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# Measuring progress toward sustainability







### Secretariat of the International Human Dimensions Programme on Global Environmental Change

CAMBRIDGE

## Inclusive Wealth Report 2014 Measuring progress toward sustainability

The *Inclusive Wealth Report 2014* is a joint initiative of the UN University – International Human Dimensions Programme (UNU-IHDP) and the UN Environment Programme (UNEP), in collaboration with the UNESCO-Mahatma Gandhi Institute of Education for Peace and Sustainable Development (UNESCO-MGIEP), ASCENT Africa Sustainability Centre, the Malaysian Industry-Government Group for High Technology (MIGHT), Science to Action (S2A), the Ministry of Environment – Government of Japan, the UN University – Institute for the Advanced Study of Sustainability (UNU-IAS), and endorsed by the Science and Technology Alliance for Global Sustainability.

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Measuring progress toward sustainability







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### Foreword

National accounts are descriptors. They describe the state of an economy and form the raw material for both assessing performance and prescribing policy. National accounts are meant to contain the kinds of information that are essential for economic evaluation. The system of national accounts currently in use throughout the world, however, suffers from extreme narrowness. Vast quantities of information relevant for economic evaluation do not appear in them. Some don't because the appropriate data are hard, even impossible, to collect; but others don't because until recently the theory and practice of economic evaluation didn't ask for them. The demand for "green national accounts" has arisen because of a growing recognition that contemporary national accounts are an unsatisfactory basis for economic evaluation. The qualifier, "green", signals that we should be especially concerned about the absence of information on society's use of the natural environment.

#### The IWR 2012

The inaugural publication on inclusive wealth (the IWR 2012), issued jointly by UNU-IHDP and UNEP, provided an account of what would ideally be needed for a comprehensive set of national accounts. The procedures recommended there were put to work in estimating changes in inclusive wealth per capita during 1990-2008 in 20 countries that represent various stages of economic development. The publication revealed that national governments and international agencies ought to go beyond even green national accounts, by reclassifying certain classes of goods and services and adding others that are currently missing. For the present, the ideal can be approximated at best crudely, which is what the IWR 2012 achieved. Data on many items that ought to be included will of necessity appear only in physical terms for some time yet, while many other items of significance (ecosystems other than forests, for example) will continue to be missing even in physical terms. Economic evaluation inevitably involves cutting corners. But it is essential for good practice to know where the corners that are being cut happen to be. That is why

the authors of the IWR 2012 went extensively into the conceptual foundations of economic evaluation.

The IWR 2012 offered a set of capital accounts for each of the 20 countries on its list, akin to balance sheets of private firms. Inclusive wealth is the social value of an economy's capital assets. The assets comprise (i) manufactured capital (roads, buildings, machines, and equipment), (ii) human capital (skills, education, health), and (iii) natural capital (sub-soil resources, ecosystems, the atmosphere). Such other durable assets as knowledge, institutions, culture, religion – more broadly, social capital – were taken to be enabling assets; that is, assets that enable the production and allocation of assets in categories (i)-(iii). The effectiveness of enabling assets in a country gets reflected in the shadow prices of assets in categories (i)-(iii). For example, the shadow price of a price of farming equipment would be low in a country racked by civil conflict, whereas it would be high elsewhere, other things being equal.

The system of national accounts (SNA) that are still being developed by the United Nations and their affiliated international agencies do not yet contain several of the additions and reclassifications that were made in the IWR 2012. That is why the empirical estimates reported in the IWR 2012 were of significance. Being a first attempt, the estimates were conducted mainly with natural capital in mind. Even within that category, attention was paid to forests, land, sub-soil resources, and the atmosphere as a sink for carbon. Estimates of human capital were restricted to education, whose measurement has a long history in economics.

The present publication extends the IWR 2012 in three ways: (a) the coverage is 140 countries; (b) the basis for the estimates of education as a capital asset is the more sophisticated approach developed by Dale Jorgenson and his collaborators; and (c) health as a form of capital asset receives attention in the main body of work. Health poses special problems of estimation, so it is worth explaining why.

#### Health capital

Health is a capital asset and should be seen as a component of a person's human capital. In order to compare the relative significance of an economy's various capital assets with one another, they have to be expressed in a common currency. That common currency is typically monetary, say, dollars. But the currency could have been any chosen commodity, or a basket of commodities, for example, a basket of consumption goods. Health capital is health status expressed in that common currency.

Good health brings three benefits to a person:

- 1. It adds directly to the person's well-being (she feels good);
- 2. It enables the person to be productive (a healthy person works better and can work for longer hours than an unhealthy person);
- 3. It contributes to her longevity (a healthy person can be expected to live longer than an unhealthy person).

Items (I) and (3) are direct benefits (they constitute aspects of a good life), while item (2) is an indirect benefit (a means to a better life). It is humanity's good fortune that good health offers the three benefits jointly (they are not in competition!). Economists have developed elaborate methods for estimating the value of each type of benefit. Some involve asking people to report their willingness to pay for the benefits ("reported preference"), while others estimate the value of the benefits to people by observing their behavior ("revealed preference"). One way to estimate the combined benefit of improved health is by recording people's willingness to pay for better health (e.g., observing how much people spend on health). Some studies estimate the benefits enjoyed from item (2) by the output lost when workers are absent owing to illness (the costs of air pollution are often estimated on the basis of lost days of work owing to bronchial congestion).

Unfortunately, there are no systematic studies of items (I) and (2) that could be used to cover the I40 countries in question. The present study confines itself to item (3), by using tables that have been prepared by economists reporting the value of a statistical life in various countries. The approach is not without its weaknesses, but a first step had to be taken, and the authors of the IWR 2014 are to be applauded for inaugurating in an official publication what is likely to be a long process of evaluation of health as a form of capital asset.

That said, I do not believe that a central finding of the publication will be overturned, no matter how refined the valuation exercise becomes in future. It is that health is the most significant component of the wealth of nations. The authors show that it swamps the value of all other forms of capital assets by an order of magnitude and more. This will come as a surprise to all of us who have thought that in a reasonably well-ordered society the various forms of capital assets are on a par with one another; after all that is what the theory of economic development tells us to expect. The estimates in the IWR 2014 tell us otherwise.

Parthe Duguth

Partha Dasgupta Chair of the IWR science advisory group and Frank Ramsey Professor Emeritus of Economics at the University of Cambridge

### Preface

There can be no doubt that, over the past two decades, many countries have done much to improve their citizens' well-being. Of course, some have a better record than others, but overall the trend has been positive. Gross domestic product (GDP), although stagnant in some highly advanced economies, has risen steadily across most of the world. Human Development Index (HDI) scores have also improved for a substantial number of countries

over the same time period. A cursory glance at these two trends might suggest that we are on the right track; that we should continue with business as usual.

That first glance would be misleading. Over the past twenty years we have seen, it is true, enormous gains in economic activity and output, and indeed as well in many of the quality of life indicators comprising the HDI. Countries have spent decades chasing production, consumption, and employment at all costs as the ticket to well-being.

On the other hand, serious questions have arisen as to the equitability and – more importantly – the sustainability of those gains. As Thomas Piketty demonstrated in his groundbreaking Capital in the Twenty-First Century, inequality is steadily growing, and will continue to as long as returns on capital exceed the rate of overall growth. In the era of globalization and instant communication, such levels of inequality, both within and across nations, are unsustainable.

Meanwhile, these gains have, as they have since the onset of Industrialization, come at a massive cost to ecosystem health, biodiversity, air quality, and climate resiliency.

One of the welcome key outcomes of the Rio+20 Conference on Sustainable Development was the agreement by countries to focus explicitly on sustainability in crafting the post-2015 development agenda. It is thus that the successors to the Millennium Development Goals will be known as the Sustainable Development Goals.

But how will we know when we are developing sustainably?

GDP growth still dominates policy planning, implementation, and evaluation for countries of all levels of development. Yet we have no way of knowing whether that growth is sustainable and inclusive – whether the activities that generate that growth will be possible in five years or fifty; whether they enrich the few at the expense of the many. Countries have spent decades chasing production, consumption, and employment at all costs as the ticket to well-being. But there is more to well-being than GDP, and it is time countries have approached policy planning strategically, and over the long term.

We hope that policy-makers will see the IWR 2014 as a useful tool, and as encouragement ... We have seen, since the seminal Brundtland Report in 1987, successive efforts call for audacity and ambition in tackling sustainability, but with only limited success. We will continue to see only limited success so long as our definitions of economic success and socioeconomic well-being continue to be based on GDP.

The case against GDP as a metric for economic success and socioeconomic well-being can be distilled into three main points: The first relates to the extent to which income alone is conflated with well-being. Although it is undoubtedly a necessary condition for well-being, it is not a sufficient one. As the World Bank's Voices of the Poor study found, poor people themselves define well-being not only in terms of income, but as "peace of mind, … belonging to a community, … safety, … [and] good health", among others.

Second, GDP measures gains in production and output at market prices, but ignores the environmental externalities produced through the production process. Nor does GDP reflect scarcity arising from dwindling natural resources, which are often public goods with no market prices.

Third, GDP represents flows only for a specified, generally short, time period. It does not provide information on the state of those capital stocks necessary to generate the income measured. Equally important, it provides no insight into whether those capital stocks – what we call inclusive wealth – are sufficient to generate consumption flows for future generations.

The Sustainable Development Goals are thus destined for only limited success as long as we are missing an adequate framework to measure progress, and do so in an integrated and holistic manner.

The Inclusive Wealth Report (IWR) aims to provide a comprehensive overview of the status of capital stocks of three key assets for nations. These assets are tracked over the past 21 years, and the sustainability implications of trends and changes in these assets are appraised. The report does not attempt to provide a comprehensive overview of human well-being. Instead, it provides guidance and insight for policy-makers on how their economies are generating income, how depreciation and reinvestment are affecting capital stocks, and whether system trajectories are sustainable.

The IWR 2014, while still suffering from incomplete data in some areas, is a significant improvement over the IWR 2012 in

both breadth and depth, particularly in the areas of education and health capital stocks. We hope that policy-makers at the international, national, and state level will see the IWR 2014 as a useful tool, and as encouragement to take the steps necessary to close gaps in data and to utilize the inclusive wealth accounts presented in the report as guidance.

We acknowledge that it may be early to use the report for practical policy-making; however, this was also the case 60 years ago, when nations began designing economic policies based on an incomplete set of GDP accounts. We are confident that countries will recognize the need for a comprehensive and integrated picture of the three pillars of sustainability, and the benefit of a tool to monitor and assess it. The report, however, should not only be useful for policy-makers but also our education systems, educators, and students – providing an understanding of the productive base available to societies and how it has to be managed to ensure sustainability of human well-being. We hope countries find the IWR 2014 useful as they gather in 2015 to finalize the post-2015 development agenda and the Sustainable Development Goals. It is time to plan – and measure – the future we want holistically, and inclusively.

v

#### Anantha Duraiappah

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#### **IWR** Team

As an initiative hosted by UNU-IHDP Secretariat, the IWR has involved many people that devoted much time and energy. Anantha Duraiappah, Executive Director of IHDP, who conceived of the idea for an IWR as early as 2008, took the lead as the IWR Director. Pablo Muñoz, Academic Officer at IHDP, who coordinated and oversaw scientific inputs to the report, is the IWR's Science Director. Elorm Darkey, Cecilia Fernandes, and Kira Petters provided analytical and quantitative support. We are equally grateful to our interns who devoted their time to this project, including Sergio de Marco and Muzaffar Yunusov. Special thanks to Katja Cloud and Louise Schenk, art and layout designers, as well as our consultants John Tkacik and Carmen Scherkenbach. We are also grateful to Sabrina Zwick for her technical and logistical support. And we would like to thank Terry Collins for helping disseminate the report through press releases and other media forums.

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### Abbreviations

| BAU      | business as usual                    |
|----------|--------------------------------------|
| BMI      | body mass index                      |
| CGE      | computable general equilibrium       |
| CLA      | physical amount of total cropland    |
|          | area of country                      |
| CO2      | carbon dioxide                       |
| DSB      | Dispute Settlement Body              |
| ECE      | estuarine and coastal ecosystems     |
| EDF      | expected damage function             |
| EGP      | Egyptian Pound                       |
| EIA      | Energy Information                   |
|          | Administration                       |
| EPA      | United States Environmental          |
|          | Protection Agency                    |
| ESVD     | Ecosystem Service Valuation          |
|          | Database                             |
| ESW      | ecosystem service wealth             |
| EU       | European Union                       |
| EU KLEMS | EU level analysis of capital (K),    |
|          | labour (L), energy (E), materials    |
|          | (M) and service (S) inputs on a      |
|          | detailed activity level: statistical |
|          | and analytical research project to   |
|          | analyse productivity and growth      |
|          | across Europe                        |
| FAO      | Food and Agricultural                |
|          | Organization of the United           |
|          | Nations                              |
| GBM      | geometric Brownian motion            |
| GDP      | gross domestic product               |
| GHG      | greenhouse gas                       |
| GIS      | geographical information systems     |
| GTAP     | Global Trade Analysis Project        |
| GTAP     | Global Timber and Forestry Data      |
|          | Project                              |
| HAD      | High Aswan Dam                       |
| нс       | human capital                        |
| HDI      | Human Development Index              |
| HS       | crop classification                  |

| HTA health technology assessment       |
|--|
| IALS International Adult Literacy      |
| Survey                                 |
| IEA International Energy Agency        |
| IW inclusive wealth                    |
| IWI Inclusive Wealth Index             |
| IWIadj Adjusted Inclusive Wealth Index |
| IWR Inclusive Wealth Report            |
| KPMG Klynveld Peat Marwick Goerdeler   |
| MA Millennium Ecosystem                |
| Assessment                             |
| MW megawatt                            |
| NAFSA Association of International     |
| Educators                              |
| NC natural capital                     |
| NCC Natural Capital Committee          |
| NDP net domestic product               |
| NIA national income account            |
| NPV net present value                  |
| NRC National Research Council          |
| NTFB value of non-timber forest        |
| benefits                               |
| NTFP non-timber forest products        |
| OECD Organisation for Economic         |
| Co-operation and Development           |
| ONS Office of National Statistics      |
| PC produced capital                    |
| PCE personal consumption               |
| expenditure                            |
| PIAAC Programme for International      |
| Assessment of Adult                    |
| Competencies                           |
| PIM perpetual inventory method         |
| PISA Programme for International       |
| Student Assessment                     |
| PLA physical amount of pastureland     |
| area available                         |
| PPI per capita income, adjusted by a   |
| private consumption                    |
| PPP purchasing power parity            |
| REDD Reducing Emissions for            |
| Deforestation and Degradation          |
| RICE Regional Integrated               |
| Climate-Economy                        |
| RPA rental price per hectare           |
| SEEA System of Environmental and       |
| Economic Accounts                      |

- SEPA ..... State Environmental Protection Administration
- SNA ..... System of National Accounts
- STEM ..... science, technology, engineering, and mathematics
- TEEB ..... The Economics of Ecosystems and Biodiversity
- TFP..... total factor productivity
- UN ..... United Nations
- UN-DESA. . United Nations Department of Economic and Social Affairs
- UNDP ..... United Nations Development Programme
- UNECE . . . . United Nations Economic Commission for Europe
- UNECE CES United Nations Economic Commission for Europe -Conference of European Statisticians
- UNEP..... United Nations Environment Programme
- UNESCO . . United Nations Educational, Scientific and Cultural Organization
- UN-OWG . . United Nations Open Working Group
- UNU-IAS . . United Nations University Institute for the Advanced Study of Sustainability
- UNU-IHDP United Nations University International Human Dimensions Programme on Global Environmental Change
- USSR..... Union of Soviet Socialist
  - Republics
- VARG..... The Value at Risk or Gain
- VLS..... value of a statistical life
- VLSY ..... value of a statistical life year
- WAVES .... Wealth Accounting for Valuation of Ecosystem Services
- WCL..... wealth in cropland
- WDR ..... World Development Report
- Wha ..... wealth per hectare
- WPL..... wealth in pastureland
- WTO ..... World Trade Organization
- WTP..... willingness to pay

# Executive summary

#### The goals of the IWR 2014

Structure and content of the IWR 2014

The primary objective of the 2014 Inclusive Wealth Report (IWR 2014) is to provide quantitative information and analysis on long-term trends in global inclusive wealth (IW), and in doing so paint a picture of how nations are performing in their efforts to sustainably improve the well-being of their citizens.

Another objective of the report is to further drive global efforts toward improving conceptual understanding – and quantitative evaluation – of the components of inclusive wealth that remain all-too poorly understood: natural capital and human capital.

The IWR 2014 strives to cement the role of the Inclusive Wealth Index (W) as the leading comprehensive indicator for measuring nations' progress on building and maintaining inclusive wealth – a central pillar of the sustainability agenda – and gauging global sustainability as part of the post-2015 development agenda as outlined in the Sustainable Development Goals. The IWR 2014 is presented in three parts and eight chapters, each beginning with key messages.

Part I comprises two chapters. Chapter I presents the empirical computations of inclusive wealth for I40 countries over the period of I990 to 2010. Particular attention is paid to changes and trends in inclusive wealth, and respective changes across human, natural, and produced capital, as well as a comparative analysis with those of GDP and HDI. Chapter I provides an analysis of per capita trends, demonstrating the role population growth plays in sustainability.

Chapter 2 provides basic policy guidance for improving the inclusive wealth of a country, and offers initial ideas on how the IWR can be utilized to address typical policy challenges, such as energy and agricultural policy. The chapter also discusses some of the pitfalls of present health investment policies, and suggests gaps that the IWR might address. The chapter concludes with suggestions on how to revise national accounts to include wealth accounts.

Part II of the report provides a detailed analysis of human capital wealth accounts. Chapters 3 and 4 offer in-depth reviews, as well as recommendations, on methodologies for generating education wealth accounts, and suggest using the lifetime income approach to measure human capital, which uses information on gender, demography, and age, among other categories, to compute the contribution of education to sustainable development.

Chapter 5 focuses on the theoretical model for computing health wealth accounts. The chapter gives insights into the contribution of health to human well-being, and the ways in which health is valued as a capital asset. There is considerable controversy involved in using valuation methods to value human health, some of which are discussed in the chapter. The authors suggest using the Value of a Statistical Life (VSL) for health wealth accounts, and provide some initial estimates for a small number of selected countries (not included in final WV country calculations).

Part III contains three chapters. Chapters 6 and 7 cover advances made in computing natural capital, while Chapter 8 describes how inclusive wealth can be used for project evaluation utilizing scenario analysis. Chapter 6 focuses especially on forest accounts, and explains improvements in calculations based on updated values for non-timber forest goods and services. These values were compiled from The Economics of Ecosystems and Biodiversity (TEEB) and Ecosystem Service Valuation Database (ESVD). Chapter 6 also recommends further research on generating a more complete computation of ecosystem services provided by forests, particularly with regard to carbon sequestration. For instance, countries could use the marginal contribution from a unit of forest maintained to inclusive wealth as a price for maintaining the forest for carbon sequestration.

Chapter 7 discusses several contentious issues involved in developing valuation estimates for ecosystem services, including methodologies for assuming benefit transfer across specific areas, or scaling up values to generate national-level figures. The final chapter of Part III provides an illustrative example for using the IW framework in project evaluation. The authors describe a model whereby a social cost-benefit analysis is computed for a project relating to investment in produced capital. The analysis makes an important contribution to existing project evaluation techniques by explicitly addressing the inter-linkages that occur across the various capital stocks. The chapter offers two case study examples, focusing on how infrastructure investments impact natural capital and health capital, respectively. Although the model looks retroactively at past projects, it offers insights into use for future scenario building that can inform investment decisions, in particular investments in produced capital.

### Key messages

#### PART I

#### **CHAPTER 1**

#### Accounting for the inclusive wealth of nations: Key findings of the IWR 2014

- Chapter I utilizes the Inclusive Wealth Index per capita as an indicator of inter-temporal human well-being to assess nations' economic progress within the context of sustainable development. The chapter covers 140 countries over the time period between 1990 and 2010.
- Empirical evidence shows average positive growth in per capita inclusive wealth and thus progress toward sustainable development in 85 of the 140 countries evaluated (approximately 60 percent). Gains in inclusive wealth were in general lesser than those in GDP and HDI: 124 of 140 nations (89 percent) experienced gains in GDP, while 135 of 140 (96 percent) showed improvement in HDI over the same period.
- Human capital is the foremost contributor to growth rates in inclusive wealth in 100 out of 140 countries. In 28 countries produced capital was the primary contributor. On average, human capital contributed 54 percent of overall gains in inclusive wealth, while produced capital contributed 33 percent and natural capital 13 percent.
- Population growth and natural capital depreciation constitute the main driving forces of declining wealth per capita in the majority of countries. Population increased in 127 of 140 countries, while natural capital declined in 127 of 140 countries. Although both factors each negatively affect growth in wealth, changes in population were responsible for greater declines.

- Produced capital, the capital type for which by far the most exhaustive (and reliable) data exists, represents only about 18 percent of the total wealth of nations. The remaining capital types, which together constitute 82 percent of wealth (54 percent in human capital and 28 percent in natural capital), are currently treated as, at best, satellite accounts in the System of National Accounts.
- After adjusting for carbon damage, oil capital gains, and total factor productivity, the number of overall progressing countries drops from 85 to 58 of 140 counties (41 percent). Results show that all three factors negatively affected inclusive wealth in most of countries; of the three, total factor productivity adjustments had the greatest negative effect.

#### CHAPTER 2

#### The IWR and policy lessons

- Chapter 2 provides basic policy guidance for improving the inclusive wealth of a country based on the findings and lessons of the Inclusive Wealth Report.
- Countries striving to improve their citizens' well-being – and do so sustainably – should reorient economic policy planning and evaluation away from targeting GDP growth as a primary objective toward incorporating inclusive wealth accounting as part of a sustainable development agenda.
- Investments in human capital in particular
  education would generate higher returns
  for IW growth, as compared to investments
  in other capital asset groups, in countries
  with high rates of population growth.
- Investments in natural capital, in particular agricultural land and forest, can produce a twofold dividend: First, they can increase IW directly; second, they can improve agricultural resiliency and food security to accommodate anticipated population growth.
- Investments in renewable energy can produce a triple dividend: First, they can

increase IW directly by adding to natural and produced capital stocks; second, they improve energy security and reduce risk due to price fluctuations for oil-importing countries; third, they reduce global carbon emissions and thus carbon-related damages.

- Investments in research and development to increase total factor productivity, which decreased in 65 precent of countries, can immediately contribute to growth in inclusive wealth in nearly every country.
- Countries should expand the asset boundary of the present System of National Accounts (SNA), which currently captures only 18 percent of a country's productive base, to include human and natural capital, which are now measured only through satellite accounts, if at all.

#### PART II

#### **CHAPTER 3**

# Human capital measurement: a bird's eye view

- Chapter 3 explores concepts and methodologies of measuring human capital for the purposes of inclusive wealth accounting.
- Measuring human capital can serve many purposes: it can help one better understand what drives economic growth; assess the long-term sustainability of a country's development path; measure the output and productivity of the educational sector; and facilitate informed discussions on social progress and well-being. In spite of this, human capital has not yet been included within the asset boundary of the SNA.
- The multifaceted nature of the concept of human capital creates substantial challenges for its measurement. By focusing on formal education and economic returns for individuals – rather than on human capital in general and all the benefits due to human

capital investment – we can begin from an empirically manageable and practical point of departure.

- All existing approaches to measuring human capital have both advantages and disadvantages. However, the monetary measures generated from the cost-based and the income-based approaches should arguably be designated a "core" status. One reason for this is to enable direct comparison of figures with those for traditional produced capital covered by the SNA, the construction of which is a primary task of national statistical offices.
- Drawing on country experiences and international initiatives in the field of human capital measurement, one may conclude that an international trend is emerging toward an income-based approach, specifically the lifetime income approach. Estimates based on this approach can be used to assess the relative contribution of a range of factors (demographic, education, and labor market) to the evolution of human capital, and facilitate corresponding policy interventions.
- Despite significant progress having been made, there remain considerable challenges regarding data availability, and detailed methodological choices inherent in applying monetary measures. Further research should therefore be encouraged, including toward the compilation of quality data for use in international and inter-temporal comparisons; the construction of experimental satellite accounts, in order to better understand and reconcile the discrepancies between estimates based on the cost-based and the income-based approaches; and, eventually, toward incorporating human capital measures into the SNA in the future.

#### **CHAPTER 4**

# Human capital: country estimates using alternative approaches

- Chapter 4 reviews and analyzes data on human wealth accounts from the IWR 2014 country sample.
- Human capital is critical to individual and societal well-being.
- The educational attainment of a country's younger cohort is frequently higher than the educational attainment of the older cohort; high levels of youth educational attainment correlate to high potential for improved well-being and economic growth in the future.
- Human capital indicators which depend solely on educational attainment information fail to capture the full potential of a country's population.
- Human capital measures including information on present and future demographic trends, education, and wage or income components are essential for appropriate policy formulation and analysis.

#### **CHAPTER 5**

#### Health capital

- Chapter 5 discusses the importance of health to human well-being and the concept of health capital.
- Health is an essential characteristic of human well-being.
- Health capital is an important part of inclusive wealth.
- The economic model of health capital presented in this chapter allows health to affect human well-being through three distinct channels: direct well-being, productivity, and longevity.
- Most health capital services influence human well-being directly rather than through the production of goods and services that are counted in GDP.

- In the absence of better estimates of the direct and productivity effects, gains in life expectancy should be used as the primary measure of health capital.
- Annual gains in health capital in the United States are worth approximately US\$10,000 per person in monetary terms.

#### PART III

#### **CHAPTER 6**

#### Forest wealth of nations

- Chapter 6 focuses on forest ecosystems as a key component of natural capital, and looks at current efforts to internalize benefits from forest ecosystem services to well-being.
- Forest ecosystems provide a huge range of tangible and intangible benefits for human well-being. These are of immense value and represent an important component of national and global wealth.
- Demographic trends and economic growth are exerting increasing pressure on forest capital. Accounting more fully for this wealth, and how it is changing as a result of economic and social activity, is urgently required. The estimates in this chapter provide a tentative first step in this direction.
- From a global perspective, in 2010 for the selected countries, forest wealth amounted to more than US\$273 trillion. On the face of it this wealth, in absolute terms, seems concentrated in relatively few countries. However, for many other countries, forest capital remains an important component of national wealth. Many of these countries (although not all) have experienced alarming losses in forest capital over the past 20 years.
- From an accounting perspective, these losses are frequently hidden from view. It is thus essential that nations pursue better

accounting to understand quantity, quality, and distribution of forest wealth. Indeed, keeping forest wealth intact – and, moreover, investing in forests to reverse past losses – is an important pre-condition for sustaining development.

#### **CHAPTER 7**

# Challenges to ecosystem service valuation for wealth accounting

- Chapter 7 reviews the use of ecosystem service valuation for wealth accounting.
- In recent years, substantial progress has been made by economists working with ecologists and other natural scientists in valuing some ecosystem goods and services.
- However, difficulties in measurement, data availability, and other limitations still preclude the valuation of certain ecosystem services.
- There is often uncertainty associated with estimated ecosystem service values, and even more so with scaling up of local values to regional or national levels or updating these values annually, which poses problems for their use in wealth accounts.
- In the absence of reliable estimates, the temptation is to use "second-best" estimates, or to transfer values from other locations; however, such methods should be used with caution and only under specific circumstances, at the risk of generating unrealistic values.
- Progress in incorporating ecological capital in wealth accounts therefore requires developing more accurate methods of valuing ecosystem goods and services and applying them to a wider range of ecosystems.

#### **CHAPTER 8**

Using inclusive wealth for policy evaluation: the case of infrastructure capital

- Chapter 8 explores strategies and methods for using inclusive wealth in project evaluation.
- Wealth accounting to date has focused primarily on the assessment of past performance in economies, by measuring changes to produced, natural, and human capital.
- In order to use inclusive wealth for policy evaluation, we must estimate the impacts of a given policy on the trajectories of the capital stocks that comprise wealth.
- Infrastructure is an important policy domain because proposed changes to current systems affect many, if not all, capital stocks, which results in capital stock interactions and trade-offs.
- A systems view of policy evaluation is necessary in order to map and quantify these impacts and trade-offs; this can be managed using conceptual and mathematical models that capture integrated physical and economic processes.
- To illustrate how one might conduct wealthbased policy evaluation, we use two infrastructure case studies – coal-fired power generation in China and the High Aswan Dam in Egypt. The case studies rely on integrated physical and economic models to quantify capital stock impacts of past infrastructure decisions.
- Such models can be used to evaluate prospective infrastructure systems as well, although doing so requires careful consideration of future uncertainty. Scenario analysis is a useful and flexible method for incorporating uncertainty into wealth-based policy evaluation.

# Inclusive wealth: an overview

Anantha Duraiappah and Nabila Jamshed

#### Context

For more than half a century we have appraised our progress as nations on the basis of how much we produce, consume, and invest; we have measured that progress in U.S. dollars and aggregated into an easy-to-compare metric: gross domestic product (GDP). We have been working under the implicit underlying assumption that the resource base upon which this growth depends is infinite. But what if it is not – what if this growth is not sustainable? And further, what if the reality of human wellbeing is not being accurately reflected in our computations of GDP; or if our GDP growth rates are not resulting in improvements in human well-being?

The dialogue surrounding what is to become the post-2015 global development agenda has recognized the shortcomings of the present development agenda, as well as the limitations inherent in using GDP as a yardstick for progress (UNITED NATIONS 2012, UNU-IHDP AND UNEP 2012). The outcome document from the 2012 global summit, the Rio+20 United Nations Conference on Sustainable Development, calls for a paradigm shift in the way we view development and growth, and for a set of Sustainable Development Goals (SDGs) that reflect that paradigm shift. At the same time, there is growing recognition that conventional national accounting frameworks have overlooked some of the most important assets a country possesses, treating them as peripheral, rather than central to human well-being.

We require a more comprehensive framework for measuring our future progress – not necessarily to replace, but rather complement, GDP – and to reveal the full extent of a country's assets, or productive base. The shift toward sustainability as a core development pillar demands an index that can quantify, measure, and track sustainability. The concept of inclusive wealth, and an inclusive wealth indicator, is a response to these deliberations and demand.

The Inclusive Wealth Report (IWR) is a biennial effort to evaluate the capacities of nations around the world to improve their citizens' well-being, and do so sustainably for the benefit of present and future generations. The report provides a more comprehensive and accurate measure of human wealth, development, and progress. The IWR validates our suspicions that GDP is an inadequate measure for assessing long-term prosperity, and reveals education, health, and the environment as investments that will truly unleash the potential of young and interconnected populations around the world for development. The Inclusive Wealth Index (*W*) will be crucial to measuring progress toward the Sustainable Development Goals, and in the planning and evaluation of sustainable development as a policy paradigm.

#### Inclusive Wealth Index: beyond GDP and HDI

Inclusive Wealth is a tool, rather than a prescription. In the first IWR in 2012, we demonstrated that the principal pillars of the wealth of nations, human capital and natural capital, have remained largely hidden to policy-makers due to the limitations of traditional economic indices. It was discovered that the biggest returns were coming from factors not accounted by our systems of national accounts, nor, by extension, reflected in GDP.

GDP is a useful and practical tool for measuring economic production, but it does not impart any information on the state of the resource base upon which production relies. The 2010 Report of the French Government's Commission on Measurement of Economic Performance and Social Progress, also known as the Stiglitz-Sen-Fitoussi Commission, pointed to a number of ways in which nations were "mis-measuring" development through using GDP (STIGLITZ ET AL. 2010). These range from measurement errors and exclusion of key variables, to incomplete and misleading data. The commission echoed the warnings of Simon Kuznets – the father of GDP – of using GDP to measure societal progress (KUZNETS 1934). The underlying framework used to compute GDP are the systems of national accounts. These national accounts have in recent years made some progress toward capturing a broader picture of the economic system, in particular by extending accounting to include the environmental system (see SEEA 2013). However, the accounts measured are still flow accounts – they measure only financial and material flows over a given time period – and thus do not reflect the sustainability dimensions of the economy. It is for this reason that recent attempts to internalize environmental externalities into national accounts, such as Green GDP, still fall short of providing an indicator to understand and track sustainability.

Another effort, the Human Development Index (HDI), was created in the 1990's as an initiative to provide an alternative to GDP in measuring human development progress in terms of life

expectancy, education, and income (UNITED NATIONS DEVELOPMENT PROGRAMME 1990-2014). Although certainly illuminating, HDI still has significant – in sustainability terms – shortcomings. A primary drawback of HDI

"... a better way to size up wealth" – *The Economist* 

is its inability to adequately incorporate the ecological dimensions of sustainable development, and that it does not integrate social goods in capital accounts to complement GDP.

The W does not reject GDP. It acknowledges GDP's practicality for tracking efficiency of resource use for production, and for providing an overview of interdependencies among economic sectors held within the system of national accounts. Neither does the W aim to modify GDP to accommodate missing elements, as Green GDP initiatives attempt. The W starts from the premise that all development is conditional on the existence of several key assets, and that the total value of these assets should not be allowed to decline if human well-being is to be furthered sustainably.

The inclusive wealth framework takes a different approach to that of earlier efforts to capture a broader sense of human well-being and progress. Inclusive wealth directs its focus not on the constituents of well-being – measuring as does the HDI specific outcomes that reflect well-being – but rather the determinants of well-being, the set of "ingredients" necessary for nations to bring about those outcomes. These determinants can be found in several pools of national capital assets, or the productive base of economies.

#### Inclusive wealth and the post-2015 Sustainable Development Goals:

The outcome document of the Rio+20 UN Conference on Sustainable Development, The Future We Want, set out to establish a broader development agenda for the Millennium Development Goals (MDGs) after their evaluation in 2015 (UNITED NATIONS 2012). The international policy dialogue on the Sustainable Development Goals (SDGs) that followed is clear on the need for long-term planning to ensure achievements are not merely temporary, but strive to improve the lives of both present and future generations. The SDGs also take sustainable develop-

"... if governments could agree to use the IWI as part of their overall economic accounting, it would be a substantial step towards true sustainable development."

#### - The Huffington Post

ment from the environmental realm to include social and economic aspects. In doing so, the SDGs offer a unique and much-needed catalyst to converge economic aspirations with the social and environmental goals, and not consider them independently, as is currently the case.

Sustainable development will be about transformative shifts (UNITED NATIONS 2013), and should include, according to an Open Working Group charged with developing the draft goals, the following arenas: poverty alleviation, food security, inclusive and quality education, gender equality, water and sanitation, sustainable

energy for all, inclusive and sustainable economic growth for all, decent work, innovation, inclusive and sustainable industrialization, reduced inequality, inclusive human settlements, sustainable consumption and production, sustainable use of oceans and terrestrial ecosystems, and inclusive societies and institutions (UNITED NATIONS 2013).

The Open Working Group of the SDGs has rightly identified key issues and priorities to guide nations' and the global community's sustainable development agenda over the coming decade. The next step will be to develop a set of indicators for each of the goals and targets included in the final ratified list. In doing so, however, they must keep in mind what the predecessors of the SDGs – the MDGs – did not: that indicators should capture the interdependencies among various goals. That is, indicators must provide information pertaining to trade-offs and synergies among the goals in an integrated and holistic manner. This will allow policy-makers to understand the trade-offs and knock-on effects of prioritizing some goals over others, and the areas in which synergies can be leveraged to achieve a multiplicity of positive outcomes across several goals.

The SDGs call for "measurements of progress on sustainable development that complement GDP". The W might offer such a
tool. The W helps countries measure sustainable development within the framework of growth and prosperity, and will facilitate integration of the SDGs into the rationale of national economic growth strategies.

The W premises development on opportunity. The underlying axiom of the inclusive wealth concept is simple and elegant: changes in the overall value of all assets in a country over time must be positive if the economy is to be considered on a sustainable trajectory (see Annex I for a description of the model and the underlying prepositions and theorems). The value of the change in each asset stock is computed using the social price – commonly called by economists the shadow price – of each asset, multiplied by the change in the physical stock of that particular asset. These prices in effect reflect the weighting preference of individuals across the various capital assets.

#### The IWR 2012

The Inclusive Wealth Index was launched with the first IWR at Rio+20 in 2012, and represented the first attempt by the international scientific and policy communities to develop a framework for quantifying and tracking sustainable development, inclusive of produced, human, and natural capital. It drew upon two decades of data for 20 countries covering three types of capital to quantify and demonstrate the impact and returns of investing in them. The report, subtitled Measuring progress toward sustainability, focused on natural wealth, and offered valuable insights for development policy. The report was experimental in nature but, as Time Magazine noted, was the first serious effort to measure the true total wealth of nations.

The results from the IWR 2012 were both promising and sobering. Promising was that 19 of the 20 countries evaluated experienced positive changes in overall wealth. Still, after factoring in population, inclusive wealth growth rates per capita level turned negative for five countries. It was clear that population growth in these five countries had outpaced growth of inclusive wealth, highlighting the oft underplayed role of population growth in determining the sustainability trajectory of countries.

The sobering factor that emerged from the IWR 2012 was the status of natural capital. The IWR found that 19 of 20 countries were depleting natural capital while failing to adequately invest in rebuilding this category, despite evidence that returns on investment in natural capital far outweighed investing in produced capital – infrastructure, buildings, roads, etc. – for a majority of countries. Although 14 out of the 20 countries witnessed positive per capita growth rates on their overall asset base, growth rates

were marginal and could easily turn negative should declines in natural capital continue apace. Moreover, the costs of natural capital declines in the IWR 2012 were conservative estimates; actual growth rates might indeed have become negative had the IWR 2012 used less cautious estimates. These results reinforced our contention that the prevailing understanding of economic development must change to meet the needs of the 21<sup>st</sup> Century.

#### The IWR 2014: what is new

The IWR 2014 has been expanded from 20 to 140 countries, and the time horizon has been updated to include data from 2009 and 2010 in addition to the original 1990 to 2008 periods. While the IWR 2012 included a special focus on natural capital, the IWR 2014 does the same for human capital.

Spending on human capital has traditionally been considered as expenditure in core national accounts. The IWR 2014 makes a powerful case for treating education and other spend-

"... this impressive research project ... is the first serious attempt to measure the total wealth of the planet's richest countries." – *TIME*  ing in human capital as investments, rather than expenditures. Education has long been considered a social good, and one that is crucial for future growth; however the IWR 2014 demonstrates it is also an engine of wealth today, and puts numbers to this value. In increasingly knowledge-based economies, education's role as driver of production has become more important than ever. That role is two-pronged:

education is positively correlated to produced capital, as well as enhancing opportunity, which is at the core of human well-being.

The two main components of human capital are education and health. However, while health is a key component of human capital, we have left it out of the main human capital wealth accounts as we did for the IWR 2012. This was done for a number of reasons: First, because of the relatively high value of health capital, it dominates and skews overall inclusive wealth figures. While we are convinced that health capital is indeed valuable, the methodology used for computing health values is still under debate; until there is consensus among health economists on these methodologies, it would be inappropriate to integrate as such into overall wealth accounts.

We have, however, included in the IWR 2014 a chapter in Part II which delves into the subject, providing a detailed analysis of health capital and the challenges and opportunities it poses for the national accounts and the computation of inclusive wealth. Sample coverage for a selected number of countries is represented in Part II of this report for health capital. Our goal is to integrate health capital in the 2016 report, given continuing progress on methodologies and database construction on health.

The education component in the main inclusive wealth accounts is unchanged in 2014. This is due to a lack of available data necessary to undertake a more detailed analysis as prescribed in the chapters addressing education in Part II of the report. However, the education wealth accounts have been expanded to account for new methodologies, and in Part II calculated for a selected number of countries in which necessary data was available. The lessons learned from this exposition will help guide in the revision and updating of education accounts for the broader set of countries for the IWR 2016.

The natural capital wealth accounts have been revised with new estimates for forest accounts, which included improved estimates for forest physical accounts and updated values for non-timber forest product goods and services taken from The Economics of Ecosystems and Biodiversity (TEEB) and Ecosystem Service Valuation Database (VAN DER PLOEG AND DE GROOT 2014). In addition, Part III discusses recent advances in using new typologies for forest accounts, with special attention paid to the challenges and opportunities involved in using social prices from economic valuation methods for ecosystem services.

Total factor productivity (TFP) was treated as a residual in the IWR 2012. The estimates were taken from the Total Economy Database (CONFERENCE BOARD 2012). In the IWR 2014, TFP is still treated as a residual, but is now generated by including natural capital as an explicit factor input to the production process. This approach allows us to extract directly the contribution of natural capital toward production, and not have it be reflected implicitly in the TFP, as was the case in 2012. We were therefore able to isolate to a closer approximation the real role technological innovation and creativity played in production, as well as other implicit capital types not yet accounted in building the inclusive wealth of the country.

The final addition for 2014 is policy. We present some first attempts at interpreting the findings of the IWR 2014 into implications for national and intergovernmental policy-makers. The report also takes a first stab at using scenario analysis for specific areas, applying inclusive wealth methodology and results to guide policy-making at the project level. The inclusive wealth framework allows using a social cost-benefit approach to project design and implementation (DASGUPTA ET AL. 1972). The first attempt focuses on produced capital, but the lessons learned can be easily transferable to the other capitals.

#### Audience and structure of the IWR 2014

The primary audience of the Inclusive Wealth Report 2014 will be researchers and policy-makers. The inclusion of environmental damage in the accounts – such as damages caused by global environmental and climate change – can be useful in determining transnational compensations, and as a guide for international negotiations on trans-boundary assets.

The report will also be useful for national economic planning agencies when considering macroeconomic fiscal policies. Changes in the various capital assets and their contributions toward inclusive wealth can provide key information as to where future investments should be targeted to generate optimal returns for increasing the overall productive base of a country.

The IWR is also targeted toward the research community. The 2014 edition identifies and elaborates on a large number of areas within the framework still in need of theoretical refinement and empirical data. For instance, the IWR 2014 does not address the issue of inequality within and among nations; yet the significance of wealth as a common denominator for measuring inequalities is becoming more evident, as recently demonstrated by Thomas Piketty in Capital in the 21st Century (PIKETTY 2014). Using inclusive wealth rather than income alone can provide a more complete picture of inequality in contemporary societies across the world.

The IWR 2014 is presented in three parts. Part I comprises two chapters. Chapter I presents the empirical computations of inclusive wealth for 140 countries over the period of 1990 to 2010. Particular attention is paid to the changes in inclusive wealth, and respective changes across human, natural, and produced capital. Chapter I also compares IWR trends with those of GDP and HDI. Importantly, Chapter I also provides an analysis of per capita trends, demonstrating the role population growth plays in sustainability.

Chapter 2 provides basic policy guidance on investment strategies to improve the inclusive wealth of a country. The chapter offers some initial ideas on how the IWR can be used to address typical policy issues such as energy or agricultural policy. The chapter also discusses some of the pitfalls of present health investment policies, and suggests gaps that the IWR might address. The chapter concludes with suggestions on how to revise national accounts to include wealth accounts.

Part 11 of the report provides a detailed analysis of human capital wealth accounts. Chapter 3 and 4 offer in-depth reviews as well as recommendations on methodologies for generating education wealth accounts, eventually suggesting a specific approach (the lifetime income approach) to measure human capital, which uses information on gender, demography, and age, among other categories, to compute the contribution of education to sustainable development.

Chapter 5 focuses on the theoretical model for computing health wealth accounts. The chapter gives insights into the contribution of health to human well-being, and the ways in which health is valued as a capital asset. There is considerable controversy involved in using valuation methods to value human health, some of which are discussed in the chapter. The authors go on to suggest using the Value of a Statistical Life (VSL) for health wealth accounts, and provide some initial estimates for a small number of selected countries.

Part III of the report contains three chapters. Chapters 6 and 7 cover advances made in computing natural capital, while Chapter 8 describes how inclusive wealth can be used for project evaluation using scenario analysis. Chapter 6 focuses especially on forest accounts, and explains improvements in calculations based on updated values for non-timber forest goods and services. These values were compiled from The Economics of Ecosystems and Biodiversity (TEEB) and Ecosystem Service Valuation Database (VAN DER PLOEG AND DE GROOT 2014). Chapter 6 also recommends further research on generating a more complete computation of ecosystem services provided by forests, particularly with regard to carbon sequestration. It is a component that might be useful in the Reducing Emissions from Deforestation and Degradation (REDD) initiative. Countries might, for instance, use the marginal contribution to inclusive wealth from a unit of forest maintained as a price for maintaining the forest for carbon sequestration. Although the wealth accounts are still experimental, the potential of using the inclusive wealth accounts for this purpose is promising.

Chapter 7 discusses several contentious issues involved in developing valuation estimates for ecosystem services, including methodologies for assuming benefit transfer across specific areas, or scaling up values to generate national-level figures. The final chapter of Part III provides an illustrative example for using the inclusive wealth framework in project evaluation. The authors describe a model whereby a social cost-benefit analysis is computed for a project relating to investment in produced capital. The analysis makes an important contribution to existing project evaluation techniques by explicitly addressing the inter-linkages that occur across the various capital stocks. The chapter offers two case study examples, focusing on how infrastructure investments impact natural capital and health capital, respectively. Although the model looks retroactively at past projects, it offers insights into use for future scenario building that can inform investment decisions, in particular investments in produced capital.

The reader is encouraged to review the data and technical notes annexed at the end of the report for a more detailed discussion of the specific methods used in the IWR 2014. A brief description of the inclusive wealth framework is provided as well, giving those unfamiliar with inclusive wealth an understanding of the concepts and definitions that make up the inclusive wealth theorem for sustainability.

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Part I: What does the data say?

# Accounting for the inclusive wealth of nations: key findings of the IWR 2014

Pablo Muñoz, Kira Petters, Shunsuke Managi, and Elorm Darkey

# KEY MESSAGES

This chapter utilizes the Inclusive Wealth Index per capita as an indicator of inter-temporal human well-being to assess nations' economic progress within the context of sustainable development. The chapter covers 140 countries over the time period between 1990 and 2010.

**Empirical evidence shows positive average** growth in per capita inclusive wealth – and thus progress toward sustainable development – in 85 of the 140 countries evaluated (approximately 60 percent). Gains in inclusive wealth among the countries analyzed were in general lesser than those in GDP and HDI; 124 of 140 nations experienced gains in GDP, while 135 of 140 showed improvement in HDI over the same period.

Human capital is the foremost contributor to growth rates in inclusive wealth in 101 out of 140 countries. In 27 countries produced capital was the primary contributor. On average, human capital contributed 55 percent of overall gains in inclusive wealth, while produced capital contributed 32 percent and natural capital 13 percent. **Population growth and the depreciation** of natural capital constitute the main driving forces of declining wealth per capita in the majority of countries. Population increased in 127 of 140 countries, while natural capital declined in 116 of 140 countries. Although both factors each negatively affect growth in wealth, changes in population were responsible for greater declines.

**Produced capital, the capital type** for which by far the most exhaustive (and reliable) data exists, represents only about 18 percent of the total wealth of nations; while the remaining capital types, which together constitute 82 percent of wealth (54 percent in human capital and 28 percent in natural capital), are at best treated as satellite accounts in the System of National Accounts.

After adjusting for carbon damage, oil capital gains, and total factor productivity, the number of overall progressing countries drops from 85 to 58 of 140 counties. Results show that all three factors negatively affected inclusive wealth in most of countries; of the three, total factor productivity adjustments had the greatest negative effect.

#### 1. Introduction<sup>1</sup>

For those interested in understanding a nation's short term progress in increasing gross economic output, there is a veritable flood of raw data, macro trend research, comparative analysis, and prescriptive advice available from which to choose. For those interested in understanding a nation's progress in improving its citizens' well-being from a long term perspective, there is significantly less information available.

For many years economists and policymakers have relied primarily on gross domestic product (GDP) as a proxy indicator for gauging countries' welfare<sup>2</sup>. But the pursuit of GDP growth has, in many instances, failed to deliver the gains in overall human welfare for which many had hoped (STIGLITZ ET AL. 2009). There is growing consensus (discussed in the introduction and throughout this report) that GDP is falling short in this regard (BARTELMUS 2014). This should not be surprising. GDP is a limited, unidimensional measure of market activity.

Well-being is, on the other hand, a complex, multidimensional concept that goes far beyond material living conditions, for which GDP might in certain cases be reasonably enlisted as a valid metric. Well-being encompasses not only income but also dimensions such as educational opportunity, health and quality of life, natural landscapes and ecosystems, and social networks and relationships, among others.

This multidimensional nature introduces significant challenges for those attempting to articulate the notion into a quantitative indicator within a consistent theoretical framework. Despite the challenges, a number of efforts have attempted to fill the gaps left by GDP, and measure, evaluate, and track meaningful progress of nations in improving well-being.

Some reactions on this issue are reflected in a resolution adopted by the UN General Assembly in 2012 and enshrined in "The Future We Want" outcome document of the Rio+20 summit, which explicitly recognizes the need for broader measures of progress to complement GDP (UNITED NATIONS 2012). The report "Resilient People, Resilient Planet", published in 2012 by the United Nations Secretary-General's High Level Panel on Global Sustainability, called for the development of an indicator for sustainable development.

The United Nations has not been the only supporter of change on this front. In 2007, the European Union launched the Beyond GDP initiative, which aims to develop indicators that integrate environmental and social aspects of progress. In 2011, the OECD launched the Better Life initiative to better track progress in improving human well-being (OECD 2014). Other contributions to the debate have come from the Beyond Economic Growth initiative at the World Bank (SOUBBOTINA 2004).

At the same time, national level programs have also made headway in pushing indicators beyond GDP. Perhaps the most famous example is Bhutan's Gross Domestic Happiness indicator. India too has delved into the issue with its Report on Green National Accounts (DASGUPTA ET AL. 2013); as has the United Kingdom with the Measuring National Well-Being initiative (OFFICE FOR NATIONAL STATISTICS 2014). France has also been at the forefront of beyond GDP, and the output of its 2008 "Commission on the measurement of economic performance and social progress", the so called Stiglitz-Sen-Fitoussi Report (STIGLITZ ET AL. 2009), has provided the philosophical and theoretical underpinnings for much of the concrete progress over the past several years.

There are now a number of proposed *beyond GDP* indicators available, including (but not exclusive of): the Human Development Index (UNITED NATIONS DEVELOPMENT PROGRAMME 1990-2013); the Happiness Index (HELLIWELL ET AL. 2012, EASTERLIN 2003, KAHNEMAN ET AL. 2006, LAYARD 2005, AND OTHERS); the Genuine Progress Indicator (KUBISZEWSKI ET AL. 2013); indicators derived from the System of National

<sup>1</sup> We gratefully acknowledge comments and suggestions provided by Matthew Agarwala, Giles Atkinson, and Nick Hanley.

<sup>2</sup> We use "welfare" and "well-being" interchangeably.

Accounts 2008 (UNITED NATIONS 2009) and the System of Integrated Economics and Environmental Accounts 2013 (UNITED NATIONS ET AL. 2012); and the Better Life Index<sup>3</sup> (OECD 2014).

It is unlikely that a single indicator can provide the information needed for understanding well-being and informing long-term policy planning in the area of sustainable development. Instead, several indicators, incorporating multiple dimensions, will be necessary.

In this context, the Inclusive Wealth Report (IWR) project proposes wealth accounting as a complementary indicator of social progress and human well-being.

Wealth accounting at one level aims to fill gaps left by traditional economic indicators, some of which are derived from the System of National Accounts (SNA). For instance, GDP largely ignores biophysical changes to environmental systems caused by economic production, yet those changes often drastically affect human well-being, both directly and indirectly.

Wealth accounting also internalizes sustainability by tracking the changes in the value of a nation's capital asset stocks - its productive base - so as to understand how a its activities and policies now will impact its future opportunity to generate well-being. This concept of opportunity is central to inclusive wealth. Rather than measure the constituents of well-being - the specific outcomes and circumstances that make up quality of life for us now - inclusive wealth measures the determinants of well-being - the capital stocks upon which nations rely to bring about those outcomes. This is a key pillar of intergenerational sustainability, and should be included in any assessment of a nation's economic performance.

Inclusive wealth accounting also allows for analysis and understanding of trade-offs and

other relationships among the various capital stock trajectories as an economy evolves, linking macroeconomic variables such as consumption and investment in several assets, including environmental capital stocks.

The IWR tracks changes in three categories of capital assets: human capital, produced capital, and natural capital. While some of the components of these capital stocks, such as manufactured capital, are extensively accounted for within the SNA, other important assets, including natural capital, human capital, and health capital, have been poorly or not at all assessed and reflected. To be sure, the SNA recognizes the importance of these capital assets, and recommends accounting not only for produced capital but also for some aspects of natural capital, such as fossil fuel and mineral deposits and forest cover. The SNA also considers the quantification of human capital of relevance; its measurement, however, falls outside the boundaries of analysis (UNITED NATIONS 2009), and is therefore excluded from nations' balance sheets.

In this chapter, we present the key findings from the wealth accounts of 140 countries in 2014. While the overall methodology for computing the Inclusive Wealth Index (W) is largely the same as it was in 2012, it has been expanded and improved in several ways worth noting: First, the country sample has been expanded from 20 countries - representing 72 percent of world GDP and 56 percent of the global population - to 140 countries, together representing 99 percent of GDP and 95 percent of the global population. Second, the time period reviewed has been extended by two years, stretching now from 1990 to 2010. Third, the methodologies for calculating several components of natural capital have been updated and improved, particularly with regard to calculating contributions of forest ecosystem services toward human well-being. Fourth, proxy variables used to represent contributions of inputs for which data is lacking, total factor productivity (TFP) estimates, have been improved by adding in the

<sup>3</sup> The index focuses on eleven key aspects of life, including not only income and jobs, but also housing, environment, social network, work-life balance, personal security, education, health, whether citizens feel part of the democratic process, and their level of satisfaction with life in general.

# FIGURE 1 A three-capital model of wealth creation



production function natural capital along with produced and human capital.

#### 2. Methods

The chapter is organized into four sections, in addition to the introduction: Section 2 presents a brief overview of the theory and methodology behind the Inclusive Wealth Index. Further details regarding the architecture of the index can be found in the conceptual and methodological annexes (Annex 1 and 2) at the end of the report. Section 3 provides some of the key results and upshots stemming from the inclusive wealth calculations. Further details on the data inputs presented in this section are located in Annex 3, while the full data set is available at www.inclusivewealthindex.org. Section 4 discusses some of the challenges and limitations of the study. Section 5 contains concluding remarks.

The Inclusive Wealth Index is premised on a body of work<sup>4</sup> (e.g., DASGUPTA 2009, ARROW ET AL. 2012, HULTEN 1992, HEAL AND KRISTRÖM 2005) for linking the discounted present value of all future consumption possibilities to the total worth of the capital assets (or wealth) in an economy.

This implies that present decisions involving countries' asset portfolio management have consequences for a population's future consumption opportunities. In wealth accounting, capital assets are both the inter-temporal means of production, as well as a direct source of human well-being by filling the current population's consumption needs (see Figure 1). Because wealth accounting is concerned with sustaining well-being over time, the approach is especially relevant for studying sustainability issues.

The wealth accounting framework can be thought as a supply-driven accounting system, in

<sup>4</sup> See also the conceptual framework in Annex 1.

which wealth is the constraint of inter-temporal consumption; it follows thereby that changes (positive or negative) in the total worth of capital assets will increase (or decrease) inter-temporal welfare. For the IWR we group capital assets into the following three categories: human capital (HC), produced capital (PC), and natural capital (NC). These three capital asset groups can all be broken down into various subcomponents; for the IWR 2014 we look especially closely at human capital (see Figure 2). For the IWR 2014, as in 2012, we have excluded health capital from the human capital asset group as this asset, according to existing methodologies, so dominates wealth capital that any changes, even modest, would disproportionately influence overall trends in capital asset stocks. Human capital accounts thus reflect only the education component of a population.

The basket of assets comprising the resource base is calculated by using the marginal contribution of each asset type to social welfare, which is represented by the social (or shadow) price of the assets under evaluation. Such prices act as a weight in the metric, resulting in the measure of wealth, or Inclusive Wealth Index (WV):

#### EQUATION 1

Wealth =  $P_{nc} x PC + P_{hc} x HC + P_{nc} x NC$ 

while the changes in wealth, also called inclusive investment, are captured by assessing changes in capital assets over time:

#### EQUATION 2

 $\Delta Wealth = Inclusive Investment = P_{pc} x \Delta PC + P_{hc} x \Delta HC + P_{nc} x \Delta NC$ 

It is worth highlighting that changes in W are driven solely by changes in the productive base (the entirety of capital assets) of the economy, as prices (the weights in the index) are assumed to be constant and represented in most of cases by the average price of the time span under evaluation in this study. In this way, changes in wealth are induced only by real changes in the physical amount of the various capital forms, and not simply by price fluctuations, which may be subject to contingent situations. Over the long run, however, price fluctuation may nonetheless be important for building other capital types. We therefore suggest accounting separately for those capital gains (or losses) due to price changes.

We represent this by, for instance, accounting oil capital gains (and losses) due to fluctuations in oil prices separately from physical changes in the resource base. As we see in Sections 3.4, this is particularly relevant for oil-extractive economies such as Saudi Arabia, Kuwait, or Iraq, as they are positively affected by the changes in oil prices. Conversely, net oil importing countries tend to see welfare negatively affected from rises in oil prices, as their basket of consumed goods and services diminishes.

In addition to oil capital gains, there are two other items in our accounts that we keep separate from W: carbon damage, and total factor productivity.

The decision to exclude damage experienced by nations as a result of increased CO2 concentrations in the atmosphere was made because those effects do not themselves constitute an asset of a particular country under analysis. Rather, these damages are negative externalities stemming from total global fossil fuel combustion. While such damages could be counted for as a component of the shadow price of burnt fossil fuel, there are complications inherent in this route, such as the reality that the countries responsible for the majority of carbon discharge are usually not experiencing proportional damages to well-being. Instead, we account for the externalities on the basis of how the damages are distributed across nations once the emissions are concentrated in the atmosphere. For further explanations as to how the calculations of these damages are allocated to different countries, refer to Annex 2 as well as Arrow et al. (2012).

Total factor productivity (TFP) is the third component accounted for separately from the resource base and W. This is because TFP does not explicitly capture the contribution of one particular asset, but instead captures the contributions of several "missing" assets – assets

#### FIGURE 2

Schematic representation of the Inclusive Wealth Index (*W*) and the Adjusted Inclusive Wealth Index (*W*adj).

Note: assets are added by evaluating their changes at their social (shadow) price.



not explicitly accounted for in our wealth calculations. TFP measures changes in aggregated output (GDP) that cannot be explained by the growth rate of observable inputs or capital assets.

In the IWR 2014 accounts we estimate TFP growth rates, isolating not only the contribution of produced and human capital to economic outputs – as is done in, for instance, the Penn World Tables (FEENSTRA ET AL. 2013) or the Total Economy Database (CONFERENCE BOARD 2014) – but also taking into consideration the contribution of natural capital to economic growth.

Thus, our TFP growth measures represent the contribution of residual production factors to GDP growth after the three capital asset groups (human, produced, and natural capital) are accounted for. In other words, the same productive base of a country can lead to an increase (decrease) in aggregate output over time due to productivity changes in the use of resources. These TFP estimates, including the contribution of natural capital, represent new developments in our wealth accounts; the IWR 2012 estimates were taken from the Total Economy Database (CONFERENCE BOARD 2012). Combining the last three factors (oil capital gains, carbon damage, and TFP) with our three capital asset groups leads us to our second measure of wealth: the Adjusted Inclusive Wealth Index (Wadj). Figure 2 illustrates the full array of capital assets accounted for within the Inclusive Wealth Index, as well as the three items integrated into the Adjusted Inclusive Wealth Index. As reflected by Equation (1), the sum of all capital assets and factors of the indices outlined in Figure 2 is found by evaluating the changes of the capital assets at the marginal shadow price.

For the IWR 2014 we have also introduced updates to the forest accounts, which affect estimates of timber and non-timber forest resource benefits. The main change is to the boundary utilized to define forest resources: specifically, we excluded cultivated forest from the forest measures. The main reason for this is that the activity of cultivating a forest – for which labor is required – is considered a production activity in the System of National Accounts (UNITED NATIONS 2009). Further, such cultivated forest, and the corresponding changes in stock over time, are part of produced capital, as well as gross capital formation, respectively. Cultivated forest is therefore accounted for under produced capital, while our revised forest accounts reflect naturally regenerated forest.

The second update in the forest accounts refers to estimates of non-timber forest benefits. In the IWR 2012, calculations were primarily based on work carried out by Lampietty and Dixon (1995), which reflected the monetary value of several ecosystem services stemming from studies carried out in the early 1990s. Since then, there has been a considerable amount of work devoted to calculating monetary values of ecosystem services. An exhaustive summary of work conducted over the past two decades can be found in the Ecosystem Service Valuation Database (VAN DER PLOEG AND DE GROOT 2010). For the IWR 2014, we make use of this database to update our figures for economic benefits of provisioning (excluding timber<sup>5</sup>), regulating, and recreational services resulting from forest area. Additional details are presented in Annex 2.

#### 3. The inclusive wealth of nations

3.1 Measuring performances based on changes in wealth

This subsection explores one of the key questions of this report: have nations been expanding or depleting their inclusive wealth over the past several decades? The direction of such changes in wealth correlates positively with the movements of inter-temporal welfare, and therefore its relevance. In the following analysis, we show changes in wealth for the 140 nations under examination. In addition, we illustrate how these changes in wealth vary when resources are measured on a per capita basis.

As we will show for several countries, the aggregated accumulation of wealth moves at a slower pace than population growth, leading to negative per capita growth rates in wealth for a considerable number of countries, including some which otherwise experienced absolute gains in wealth.

We estimated changes in wealth by calculating annual average growth rates in wealth and population. In some instances, which are explicitly pointed out, the information is presented in terms of the changes with respect to a base year of analysis that in our case refers to 1990.

Our estimates show that 128 of the 140 countries assessed (91 percent) experienced a positive annual average growth rate in wealth, while the remaining 12 countries exhibited negative growth (see Figure 3-a). On a per capita basis, the number of countries showing positive growth rates in wealth fell from 128 to 85 (60 percent). Per capita, 55 countries experienced negative growth rates in W (40 percent) (see Figure 3-b). Figure 3 also shows changes in per capita wealth after adjusting by TFP, damages from climate change, and oil capital gains (see Figure 3-c). Only 58 of the 140 counties experienced an increase in Adjusted Inclusive Wealth. For further details on the results of Wadj, refer to Sections 3.4.

In Figure 4 the above trends in wealth and wealth per capita are further investigated by identifying those countries, and their region of precedence, that turn into negative wealth growth.

Of the twelve countries with negative W growth rates (Figure 4: quadrants II and III), ten are negative both in absolute terms as well as per capita (quadrant III). In two of twelve countries – Russia and Ukraine – figures improve when population is factored in (quadrant II), due to both countries having experienced negative population growth over the time period. As the population decreases in these countries, there are relatively more resources available per person in relation to the base period. Moldova experienced negative growth in per capita W despite having also had a declining population.

Ten nations (Albania, Armenia, Bulgaria, Croatia, Estonia, Hungary, Kazakhstan, Latvia, Lithuania, and Romania) had negative population growth but positive changes in wealth;

<sup>5</sup> Benefits resulting from timber are measured separately by using a different methodology – see Annex 2 for further details.

# FIGURE 3

Annual average growth rates in W, W per capita and Wadj for the 140 countries assessed in the IWR 2014 during the time period between 1990 and 2010



FIGURE 3 a: Growth in Inclusive Wealth Index

FIGURE 3 b: Growth in Inclusive Wealth Index per capita



FIGURE 3 c: Growth in Adjusted Inclusive Wealth Index





hence, growth in wealth remained positive in per capita terms.

Out of the 128 countries with positive absolute growth in wealth (quadrant 1 and IV in Figure 4), 83 also experienced per capita growth in wealth (quadrant 1 in Figure 4). The remaining 45 countries experienced declining wealth on a per capita basis (see quadrant IV in Figure 4).

3.2 On the contribution of human, produced, and natural capital to  $W^{\rm 6}$ 

This subsection breaks down the contributions of each capital asset group to the total inclusive wealth average growth rates (see Figure 5). In particular we look at trajectories of the individual capital asset groups, which do not always reflect the trajectories of total inclusive wealth.

In 137 nations, human capital experienced positive growth during the period of 1990 to

2010. In the case of produced capital, 132 of 140 experienced positive growth. Natural capital

accounts experienced positive growth in only 24 countries<sup>7</sup> (see Figure 5).

Per capita, 138 of the 140 countries assessed experienced positive contribution in human capital to inclusive wealth; 117 of 140 exhibit a positive contribution in produced capital; whilst the contribution of natural capital is positive for only 13 nations (see Appendix 1).

Regarding the magnitude of the average country<sup>8</sup> contribution of each capital category to the growth rates in inclusive wealth, the general trend shows that human capital has been the major contributor (55 percent), followed by produced capital (32 percent), and natural capital (13 percent). This pattern can also be seen at regional and sub-regional levels, with Europe being the only exception. In particular in Eastern and Northern Northern

America

Oceania

<sup>6</sup> In the analysis we still use the average annual growth rates in inclusive wealth per capital as our reference indicator to investigate the changes in wealth.

<sup>7</sup> There are two countries with zero growth rates included in this group.

<sup>8</sup> This is an average across countries, where every nation is equally weighted.





#### TABLE 1

Relative contribution (in percentage) of human, produced, and natural capital to W growth by sub-regions, regions, and total world average

Note: The figures represent the average relative contribution by asset category of those countries comprising the (sub-) region to growth in W. Contributions with negative sign as in the case of natural capital are taken in absolute numbers.

|                                 | Human<br>Capital | Produced<br>Capital | Natural<br>Capital |
|---------------------------------|------------------|---------------------|--------------------|
| Africa                          | 62               | 20                  | 19                 |
| Eastern Africa                  | 56               | 24                  | 20                 |
| Middle Africa                   | 47               | 13                  | 40                 |
| Northern Africa                 | 57               | 29                  | 14                 |
| Southern Africa                 | 66               | 27                  | 7                  |
| Western Africa                  | 72               | 12                  | 15                 |
| Asia                            | 54               | 32                  | 14                 |
| Eastern Asia                    | 29               | 56                  | 15                 |
| South-Central Asia              | 60               | 27                  | 12                 |
| South-Eastern Asia              | 46               | 37                  | 17                 |
| Western Asia                    | 61               | 26                  | 13                 |
| Europe                          | 44               | 50                  | 6                  |
| Eastern Europe                  | 36               | 51                  | 14                 |
| Northern Europe                 | 38               | 55                  | 7                  |
| Southern Europe                 | 50               | 48                  | 2                  |
| Western Europe                  | 55               | 45                  | 1                  |
| Latin America and the Caribbean | 61               | 26                  | 13                 |
| Caribbean                       | 67               | 23                  | 10                 |
| Central America                 | 64               | 26                  | 10                 |
| South America                   | 56               | 28                  | 16                 |
| Northern America                | 54               | 41                  | 5                  |
| Northern America                | 54               | 41                  | 5                  |
| Oceania                         | 49               | 31                  | 21                 |
| Australia/New Zealand           | 48               | 43                  | 8                  |
| Melanesia                       | 49               | 18                  | 33                 |
| Total World Average             | 55               | 32                  | 13                 |

Europe, produced capital has been the major contributor to overall growth in inclusive wealth. There are possible explanations for this: some of these countries had already achieved high levels of education across their populations, thus leaving them with less room to improve from already high levels of human capital. In some areas, high rates of migration likely resulted in low growth in human capital. In other areas, natural capital makes up a relatively small part of growth in total wealth. See Table I for further information. Patterns on the capital type contribution to inclusive wealth at an individual country level show that human capital is the major contributor to the growth rates in wealth for 101 out of 140 countries. In 27 of 140 countries produced capital is the greatest contributor to inclusive wealth. In only 12 of 140 countries did natural capital make up the largest contributor affecting the changes in wealth (negatively). See Figure 5 and Appendix 1 for additional details.

#### BOX 1 Changes in worldwide aggregate inclusive wealth

In order to see how the global economy is performing in terms of inter-temporal welfare, we carried out experimental wealth accounts worldwide. We did so by converting countries' wealth into international dollars (or Geary-Khamis dollars) using purchasing power parity (PPP) exchange rates<sup>9</sup>, and aggregating the wealth across all countries for the time period<sup>10</sup> of 1992 to 2010. Finally, we calculated changes of global inclusive wealth on the per capita basis. The results are illustrated in Figure 6.

Changes in global wealth per capita were largely stagnant throughout the 1990s, and increasingly positive changes in the period from 2000 to 2010. By 2010, the performance of the global economy from a wealth perspective had shown an increase of only 6 percent with respect to 1992. The major

9 We assume PPP to be a proxy variable for adjusting in the right direction the price differences of inter-temporal consumption of goods and services between countries.

10 Because data was unavailable for some countries in 1990 and 1991, we used 1992 as a baseline to maintain consistency.

positive changes can be observed for produced capital (56 percent growth from 1992 levels), followed by human capital (8 percent growth). Natural capital experienced a decline of about 30 percent from 1992 levels.

On average, human capital is the main source of wealth, making up 57 percent of total global inclusive wealth. Natural capital is responsible for 23 percent of total inclusive wealth, while produced capital makes up the remaining 20 percent.

Relatively low increases in human capital, combined with vast losses in natural capital, largely explain the anemic overall growth in inclusive wealth worldwide, despite enormous gains in produced capital.

From an economic production perspective, the outlook is much rosier: worldwide GDP per capita grew, albeit with some fluctuations around 2008, by over 50 percent in the past two decades – a trajectory similar to that of produced capital (as would be expected). These differences in progress as measured by W and GDP underscore the need for integrating sustainability into economic evaluation and policy planning.

Regional trends are shown in Appendix 2, using the same variables plotted in Figure 6.

#### FIGURE 6





#### 3.3 Wealth compositions

In this subsection we explore the sources of nations' wealth by looking at the composition of individual countries' productive bases valued at their shadow price. Composition is shown for our three capital asset stock groups at a country, sub-regional, regional, and global level. Country-level results are shown in Figure 7, which depicts the relative importance of each capital type in total wealth.

In addition to the overview of the countries' capital portfolio presented in Figure 7, we further explore how the overall capital portfolio is composed for an average nation on a global level. When analyzing the mean composition across the country sample for each capital type, the shares of the average country over the period of study clearly demonstrate the importance of human capital wealth, with a representation of 54 percent (see Figure 8-a).Developments over time show that, while for the average country the contribution of human capital and produced capital to the total wealth increased, the share of natural capital has declined (see Figure 8-b).

The dominance of human capital over the other two capital types tends to be a fair representation of the sample, as this trend holds true for 100 out of the 140 countries evaluated. What is more, in 89 of these 101 countries the percentage of human capital in total wealth is 50 percent or higher. For those countries in which human capital is not the most important relative source of wealth (40 of 140), natural capital is the most important asset category in all but one. The only country obtaining the majority of its wealth from produced capital (with 49 percent) is the Republic of Moldova.

Of those nations with a high proportion of natural capital in comparison to the other asset categories, 13 countries had 80 percent or more of their total wealth held in natural capital<sup>11</sup>. Interestingly, 11 of 13 generate wealth from renewable resources such as forest and agricultural land. Only Iraq and Kuwait had fossil fuelbased natural capital wealth. In the overall average for our complete country sample, 75 percent of a country's natural capital is attained from renewable resources, while 25 percent is based on non-renewable resources such as fossil fuels and minerals.

A more specific analysis is presented in Table 2, which includes a disaggregation of average capital type at the regional and sub-regional level. These breakdowns show that European countries in general, and Western European countries in particular, have a very low share of natural capital in relation to that of human capital, which tends to be high. In fact, the only regions in which natural capital is the most important source of wealth are: Middle Africa, South America, and Melanesia. The wealth portfolios of the sub-regions of Latin America and the Caribbean are also interesting since each of them reveals a different trend, contrary to common understanding of regional economic similarities (see Table 2).

An interesting observation can be seen in the correlations between the composition of human and natural capital (see Figure 9): countries with a high share of human capital are generally likely to have a low share of natural capital, and vice versa. This is particularly true in the case of high-income<sup>12</sup> countries, which have relatively high wealth held in human capital, and relatively low wealth in natural capital (see also Table 3). This is largely explained by the relative stability of produced capital's proportion in total wealth. Produced capital's share tends toward around 18 percent in most nations; for 70 percent of the countries

<sup>11</sup> The following countries are part of the group of 13 countries presenting a high level of natural capital in

total wealth: Bolivia, Central Africa, Congo, Democratic Republic of the Congo, Guyana, Iraq, Kuwait, Lao People's Democratic Republic, Liberia, Mongolia, Mozambique, Papua New Guinea, and Zambia.

<sup>12</sup> In total there are 42 high-income economies, 35 upper middle-income economies, 37 lower middle-income economies, and 26 low-income economies.

#### FIGURE 7

Percentage of human, produced, and natural capital in total wealth, annual average for 1990–2010.



FIGURE 7 a: Percentage of natural capital in total wealth

FIGURE 7 b: Percentage of produced capital in total wealth



FIGURE 7 c: Percentage of human capital in total wealth



#### FIGURE 8

Developments in the composition of wealth by capital form, 1992–2010





Average wealth composition across countries, mean 1992–2010



Key

Natural Capital

#### FIGURE 8 b

Developments in the country average wealth composition



#### **FIGURE 9**

Percentage shares of human capital and natural capital in total wealth, average 1990–2010

measured, produced capital made up between 12 and 28 percent of total wealth.

These shares should be interpreted with caution. They show the worth of one capital form in relation to the total wealth of a country, but they do not provide information about the absolute wealth of the different capital types. Norway, for example, has a high level of total wealth in natural capital; however, its share still only amounts to 12 percent due to Norway's high level of human capital wealth.

|                                 | Human<br>Capital | Produced<br>Capital | Natural<br>Capital |
|---------------------------------|------------------|---------------------|--------------------|
| Africa                          | 47               | 13                  | 40                 |
| Eastern Africa                  | 47               | 12                  | 41                 |
| Middle Africa                   | 16               | 7                   | 77                 |
| Northern Africa                 | 56               | 18                  | 26                 |
| Southern Africa                 | 59               | 18                  | 23                 |
| Western Africa                  | 52               | 12                  | 37                 |
| Asia                            | 51               | 18                  | 31                 |
| Eastern Asia                    | 47               | 22                  | 31                 |
| South-Central Asia              | 54               | 19                  | 26                 |
| South-Eastern Asia              | 46               | 16                  | 39                 |
| Western Asia                    | 53               | 16                  | 31                 |
| Europe                          | 66               | 26                  | 8                  |
| Eastern Europe                  | 57               | 28                  | 15                 |
| Northern Europe                 | 67               | 24                  | 9                  |
| Southern Europe                 | 70               | 24                  | 6                  |
| Western Europe                  | 70               | 28                  | 2                  |
| Latin America and the Caribbean | 52               | 17                  | 31                 |
| Caribbean                       | 68               | 22                  | 10                 |
| Central America                 | 58               | 18                  | 25                 |
| South America                   | 39               | 13                  | 48                 |
| Northern America                | 62               | 19                  | 19                 |
| Northern America                | 62               | 19                  | 19                 |
| Oceania                         | 45               | 17                  | 38                 |
| Australia/New Zealand           | 53               | 24                  | 23                 |
| Melanesia                       | 37               | 9                   | 53                 |
| Total World Average             | 54               | 18                  | 28                 |

# TABLE 2

Regional and subregional composition (in percentage) of wealth by capital form, average 1990–2010

|                     | Human<br>Capital | Produced<br>Capital | Natural<br>Capital |
|---------------------|------------------|---------------------|--------------------|
| High Income         | 64               | 24                  | 12                 |
| Low income          | 47               | 12                  | 41                 |
| Lower Middle Income | 46               | 15                  | 39                 |
| Upper Middle Income | 55               | 19                  | 25                 |
| Total World Average | 54               | 18                  | 28                 |

#### TABLE 3

Wealth composition (in percentage) based on income-based country classification, average 1990-2010

#### 3.4 Adjusted Inclusive Wealth Index

This subsection investigates how countries' performances based on W are impacted after taking into account the following three factors: 1) climate change, accounting particularly for those damages that nations have suffered as a result of increased carbon concentration in the atmosphere; 2) total factor productivity changes, capturing exogenous contributions of multiple missing factors to economic growth 3) oil capital gains, reflecting how changes in oil prices may increase or decrease the value of a country's productive base. Similarly to the IWR 2012, we refer to this adjusted figure as "Adjusted Inclusive Wealth Index" (Wadj) (see Annex 2 for further explanations).

A nation's wealth could in principle be either positively or negatively affected by each of the three adjustment factors. For instance, Nordhaus and Boyer (2000) predict that some countries will benefit from the changes in climate, while others will be negatively impacted by those perturbations. In the case of change in oil prices, countries that are net oil producers would observe positive changes in wealth, while countries that are dependent upon oil imports will be negatively impacted. Total factor productivity adjustments can also go either way, as less efficient use of capital results in reduced productivity from the previous year.

An overview of the estimates is illustrated in Figure 10, which shows how each of the three adjustments contributes to W. In addition, Figure 10 displays the final adjusted measure of wealth, Wadj.

Out of 55 countries for which we had reported negative growth in W per capita, 10 moved to positive growth rate after wealth adjustments. The remaining 45 countries still experienced negative growth in inclusive wealth after adjustments.

Remarkably, of the 85 countries that had shown positive W per capita in section 3.1, 37 moved into the negative bracket after adjustments. Considering this, the global outlook becomes far less optimistic. We have now identified 82 countries – or over half our sample – that are facing long-term issues in sustaining current consumption patterns. As the remaining 48 countries of the pre-adjusted positive bracket remained positive after adjustments, there are in total 58 countries exhibiting a positive trend.

When we drill down into specific adjustment impacts, TFP shows negative growth in 91 out of the 140 studied countries. The average growth rates in TFP range from  $\pm$  6 percent, and thus had a considerable impact on results in several countries, such as China, which moved into a negative position in Wadj due to negative changes in TFP.

In the case of climate change, results confirm that most of the countries (134 of 140) would experience negative economic consequences from this phenomenon. The six remaining countries would exhibit improvements in the productive base. Regarding the magnitude of the climate change damages in relation to the inclusive wealth of the countries, it is estimated that such effects, at less than 0.25 percent, are still relatively low.

Trends for oil capital gains illustrate that most of the nations (119 of 140) suffer from increases in the price of oil. The remaining 21 countries<sup>13</sup> show positive changes (essentially a positive economic re-evaluation of oil reserves). Three countries, for example, enjoy high oil capital gains at a rate of at least 3 percent: Kuwait (7.0 percent), Iraq (5.9 percent), and Saudi Arabia (3.5 percent). These magnitudes offset unadjusted negative trends shown in Wfor these nations, and in some cases even the negative impacts caused by climate change and TFP. A key reason why we prefer to keep oil capital gains separate from the initial W estimates is that the positive changes in wealth for

<sup>13</sup> This group is comprised by the following oilextractive economies: Kuwait, Iraq, Saudi Arabia, Iran, Kazakhstan, Venezuela, Canada, Norway, Nigeria, Russia, Algeria, Ecuador, Trinidad and Tobago, United Arab Emirates, Qatar, Yemen, Gabon, Cameroon, and Congo.

the above three countries were not the result of improved management of the countries' asset portfolios, but rather by the rises in oil prices that resulted primarily from changes in oil supply, demand, and geo-political trends.

3.5 Measuring economic performance: a comparison of inclusive wealth, GDP, and HDI

As we discussed at the beginning of the chapter, there exists a number of indicators to evaluate nations' economic performance and progress. Two of the most commonly used are gross domestic product (GDP) and the Human Development Index (HDI). GDP measures the market monetary value of all final goods and services produced in a given economy over a period of time (generally one calendar year). HDI measures a nation's performance as pertains to a selection of outcomes seen as critical to human well-being, such as life expectancy and educational attainment, in addition to income.

In this section, we provide an overview of country trends as measured by GDP and HDI alongside inclusive wealth. In doing so, we can identify how these measures of progress converge or diverge in the assessing of nations' performance. Figure 11 presents these measures in terms of the average percentage growth rates in W per capita, GDP per capita, and HDI over the period<sup>14</sup> of 1990 to 2010.

Figure 12 shows that W growth rates are in general more moderate than those of GDP and HDI. In most cases this can be explained by the additional factor included in the W, natural capital. As most countries experienced declines in natural capital, total W growth is decelerated as compared to GDP and HDI.

From this section, we identified a positive growth rate for 86 out of 140 countries<sup>15</sup> (61 percent); while in the case of the HDI 135 out of 140 countries (96 percent) show positive progress. The five countries with negative HDI growth are Lesotho, Swaziland, Tajikistan, United Arab Emirates, and Zimbabwe. In terms of GDP, 124 of 140 countries<sup>16</sup> (89 percent) show an overall increase in GDP growth over the past 21 years. Most of the countries with negative growth rates are located in Africa (9 countries) and Asia (3 countries). When looking at the three measures simultaneously for every country, we see consistent signs of progress in 80 of the 140 countries assessed (57 percent) across the three indicators. Signs of regress in all three indices are found only in Tajikistan, the United Arab Emirates, and Zimbabwe.

As the three measurements capture mostly different aspects of a system, the evaluation obtained about a country's performance is not always consistent in terms of progress or regress. Indeed, 42 out of 140 countries reveal, for example, a negative trend in average W growth rate, but positive growth rates in GDP and HDI.

When nations are grouped according to income level, it is conspicuous that for all three measurements high-income economies mostly show a positive development. There are, however, a few high-income economies that show a negative development from the *W* perspective: United Arab Emirates, Kuwait, Qatar, Saudi Arabia, and Trinidad and Tobago. These are all countries which the extraction of fossil fuels plays an important role, and the depletion in

<sup>14</sup> Due to missing data, Afghanistan, Cambodia, Czech Republic, and Estonia are compared for the period of 2000 to 2010, and Iraq, Nigeria, and United Arab Emirates for 2005 to 2010.

<sup>15</sup> For the whole period of study 55 countries with a negative W growth rate can be identified, but for this comparison the period for some countries had to be changed to the period of 2000 to 2010, thus the average growth rate for Afghanistan switched from negative to positive.

<sup>16</sup> Negative in HDI: Lesotho, Swaziland, Tajikistan, United Arab Emirates, Zimbabwe. Negative in GDP: Burundi, Central Africa, Côte d'Ivoire, Democratic Republic of Congo, Gabon, Haiti, Kyrgyzstan, Nigeria, Republic of Moldova, Serbia, Sierra Leone, Tajikistan, Togo, Ukraine, United Arab Emirates, Zimbabwe.





# FIGURE 11

Average annual growth rates of  $\,W$  per capita, GDP per capita, and HDI, period 1990–2010 (in percentage)



FIGURE 11 b: GDP per capita



FIGURE 11 c: HDI



#### FIGURE 12 a: HDI vs. W per capita



### FIGURE 12

Comparing annual avarage growth in W per capita, GDP per capita, and HDI, period 1990–2010

Note: as previously mentioned, due to missing data, Afghanistan, Cambodia, Czech Republic, and Estonia are compared for the period of 2000 to 2010, and Iraq, Nigeria, and United Arab Emirates for 2005 to 2010.

| ₩per | capita |
|------|--------|
|------|--------|



# FIGURE 12 b: GDP per capita vs. W per capita

#### FIGURE 12 c: GDP per capita vs. HDI



non-renewable resources explains to a large extent the negative growth rates in W. The increases in human and produced capital during the period of analysis were not high enough to compensate for the loss in natural capital. From this information, it is recommendable that these countries invest in other forms of capital development in order to compensate for non-renewable natural capital depletion and return to a sustainable trajectory. For low-income economies, we see that 18 of 26 countries experience negative W growth, even though most exhibit high growth rates in GDP and HDI (Figure 12 a-b).

In four countries – Haiti, Kyrgyzstan, Serbia, and Ukraine – we see positive development in terms of W and HDI but declines in GDP. All these countries did well in increasing their human capital, but, as suggested also by the negative GDP growth rate, they should increase their productive base, particularly their produced capital. When comparing HDI to W and GDP, we see two countries – Swaziland and Lesotho – with declining HDI growth rates but positive in the other indices. This indicates that while both countries have increased human capital, there are still conditions that need further improvements to assure peoples' well-being, for instance in life expectancy.

Most importantly, the comparison demonstrates that excluding environment in GDP and HDI provides usually a more optimistic impression of nations' development, one based in short-term performance rather than long-term potential. The evidence supports the contention that a broader set of indicators is needed to adequately measure nations' economic performance and progress in order to accommodate sustainability objectives, as well as improve comprehensive well-being assessments.

# 3.6 Understanding changes in human capital accounts

Human capital indicators, such as average years of total educational attainment (hereafter education) show that almost all countries in the world achieved growth in this metric<sup>17</sup> over the past two decades. This positive trend is in line with the IWR human capital accounts shown in previous sections that reflect the educational component of human capital (but exclude health). Findings not only reveal an increasing positive trend in this measure, but also show that human capital is the most important, and largest, source of inclusive wealth as well for the majority of countries.

Indeed, such evolution in both metrics should not come as a surprise as education is one of the inputs in the IWR human capital framework. However, along with education the method considers other variables to arrive at the final human capital outcome of a country, including number of adults<sup>18</sup> that achieved the average education level, and expected lifetime working period for the average person. Because there are several inputs interacting in the methodological framework, our interest in this section is to explore the following questions: Which inputs are driving the changes in the IWR human capital accounts? What is the relative importance of these human capital driving forces in different countries<sup>19</sup> and regions for the growth in human capital? To answer these questions we carried out a decomposition analysis of those interacting factors that drive the changes in this category of the accounts.

As presented in Chapter 4 and Annex 2, the method used for calculating the wealth of human capital consists in general of three multiplied components, or "terms"<sup>20</sup>. "Term I"

17 Barro and Lee (2010) database reports that only 2 out of the 140 countries here studied present a regress in the average year of total education. These countries are Namibia and Tajikistan.

18 We define adults as those individuals who reach the age of five plus the average years of total educational attainment.

19 The study of national polices boosting these results is out of the scope of analysis.

is a function of education; "Term II" provides information on the population of a country that has reached the average education level; and "Term III" is the present value of the average labor compensation per unit of human capital received by workers over an entire life's working period.

Table 4 illustrates the results as percentage changes in human capital and the breakdown for the three drivers under analysis (Terms I, II, and III), as well as showing the relative importance among the drivers of total change in human capital. For simplicity, results are additionally presented as the change between the years 1990 and 2010 and the average among countries comprising each region and subregion. Results at country level can be additionally found in Appendix 3.

Table 4 shows that Term II, representing the number of people in a country who achieved the average education level, is the main driving force of the changes in human capital on a global and regional scale, with Europe as the only exception. Similar results also apply at country level, with 103 of 140 countries exhibiting this pattern (see Appendix 3). On average globally (the average of the 140 countries in our sample), Term II is responsible for 48 of the 65 percent change in human capital from 1990 to 2010, demonstrating that Term II is the most significant of the three terms in driving change. For this driver, Africa (61 percent) and Asia (78 percent) exhibit the strongest development in relation to the changes in other regions. In particular Western Asia shows a remarkable

and Klenow and Rodríguez–Clare (1997) where the Wealth of Human Capital =  $\underbrace{e^{(\rho \cdot Edu)}}_{Term II} \cdot \underbrace{P_{5+Edu}}_{Term II} \cdot \underbrace{\int_{t=0}^{T} \bar{r} \cdot e^{-\delta \cdot t} dt}_{Term III}$ 

being:  $\rho$ = return on human capital; Edu= education; P<sub>5+edu</sub>= adult population; *T*= life's working period of the average person;  $\vec{r}$  = compensation per unit of human capital that is kept constant at the average value of the period 1990–2010; and  $\delta$ = discount rate. For further details on this formulation, refer to Chapter 4 and Annex 2. Note that the only variable changing in Term III is *T*.

<sup>20</sup> The formulation used for estimating human capital follows the method described in Arrow et al. (2012)

improvement, with an average increase of over 132 percent; this enormous growth is largely due to vast changes in the United Arab Emirates (455 percent), Qatar (367 percent), and Bahrain (219 percent).

Term I was identified as the second largest driver in our human capital accounts in the global average, representing 19 percent out of the 65 percent total change in human capital between 1990 and 2010. Such change represented about 32 percent of the total increase in human capital for the average country. In the case of Europe. Term I was the most significant source of growth, representing 15 percent of the 21 percent total increase in human capital; for this region, this source leads almost half (48 percent) of the total changes in relation to the other two terms (see Table 4). Overall, 100 of 140 countries count Term I as the second largest driving force of human capital, for 35 countries it is the most important driver of change.

In contrast to the trajectories of Terms I and II, Term III actually declined in the average country by two percent from the base year of 1990. In other words, despite increases in the embodied human capital per person and the total number of educated people, the time period that the average person stays in the labor market (Term III) decreased in the majority of countries (86 of 140). Term III's contribution to total human capital did not have a significant impact on total human capital growth in relation to Term I and II, representing only around 9 percent for the average country (see Table 4).

Picturing the overall development of human capital through the prism of the different drivers of human capital has shown that, in most of the cases, the number of people who reach the average education level is the primary source of growth in total human capital, with Europe as the only exception due to relatively low changes in population. It should be noted, however, that a per capita adjustment would diminish the importance of this driver (Term II).

Developing countries saw higher gains in terms of the embodied physical human capital as represented by Term I compared to those of Europe, Northern America, and Oceania. This is partly because most of the countries in those regions had already achieved high educational standards in the past, and marginal changes diminished over time.

# 4. Practical considerations and study limitations

While the accounting and management of capital assets are important to understanding inter-temporal benefit flows, it is challenging in several cases to obtain data on the complete set of assets available in an economy, as well as the measurement of the full contribution of a particular asset to human well-being, due to the variety of benefits that may result from a specific capital stock. In this section we elaborate on specific challenges in the context of the IWR 2014 wealth accounts, as well as other practical considerations and limitations that should be taken into account together with the findings presented in this chapter.

In compiling wealth accounts for the IWR 2014, we substantially extended the country coverage from the IWR 2012 sample (from 20 to 140 countries). In actuality, we assessed more than 140 countries in terms of human, produced, and natural capital; however, while over 180 countries had available data for produced capital and at least one type of natural capital – a requirement for our assessment – data on human capital was available for only 144. Intersecting these requirements led to a reduction to the final sample size of 140. Thus, existing country data on required input variables was one of the restrictions encountered in our effort to further extend **W** accounting.

Despite the efforts of trying to better understand nations' resource base, there are still several areas for which assets are incompletely described. For instance, explicit estimates on the quantification of social capital are still missing from the IWR 2014, despite our efforts to extend our capital asset groups to include this category. For natural capital, we are still missing fisheries,
#### TABLE 4

Decomposition analysis of the three terms accounted in for human capital and their contribution to the changes in human capital

|                                 | perce  | ntage chai<br>to 1 | nge with res<br>990 | pect  | percentage contribution<br>to human capital growth of each term |         |          |       |
|---------------------------------|--------|--------------------|---------------------|-------|---|---------|----------|-------|
|                                 | Term I | Term II            | Term III            | total | Term I  | Term II | Term III | total |
| Africa                          | 19     | 61                 | -1                  | 79    | 24  | 72      | 4        | 100   |
| Eastern Africa                  | 18     | 56                 | -2                  | 73    | 25  | 71      | 3        | 100   |
| Middle Africa                   | 16     | 68                 | 0                   | 84    | 18  | 80      | 2        | 100   |
| Northern Africa                 | 27     | 49                 | -6                  | 69    | 33  | 60      | 8        | 100   |
| Southern Africa                 | 18     | 52                 | -6                  | 64    | 24  | 67      | 9        | 100   |
| Western Africa                  | 20     | 72                 | 2                   | 94    | 21  | 76      | 3        | 100   |
| Asia                            | 22     | 78                 | -3                  | 98    | 26  | 65      | 8        | 100   |
| Eastern Asia                    | 18     | 25                 | -7                  | 36    | 37  | 46      | 16       | 100   |
| South-Central Asia              | 24     | 54                 | 0                   | 78    | 33  | 62      | 5        | 100   |
| South-Eastern Asia              | 22     | 54                 | -3                  | 73    | 29  | 66      | 5        | 100   |
| Western Asia                    | 22     | 132                | -3                  | 151   | 15  | 74      | 11       | 100   |
| Europe                          | 15     | 8                  | -3                  | 21    | 48  | 36      | 16       | 100   |
| Eastern Europe                  | 13     | 1                  | -6                  | 8     | 49  | 29      | 21       | 100   |
| Northern Europe                 | 16     | 8                  | -4                  | 20    | 48  | 37      | 15       | 100   |
| Southern Europe                 | 16     | 12                 | -3                  | 26    | 50  | 37      | 14       | 100   |
| Western Europe                  | 16     | 13                 | 2                   | 31    | 47  | 41      | 12       | 100   |
| Latin America and the Caribbean | 21     | 41                 | 0                   | 62    | 35  | 59      | 6        | 100   |
| Caribbean                       | 17     | 27                 | -3                  | 41    | 35  | 52      | 13       | 100   |
| Central America                 | 24     | 56                 | 1                   | 80    | 31  | 67      | 2        | 100   |
| South America                   | 21     | 38                 | 2                   | 61    | 38  | 56      | 6        | 100   |
| Northern America                | 12     | 25                 | -4                  | 34    | 29  | 62      | 9        | 100   |
| Northern America                | 12     | 25                 | -4                  | 34    | 29  | 62      | 9        | 100   |
| Oceania                         | 9      | 42                 | 1                   | 52    | 17  | 80      | 3        | 100   |
| Australia/New Zealand           | 8      | 35                 | 3                   | 45    | 17  | 77      | 6        | 100   |
| Melanesia                       | 10     | 49                 | 0                   | 59    | 18  | 82      | 0        | 100   |
| Total World Average             | 19     | 48                 | -2                  | 65    | 32  | 59      | 9        | 100   |

Note: Term I = human capital embodied per person; Term II = adults who reached the average education level; and Term III = capitalized labor compensation per unit of human capital. Negative changes in Term III are considered in absolute numbers to estimate the relative contribution among Term I, II, and III.

water accounts, and several ecosystem services (e.g., estuarine and coastal ecological systems<sup>21</sup>), due to a lack of data interpreting the dynamic of these important services for human well-being. Moreover, in order to include a country in our sample we required data for at least one of the natural capital assets, along with produced and human capital. While data availability in renewable resources tends to be relatively good across nations, estimates on mineral reserves tend to be incomplete, despite a prevalence of mineral extraction (or production) flow reports.

A key challenge for this report was in capturing the true contribution of particular assets to human well-being. To truly understand this, we must understand the full set of benefits resulting from a given asset's diverse roles. For example, human capital has been defined as "the knowledge, skills, competencies and attributes embodied in individuals that facilitate the creation of personal, social and economic well-being" (OECD 2001A, P.18). How, though, does one measure such stocks in a population? One compelling solution is to look at the returns of such human capital in future income streams, also referred to as the income-based approach (see Chapters 3, 4, and Annex 2). However, the inter-temporal sum of the marginal payments for such labor (wages) reflects only the market compensation for the human capital contribution to production. Yet we know there are other benefits for human wellbeing, from human capital investment including health, subjective well-being, informed citizens, and willingness to cooperate, among others (see Chapter 3).

In the case of produced capital, the System of National Accounts (SNA) captures investments in produced capital by the market value of the capital goods and services acquired in a time period, which is also what our accounts are reflecting. Nonetheless, capital goods recorded in the SNA that serve to satisfy infrastructural needs may bring about amenities to a population beyond its central functionalities. Natural capital benefits

resulting from changes in agricultural land are, for example, measured based on the contribution of this resource to the production system only. Any other role of agricultural land to human well-being has so far been excluded. For climate change damages, results should also be taken cautiously, since these estimates rely on a single study (NORDHAUS AND BOYER 2000), conducted more than a decade ago. Incorporating recent knowledge and additional study estimates would, given the considerable variability in the findings among the works in this area, help to improve accuracy on the level of damages resulting from (anthropogenic) changes in climate. Other aspects of note are the inherent uncertainties related to modeling any similar phenomenon, such as unknown functional relationships, errors in the prediction of future parameters in natural events and/or human behavior, and the lack of measurement of some important impacts<sup>22</sup> affecting particular countries (SCHNEIDER ET AL. 2002).

To compensate for such missing and incomplete data on country balance sheets and capital asset contributions to human well-being, we introduced total factor productivity (TFP) as a proxy variable for Adjusted Inclusive Wealth Index calculations. In our accounts, TFP represents the contribution to production of multiple implicit factors after produced, human, and natural capital items have been isolated. However, it is worth remarking this multifactor variable again accounts only for "market" production, and as such excludes our absent assets' informal and "direct" (e.g., several ecosystem services) contributions to human well-being.

Another important component of the index is the associated value attached to changes in physical stocks. Theoretical models refer to shadow (social) prices, which measure the marginal contribution of an asset to inter-temporal welfare. In several cases, however, empirical applications use

<sup>21</sup> See Chapter 7 in this report and Chapter 8 in the IWR 2012 for insights into this type of natural capital.

<sup>22</sup> In the work of Nordhaus and Boyer (2000) Carbon damages have been measured in the following areas: agriculture, sea-level rise, other vulnerable market sectors (energy systems), health, and non-market amenity impacts. For further details see Nordhaus and Boyer (2000).

market prices as a proxy for unavailable shadow prices. This practice may distort an asset's contribution to well-being in the case of market failures, such as environmental externalities.

The fact that a shadow price collects information of an asset's benefits for human well-being "beyond the market" brings with it some difficulties in comparing wealth (or per capita wealth) in two or more countries at a given point in time. There are several issues that may arise from this. For example, two individuals with identical physical human capital (i.e., skills, knowledge, or others) in different nations may still generate substantially different human capital wealth per person. One of the reasons for this could be that the citizen in the first country faces a labor market condition in which the resource in analysis is abundant, while the person in the second nation experiences the opposite situation - that is, a scarcity in the respective production factor. This would lead to divergent shadow prices for each individual's lifetime earnings - a key input for the lifetime income approach framework.

The previous issue would also be reflected on the consumption side, as an inter-temporal basket of consumed goods and services may have different prices in different countries. This should be corrected by using a sort of international dollars as adjusted by purchasing power parity, but across an inter-temporal framework. Other valuation issues might also arise when using a wealth measure at the level. For instance, using the marginal price at a specific period in time to value a complete forest stock ignores changes in willingness to pay over time in those cases when the resource is being depleted. These are some of the issues that could lead to misleading conclusions when comparing wealth (or per capita wealth) at the level.

Other caveats are inherent to the methodologies used to quantify the amount of a resource available in a country. For human capital, for example, the lifetime working period forecasts are based on contemporaneous indicators of mortality and labor force across age cohorts, and not in temporal projections (for details on this see Chapter 4). In the case of natural capital, one should be particularly cautious with the interpretation of values used in non-timber forest accounts, since the estimated benefits per unit of forest are based on global average. This may in some cases not be representative for all countries (see Annex 2 and Chapter 7 for further details).

Additional limitations refer to linear interpolations carried out to obtain information on the annual basis. For instance, the indicator on the average years of total schooling is reported every five years in Barro and Lee (2013), meaning that we were forced to perform estimates of the indicator for intermediate years. Similar interpolations were carried out to estimate annual survival rates in the human capital accounts, as well as annual growth of forest stock in our natural capital accounts.

Despite these limitations, we believe that the trends we have observed in a wide range of capital assets, and their analysis, provide important insight and knowledge into discussions of sustainable economic performance.

#### 5. Final remarks

In this chapter, we assessed nations' performance in the light of changes in inclusive wealth – that is, the changes in the aggregated value of capital assets contributing to human well-being. Evaluating the changes in wealth of a particular country provides us with insights into its economy's ability to sustain present levels of welfare over the long term. Changes in wealth were carried out from an inclusive perspective by factoring in not only produced capital, but also human and natural capital. For the IWR 2014, we expanded our country sample, now covering 140 countries over a twenty one-year period (1990 to 2010).

Empirical evidence shows that nations' growth in inclusive wealth, while generally positive, has been considerably less rapid than growth in GDP, or even HDI. While 124 countries demonstrated growth in GDP and 135 in HDI (out of the 140 countries we looked at), only 85 countries showed progress in inclusive wealth. This is primarily because of a factor missing from both GDP and HDI: the environment. Depleting forests or extracting fossil fuels and mineral resources will increase GDP in the short run, but will endanger consumption potentials of a nation in the future if resources are not appropriately devoted to building the resource base, including not only human or produced capital, but also other forms of natural capital. Indeed, many countries that improved GDP might merely be transforming natural wealth into present consumption.

The depreciation of natural capital, together with population growth, constitutes the main driving force behind declining per capita wealth in the majority of countries. Population increased in 127 of 140 countries, while natural capital declined in 116 out of 140 countries. While population was a greater factor, both negatively impact growth in inclusive wealth per person. The study also reveals that human capital was the major contributor to growth rates in wealth for 101 of 140 countries. Even more, human capital is the single largest source of wealth in most countries. Produced capital came in second, representing the greatest contribution of wealth per person in 27 out of 140 countries. It is worth noting that produced capital, the capital type for which the vast majority of data and corresponding analysis exists, represents on average only about 18 percent of total wealth; while the rest of the capital types, which constitute the remaining 82 percent (54 percent human capital and 28 percent natural capital), are at best treated as satellite accounts in the System of National Accounts.

When the Inclusive Wealth Index is adjusted by carbon damages, oil capital gains, and total factor productivity, results show that all three factors negatively affected inclusive wealth growth in most countries.

While the estimates in this report are often incomplete – due both to missing data, as well as to difficulties in capturing the complete set of benefits to human well-being from particular assets – we consider that the trends we have uncovered provide important insight into understanding the development pathways of nations over the past two decades.

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## Appendices

## Appendix 1: Key statistics

Data as averages containing measurements from 1990–2010\*

|                                  |      |                 | ₩ per<br>Capita | Popu-<br>lation | XV             |     |              |                         |
|----------------------------------|------|-----------------|-----------------|-----------------|----------------|-----|--------------|-------------------------|
| Country                          | Code | Income<br>level | Growth<br>rate  | Growth<br>rate  | Growth<br>rate |     | Contri<br>ca | bution by<br>pital form |
|                                  |      |                 |                 |                 |                | нс  | PC           | NC                      |
| Africa                           |      |                 | -0.8            | 2.3             | 1.5            | 1.3 | 0.4          | -0.3                    |
| Eastern Africa                   |      |                 | -0.8            | 2.3             | 1.5            | 1.3 | 0.5          | -0.3                    |
| Burundi                          | BDI  | LI              | -0.1            | 2.0             | 1.9            | 1.9 | 0.2          | -0.2                    |
| Kenya                            | KEN  | LI              | 0.2             | 2.8             | 3.0            | 2.5 | 0.5          | 0.0                     |
| Mozambique                       | MOZ  | LI              | -2.4            | 2.8             | 0.3            | 0.4 | 0.2          | -0.3                    |
| Mauritius                        | MUS  | UMI             | 1.3             | 1.0             | 2.3            | 1.2 | 1.2          | 0.0                     |
| Malawi                           | MWI  | LI              | -1.3            | 2.3             | 1.0            | 1.0 | 0.3          | -0.2                    |
| Rwanda                           | RWA  | LI              | 0.7             | 2.1             | 2.8            | 2.3 | 0.5          | 0.0                     |
| United Republic of Tanzania      | TZA  | LI              | -2.3            | 2.9             | 0.5            | 0.5 | 0.8          | -0.8                    |
| Uganda                           | UGA  | LI              | -0.3            | 3.2             | 2.9            | 2.2 | 1.1          | -0.5                    |
| Zambia                           | ZMB  | LMI             | -2.3            | 2.6             | 0.2            | 0.3 | 0.2          | -0.2                    |
| Zimbabwe                         | ZWE  | LI              | -1.1            | 0.9             | -0.2           | 0.7 | 0.0          | -0.9                    |
| Middle Africa                    |      |                 | -2.2            | 2.5             | 0.3            | 0.5 | 0.1          | -0.3                    |
| Central African Republic         | CAF  | LI              | -2.1            | 2.0             | 0.0            | 0.1 | 0.0          | -0.1                    |
| Cameroon                         | CMR  | LMI             | -1.7            | 2.4             | 0.6            | 1.0 | 0.2          | -0.6                    |
| Democratic Republic of the Congo | COD  | LI              | -2.7            | 3.0             | 0.3            | 0.5 | 0.0          | -0.2                    |
| Congo                            | COG  | LMI             | -2.6            | 2.7             | 0.0            | 0.0 | 0.3          | -0.4                    |
| Gabon                            | GAB  | UMI             | -1.7            | 2.4             | 0.7            | 1.0 | 0.1          | -0.4                    |
| Northern Africa                  |      |                 | 0.1             | 1.7             | 1.8            | 1.4 | 0.7          | -0.3                    |
| Morocco                          | MAR  | LMI             | 1.1             | 1.3             | 2.4            | 1.5 | 0.9          | 0.0                     |
| Algeria                          | DZA  | UMI             | -0.7            | 1.7             | 0.9            | 1.1 | 0.5          | -0.7                    |
| Egypt                            | EGY  | LMI             | 0.6             | 1.8             | 2.5            | 1.9 | 0.9          | -0.4                    |
| Sudan (former)                   | SDN  | LMI             | -1.6            | 2.5             | 0.9            | 0.8 | 0.5          | -0.4                    |
| Tunisia                          | TUN  | UMI             | 1.1             | 1.2             | 2.4            | 1.7 | 0.8          | -0.1                    |
| Southern Africa                  |      |                 | 0.1             | 1.8             | 1.8            | 1.4 | 0.6          | -0.1                    |
| Botswana                         | BWA  | UMI             | -0.2            | 1.9             | 1.7            | 1.0 | 1.1          | -0.4                    |
| Lesotho                          | LSO  | LMI             | 0.9             | 1.4             | 2.3            | 1.4 | 0.9          | 0.0                     |
| Namibia                          | NAM  | UMI             | -0.8            | 2.4             | 1.6            | 1.5 | 0.3          | -0.2                    |
| Swaziland                        | SWZ  | LMI             | 0.3             | 1.6             | 1.9            | 1.6 | 0.3          | 0.0                     |
| South Africa                     | ZAF  | UMI             | 0.1             | 1.6             | 1.7            | 1.4 | 0.4          | -0.1                    |
| Western Africa                   |      |                 | -1.0            | 2.8             | 1.7            | 1.7 | 0.3          | -0.3                    |
| Benin                            | BEN  | LI              | -1.2            | 3.1             | 1.8            | 1.9 | 0.4          | -0.5                    |
| Côte d'Ivoire                    | CIV  | LMI             | -0.5            | 2.3             | 1.8            | 1.7 | 0.1          | 0.0                     |

\* For Croatia, Kazakhstan, Kyrgyzstan, Lithuania, Russian Federation, Slovenia, Tajikistan, and Ukraine the wealth account starts from 1991; for Czech Republic and Slovakia from 1992. \*\* GDP and HDI for Afghanistan, Cambodia, Czech Republic, and Estonia are averaged for the period 2000–2010 and Iraq, Nigeria, and United Arab Emirates for 2005–2010.

|                | ₩ adjusted       | per capita           |         | Percentage of inclus |    | iclusive we | ealth by capit         | al form                        | m Other indicators** |                |  |
|----------------|------------------|----------------------|---------|----------------------|----|-------------|------------------------|--------------------------------|----------------------|----------------|--|
| Growth<br>rate | Contri           | bution by adj        | ustment | нс                   | PC | NC          | Brea                   | akdown NC                      | GDP per<br>capita    | HDI            |  |
|                | Carbon<br>Damage | Oil Capital<br>Gains | TFP     |                      |    |             | Renewable<br>resources | Non-<br>renewable<br>resources | Growth<br>rate       | Growth<br>rate |  |
| -1.4           | -0.1             | 0.0                  | -0.5    | 47                   | 13 | 40          | 36                     | 4                              | 1.2                  | 1.1            |  |
| -1.6           | -0.1             | -0.2                 | -0.5    | 47                   | 12 | 41          | 40                     | 1                              | 1.7                  | 1.3            |  |
| 0.4            | -0.1             | -0.1                 | 0.7     | 66                   | 14 | 20          | 20                     | 0                              | -1.1                 | 1.3            |  |
| 0.1            | -0.2             | -0.3                 | 0.3     | 72                   | 14 | 14          | 14                     | 0                              | 0.2                  | 0.5            |  |
| -5.0           | 0.0              | -0.1                 | -2.4    | 14                   | 4  | 82          | 81                     | 1                              | 4.4                  | 2.3            |  |
| 0.8            | -0.1             | -0.4                 | 0.0     | 76                   | 22 | 2           | 2                      | 0                              | 3.5                  | 0.8            |  |
| -3.0           | -0.1             | -0.2                 | -1.3    | 36                   | 18 | 46          | 46                     | 0                              | 1.5                  | 1.7            |  |
| 1.3            | -0.2             | -0.2                 | 1.1     | 79                   | 13 | 7           | 7                      | 0                              | 3.2                  | 3.1            |  |
| -5.8           | -0.1             | -0.1                 | -3.2    | 17                   | 11 | 72          | 71                     | 1                              | 2.6                  | 1.4            |  |
| -2.7           | -0.2             | -0.2                 | -2.0    | 61                   | 18 | 20          | 20                     | 0                              | 3.4                  | 2.0            |  |
| -2.0           | 0.0              | 0.0                  | 0.5     | 9                    | 4  | 86          | 85                     | 1                              | 0.6                  | 0.5            |  |
| 0.2            | -0.1             | -0.2                 | 1.7     | 42                   | 3  | 55          | 50                     | 5                              | -1.4                 | -0.7           |  |
| -1.8           | 0.0              | 0.3                  | 0.2     | 16                   | 7  | 77          | 73                     | 4                              | -0.8                 | 0.4            |  |
| -2.1           | 0.0              | 0.0                  | 0.0     | 3                    | 2  | 95          | 95                     | 0                              | -0.7                 | 0.5            |  |
| -2.6           | -0.1             | 0.0                  | -0.7    | 34                   | 10 | 56          | 55                     | 1                              | 0.1                  | 0.6            |  |
| -0.6           | 0.0              | 0.0                  | 2.2     | 14                   | 2  | 85          | 84                     | 0                              | -3.1                 | 0.0            |  |
| -2.9           | -0.1             | 0.7                  | -0.8    | 1                    | 8  | 91          | 82                     | 9                              | 0.4                  | 0.2            |  |
| -0.8           | -0.1             | 0.8                  | 0.2     | 28                   | 14 | 58          | 47                     | 11                             | -0.7                 | 0.5            |  |
| -0.4           | -0.1             | 0.1                  | -0.6    | 56                   | 18 | 26          | 15                     | 11                             | 2.8                  | 1.4            |  |
| 0.7            | -0.1             | -0.3                 | 0.0     | 73                   | 20 | 7           | 7                      | 0                              | 2.4                  | 1.4            |  |
| -0.4           | -0.1             | 0.9                  | -0.5    | 37                   | 27 | 37          | 5                      | 31                             | 1.0                  | 1.2            |  |
| -0.4           | -0.2             | -0.3                 | -0.6    | 72                   | 16 | 13          | 1                      | 12                             | 3.1                  | 1.4            |  |
| -1.1           | -0.1             | 0.5                  | 0.1     | 27                   | 4  | 68          | 60                     | 9                              | 4.1                  | 1.6            |  |
| -0.9           | -0.1             | -0.2                 | -1.7    | 71                   | 24 | 5           | 2                      | 3                              | 3.4                  | 1.3            |  |
| -1.3           | -0.1             | -0.1                 | -1.1    | 59                   | 18 | 23          | 19                     | 4                              | 1.9                  | 0.1            |  |
| -2.7           | -0.2             | -0.1                 | -2.2    | 32                   | 17 | 51          | 50                     | 1                              | 3.3                  | 0.4            |  |
| 0.4            | -0.1             | -0.1                 | -0.3    | 76                   | 24 | 1           | 1                      | 0                              | 2.2                  | -0.2           |  |
| -2.5           | 0.0              | -0.1                 | -1.6    | 53                   | 11 | 35          | 35                     | 0                              | 1.7                  | 0.3            |  |
| -1.5           | -0.1             | -0.1                 | -1.6    | 71                   | 24 | 5           | 3                      | 2                              | 1.0                  | 0.0            |  |
| -0.3           | -0.1             | -0.2                 | -0.1    | 62                   | 17 | 21          | 5                      | 16                             | 1.1                  | 0.0            |  |
| -1.7           | -0.1             | -0.1                 | -0.5    | 52                   | 12 | 37          | 33                     | 3                              | 0.5                  | 1.4            |  |
| -2.9           | -0.1             | -0.2                 | -1.3    | 52                   | 10 | 38          | 38                     | 0                              | 1.1                  | 1.6            |  |
| -0.9           | -0.2             | -0.1                 | -0.1    | 62                   | 11 | 27          | 26                     | 1                              | -0.5                 | 0.9            |  |

 HI
 High income
 UMI
 Upper middle income

 LMI
 Lower middle income
 LI
 Low income

## Appendix 1: Key statistics

Data as averages containing measurements from 1990–2010\*

|                                  |      |                 | ₩ per<br>Capita | Popu-<br>lation |                | XA  | /           |                          |
|----------------------------------|------|-----------------|-----------------|-----------------|----------------|-----|-------------|--------------------------|
| Country                          | Code | Income<br>level | Growth<br>rate  | Growth<br>rate  | Growth<br>rate |     | Contr<br>ca | ibution by<br>pital form |
|                                  |      |                 |                 |                 |                | нс  | PC          | NC                       |
| Ghana                            | GHA  | LMI             | -0.7            | 2.5             | 1.8            | 1.4 | 0.7         | -0.3                     |
| Gambia                           | GMB  | LI              | 0.0             | 3.0             | 3.0            | 2.1 | 0.7         | 0.1                      |
| Liberia                          | LBR  | LI              | -3.2            | 3.2             | -0.1           | 0.5 | -0.1        | -0.5                     |
| Mali                             | MLI  | LI              | -1.6            | 2.9             | 1.3            | 1.2 | 0.3         | -0.2                     |
| Mauritania                       | MRT  | LI              | 0.3             | 2.8             | 3.1            | 2.8 | 0.5         | -0.2                     |
| Niger                            | NER  | LI              | -1.1            | 3.5             | 2.4            | 2.6 | 0.1         | -0.3                     |
| Nigeria                          | NGA  | LMI             | -1.8            | 2.5             | 0.6            | 1.2 | 0.0         | -0.6                     |
| Senegal                          | SEN  | LMI             | -1.0            | 2.7             | 1.7            | 1.3 | 0.7         | -0.3                     |
| Sierra Leone                     | SLE  | LI              | -0.9            | 2.0             | 1.1            | 1.4 | 0.1         | -0.3                     |
| Togo                             | TGO  | LI              | -0.6            | 2.5             | 1.9            | 2.3 | 0.0         | -0.4                     |
| Asia                             |      |                 | 0.0             | 2.1             | 2.1            | 1.6 | 0.8         | -0.2                     |
| Eastern Asia                     |      |                 | 1.0             | 0.7             | 1.7            | 0.6 | 1.2         | -0.2                     |
| China                            | CHN  | UMI             | 2.0             | 0.8             | 2.8            | 0.9 | 2.0         | -0.2                     |
| Japan                            | JPN  | HI              | 0.9             | 0.2             | 1.1            | 0.4 | 0.7         | 0.0                      |
| Republic of Korea                | KOR  | HI              | 2.4             | 0.6             | 2.9            | 1.0 | 1.9         | 0.0                      |
| Mongolia                         | MNG  | LMI             | -1.2            | 1.2             | -0.1           | 0.1 | 0.3         | -0.5                     |
| South-Central Asia               |      |                 | 0.3             | 1.7             | 2.0            | 1.6 | 0.6         | -0.1                     |
| Afghanistan                      | AFG  | LI              | -0.8            | 4.5             | 3.7            | 3.5 | 0.2         | 0.0                      |
| Bangladesh                       | BGD  | LI              | 1.4             | 1.7             | 3.2            | 2.2 | 1.0         | 0.0                      |
| India                            | IND  | LMI             | 0.7             | 1.7             | 2.5            | 1.5 | 1.1         | -0.1                     |
| Iran (Islamic Republic of)       | IRN  | UMI             | -0.7            | 1.5             | 0.8            | 0.9 | 0.3         | -0.4                     |
| Kazakhstan                       | KAZ  | UMI             | 0.3             | -0.2            | 0.2            | 0.3 | 0.2         | -0.3                     |
| Kyrgyzstan                       | KGZ  | LI              | 0.2             | 1.0             | 1.1            | 0.6 | 0.2         | 0.3                      |
| Sri Lanka                        | LKA  | LMI             | 1.2             | 0.9             | 2.1            | 1.5 | 0.7         | 0.0                      |
| Maldives                         | MDV  | UMI             | 3.2             | 1.8             | 5.1            | 3.0 | 2.1         | 0.0                      |
| Nepal                            | NPL  | LI              | -1.6            | 2.3             | 0.7            | 0.9 | 0.5         | -0.8                     |
| Pakistan                         | PAK  | LMI             | 0.6             | 2.2             | 2.9            | 2.4 | 0.6         | -0.1                     |
| Tajikistan                       | TJK  | LI              | -1.0            | 1.3             | 0.2            | 0.8 | -0.6        | 0.0                      |
| South-Eastern Asia               |      |                 | 0.2             | 1.7             | 1.9            | 1.3 | 1.0         | -0.3                     |
| Indonesia                        | IDN  | LMI             | 0.0             | 1.3             | 1.3            | 1.0 | 0.8         | -0.4                     |
| Cambodia                         | KHM  | LI              | -1.3            | 2.0             | 0.6            | 0.8 | 0.4         | -0.5                     |
| Lao People's Democratic Republic | LAO  | LMI             | -1.5            | 2.0             | 0.4            | 0.3 | 0.4         | -0.2                     |
| Myanmar                          | MMR  | LI              | -1.3            | 1.0             | -0.3           | 0.3 | 0.3         | -1.0                     |

\* For Croatia, Kazakhstan, Kyrgyzstan, Lithuania, Russian Federation, Slovenia, Tajikistan, and Ukraine the wealth account starts from 1991; for Czech Republic and Slovakia from 1992. \*\* GDP and HDI for Afghanistan, Cambodia, Czech Republic, and Estonia are averaged for the period 2000–2010 and Iraq, Nigeria, and United Arab Emirates for 2005–2010.

|                | ∭ adjusted       | djusted per capita Percentage of inc |         |    | ICLUSIVE WE | ealth by capit | al form             | Other indicators**             |                   |                |
|----------------|------------------|--------------------------------------|---------|----|-------------|----------------|---------------------|--------------------------------|-------------------|----------------|
| Growth<br>rate | Contri           | bution by adj                        | ustment | нс | PC          | NC             | Brea                | akdown NC                      | GDP per<br>capita | HDI            |
| -              | Carbon<br>Damage | Oil Capital<br>Gains                 | TFP     |    |             |                | Renewable resources | Non-<br>renewable<br>resources | Growth<br>rate    | Growth<br>rate |
| -3.4           | -0.2             | -0.3                                 | -2.2    | 45 | 23          | 32             | 30                  | 1                              | 2.4               | 1.2            |
| 1.6            | -0.2             | -0.4                                 | 2.1     | 61 | 8           | 31             | 31                  | 0                              | 0.7               | 1.5            |
| -0.7           | 0.0              | -0.1                                 | 2.8     | 11 | 7           | 81             | 81                  | 0                              | 1.3               | 0.9            |
| -3.9           | -0.1             | 0.0                                  | -2.1    | 36 | 7           | 57             | 57                  | 0                              | 1.9               | 2.7            |
| -1.1           | -0.1             | -0.6                                 | -0.7    | 78 | 12          | 11             | 3                   | 8                              | 1.3               | 1.3            |
| -2.8           | -0.1             | -0.1                                 | -1.5    | 66 | 18          | 16             | 15                  | 1                              | 0.1               | 2.1            |
| -2.8           | -0.1             | 2.1                                  | -2.9    | 44 | 12          | 44             | 16                  | 28                             | -0.6              | 1.3            |
| -1.8           | -0.1             | -0.3                                 | -0.3    | 40 | 15          | 46             | 45                  | 1                              | 0.8               | 1.2            |
| -0.1           | -0.1             | -0.3                                 | 1.2     | 55 | 3           | 41             | 41                  | 0                              | -1.2              | 1.7            |
| -1.4           | -0.1             | -0.3                                 | -0.4    | 70 | 15          | 15             | 14                  | 1                              | -0.8              | 0.9            |
| 0.2            | -0.1             | 0.5                                  | -0.5    | 51 | 18          | 31             | 15                  | 16                             | 3.2               | 1.1            |
| -0.9           | -0.1             | -0.2                                 | -1.7    | 47 | 22          | 31             | 24                  | 7                              | 4.3               | 1.0            |
| -3.7           | 0.0              | -0.2                                 | -5.5    | 49 | 19          | 32             | 21                  | 11                             | 9.6               | 1.7            |
| 0.3            | 0.0              | -0.1                                 | -0.4    | 64 | 35          | 1              | 1                   | 0                              | 0.8               | 0.4            |
| 1.1            | -0.2             | -0.4                                 | -0.7    | 67 | 28          | 5              | 5                   | 0                              | 4.8               | 1.0            |
| -1.4           | 0.0              | -0.1                                 | -0.1    | 7  | 6           | 87             | 70                  | 17                             | 2.1               | 0.8            |
| 1.5            | -0.1             | 0.0                                  | 0.4     | 54 | 19          | 26             | 11                  | 16                             | 3.3               | 1.3            |
| 0.5            | -0.1             | -0.1                                 | 1.5     | 67 | 25          | 8              | 5                   | 3                              | 10.0              | 4.5            |
| 1.1            | -0.1             | -0.2                                 | 0.0     | 78 | 16          | 6              | 6                   | 0                              | 3.5               | 1.7            |
| -1.1           | -0.2             | -0.2                                 | -1.5    | 65 | 14          | 21             | 9                   | 12                             | 4.8               | 1.5            |
| 0.0            | 0.0              | 2.4                                  | -1.7    | 26 | 12          | 63             | 3                   | 60                             | 2.7               | 1.6            |
| 8.9            | 0.0              | 1.2                                  | 0.5     | 24 | 19          | 57             | 1                   | 56                             | 3.3               | 0.6            |
| 3.4            | -0.1             | -0.5                                 | 2.2     | 37 | 24          | 38             | 17                  | 22                             | -0.2              | 0.0            |
| -0.4           | -0.1             | -0.3                                 | -1.3    | 80 | 15          | 6              | 6                   | 0                              | 4.3               | 0.8            |
| 4.9            | -0.1             | -0.6                                 | 2.4     | 77 | 23          | 0              | 0                   | 0                              | 4.8               | 0.7            |
| -3.5           | -0.1             | -0.1                                 | -1.7    | 31 | 10          | 58             | 58                  | 0                              | 2.2               | 1.5            |
| -0.8           | -0.1             | -0.3                                 | -1.0    | 69 | 17          | 14             | 7                   | 6                              | 2.0               | 1.5            |
| 3.9            | -0.1             | -1.0                                 | 5.0     | 43 | 38          | 20             | 5                   | 15                             | -1.0              | 0.0            |
| -1.4           | -0.1             | -0.3                                 | -1.2    | 46 | 16          | 39             | 35                  | 4                              | 4.7               | 1.4            |
| -1.5           | -0.1             | -0.2                                 | -1.2    | 47 | 13          | 40             | 32                  | 8                              | 3.5               | 1.3            |
| -4.5           | -0.1             | -0.2                                 | -2.8    | 23 | 6           | 71             | 71                  | 0                              | 6.7               | 1.8            |
| -0.5           | 0.0              | 0.0                                  | 1.1     | 10 | 3           | 87             | 87                  | 0                              | 5.1               | 1.7            |
| -2.4           | 0.0              | -0.1                                 | -0.9    | 18 | 2           | 79             | 74                  | 5                              | 8.5               | 2.4            |

 HI
 High income
 UMI
 Upper middle income

 LMI
 Lower middle income
 LI
 Low income

## Appendix 1: Key statistics

Data as averages containing measurements from 1990–2010\*

|                      |      |                 | ₩ per<br>Capita | Popu-<br>lation |                | XA   | /            |                         |
|----------------------|------|-----------------|-----------------|-----------------|----------------|------|--------------|-------------------------|
| Country              | Code | Income<br>level | Growth<br>rate  | Growth<br>rate  | Growth<br>rate |      | Contri<br>ca | bution by<br>pital form |
|                      |      |                 |                 |                 |                | нс   | PC           | NC                      |
| Malaysia             | MYS  | UMI             | 0.4             | 2.2             | 2.7            | 2.0  | 1.0          | -0.3                    |
| Philippines          | PHL  | LMI             | 0.5             | 2.1             | 2.6            | 1.7  | 0.8          | 0.0                     |
| Singapore            | SGP  | HI              | 1.9             | 2.6             | 4.6            | 2.7  | 1.9          | 0.0                     |
| Thailand             | THA  | UMI             | 1.2             | 1.0             | 2.2            | 0.9  | 1.5          | -0.1                    |
| Vietnam              | VNM  | LMI             | 1.9             | 1.4             | 3.3            | 1.7  | 1.5          | 0.1                     |
| Western Asia         |      |                 | -0.6            | 3.1             | 2.5            | 2.0  | 0.7          | -0.3                    |
| United Arab Emirates | ARE  | HI              | -3.0            | 7.4             | 4.2            | 4.0  | 0.6          | -0.4                    |
| Armenia              | ARM  | LMI             | 1.0             | -0.7            | 0.4            | -0.2 | 0.6          | -0.1                    |
| Bahrain              | BHR  | HI              | 0.8             | 4.9             | 5.7            | 4.5  | 1.4          | -0.2                    |
| Cyprus               | CYP  | HI              | 1.0             | 1.8             | 2.8            | 2.2  | 0.6          | 0.0                     |
| Iraq                 | IRQ  | LMI             | -2.9            | 3.0             | 0.1            | 0.4  | 0.0          | -0.4                    |
| Israel               | ISR  | HI              | 0.9             | 2.5             | 3.4            | 2.5  | 0.9          | 0.0                     |
| Jordan               | JOR  | UMI             | 0.7             | 3.0             | 3.7            | 2.9  | 0.8          | 0.0                     |
| Kuwait               | KWT  | HI              | -1.3            | 1.4             | 0.0            | 0.2  | 0.4          | -0.6                    |
| Qatar                | QAT  | HI              | -3.7            | 7.0             | 2.9            | 2.2  | 1.3          | -0.5                    |
| Saudi Arabia         | SAU  | HI              | -1.3            | 2.7             | 1.3            | 1.3  | 0.4          | -0.3                    |
| Syrian Arab Republic | SYR  | LMI             | -0.3            | 2.6             | 2.2            | 2.1  | 0.6          | -0.4                    |
| Turkey               | TUR  | UMI             | 0.9             | 1.5             | 2.4            | 1.5  | 1.0          | 0.0                     |
| Yemen                | YEM  | LMI             | -0.4            | 3.6             | 3.2            | 3.0  | 0.5          | -0.3                    |
| Europe               |      |                 | 1.0             | 0.2             | 1.2            | 0.6  | 0.6          | 0.0                     |
| Eastern Europe       |      |                 | 0.8             | -0.3            | 0.5            | 0.2  | 0.3          | -0.1                    |
| Bulgaria             | BGR  | UMI             | 1.0             | -0.8            | 0.1            | -0.2 | 0.3          | 0.0                     |
| Czech Republic       | CZE  | HI              | 1.3             | 0.1             | 1.4            | 0.6  | 0.8          | -0.1                    |
| Hungary              | HUN  | HI              | 1.3             | -0.2            | 1.1            | 0.7  | 0.5          | 0.0                     |
| Republic of Moldova  | MDA  | LMI             | -0.2            | -1.0            | -1.2           | -0.5 | -0.8         | 0.1                     |
| Poland               | POL  | HI              | 1.1             | 0.0             | 1.1            | 0.5  | 0.7          | -0.1                    |
| Romania              | ROU  | UMI             | 1.0             | -0.4            | 0.6            | 0.2  | 0.6          | -0.1                    |
| Russian Federation   | RUS  | UMI             | 0.1             | -0.2            | -0.1           | 0.2  | 0.0          | -0.3                    |
| Slovakia             | SVK  | н               | 1.1             | 0.2             | 1.3            | 0.6  | 0.7          | 0.0                     |
| Ukraine              | UKR  | LMI             | 0.4             | -0.7            | -0.3           | 0.0  | -0.2         | -0.1                    |
| Northern Europe      |      |                 | 0.9             | 0.2             | 1.2            | 0.6  | 0.7          | -0.1                    |
| Denmark              | DNK  | HI              | 0.5             | 0.4             | 0.9            | 0.2  | 0.7          | 0.0                     |
| Estonia              | EST  | н               | 1.7             | -0.8            | 0.9            | 0.1  | 0.8          | 0.0                     |

\* For Croatia, Kazakhstan, Kyrgyzstan, Lithuania, Russian Federation, Slovenia, Tajikistan, and Ukraine the wealth account starts from 1991; for Czech Republic and Slovakia from 1992. \*\* GDP and HDI for Afghanistan, Cambodia, Czech Republic, and Estonia are averaged for the period 2000–2010 and Iraq, Nigeria, and United Arab Emirates for 2005–2010.

|                | ₩ adjusted       | per capita           |         | Perce | entage of ir | nclusive we | ealth by capit         | al form                        | Other ind         | licators**     |
|----------------|------------------|----------------------|---------|-------|--------------|-------------|------------------------|--------------------------------|-------------------|----------------|
| Growth<br>rate | Contri           | bution by adj        | ustment | нс    | PC           | NC          | Brea                   | akdown NC                      | GDP per<br>capita | HDI            |
|                | Carbon<br>Damage | Oil Capital<br>Gains | TFP     |       |              |             | Renewable<br>resources | Non-<br>renewable<br>resources | Growth<br>rate    | Growth<br>rate |
| -0.9           | -0.1             | -0.1                 | -1.2    | 60    | 16           | 23          | 12                     | 12                             | 3.6               | 0.9            |
| -0.9           | -0.1             | -0.4                 | -0.8    | 65    | 24           | 11          | 10                     | 1                              | 1.7               | 0.6            |
| -2.1           | 0.0              | -1.1                 | -2.9    | 70    | 30           | 0           | 0                      | 0                              | 3.7               | 0.8            |
| -1.6           | -0.1             | -0.5                 | -2.2    | 53    | 32           | 14          | 11                     | 4                              | 3.5               | 0.9            |
| 1.9            | -0.1             | -0.2                 | 0.3     | 66    | 13           | 21          | 16                     | 5                              | 6.0               | 1.7            |
| 0.6            | -0.1             | 1.6                  | -0.3    | 53    | 16           | 31          | 3                      | 28                             | 1.6               | 0.7            |
| 0.3            | -0.1             | 3.3                  | 0.3     | 42    | 16           | 43          | 0                      | 43                             | -8.7              | -0.4           |
| 0.8            | -0.1             | -0.8                 | 0.6     | 71    | 25           | 4           | 4                      | 0                              | 3.6               | 0.7            |
| 0.3            | -0.2             | -0.2                 | -0.2    | 68    | 28           | 4           | 0                      | 4                              | 0.5               | 0.5            |
| -0.1           | -0.1             | -0.3                 | -0.7    | 80    | 19           | 1           | 1                      | 0                              | 1.8               | 0.4            |
| 8.3            | 0.0              | 5.9                  | 5.5     | 12    | 3            | 86          | 0                      | 85                             | 3.9               | 0.5            |
| 0.0            | 0.0              | -0.2                 | -0.7    | 78    | 22           | 0           | 0                      | 0                              | 1.9               | 0.5            |
| -1.9           | -0.1             | -0.9                 | -1.6    | 71    | 24           | 5           | 2                      | 3                              | 2.4               | 0.8            |
| 5.7            | -0.1             | 7.0                  | 0.2     | 9     | 6            | 85          | 0                      | 85                             | 4.5               | 0.5            |
| -3.4           | -0.1             | 4.1                  | -3.3    | 22    | 11           | 68          | 0                      | 68                             | 3.3               | 0.5            |
| 2.2            | -0.1             | 3.5                  | 0.1     | 33    | 8            | 59          | 13                     | 47                             | 0.3               | 0.9            |
| -2.8           | -0.1             | -0.3                 | -2.1    | 70    | 16           | 14          | 0                      | 13                             | 2.8               | 0.7            |
| 0.8            | -0.1             | -0.2                 | 0.2     | 73    | 20           | 8           | 6                      | 1                              | 2.4               | 1.2            |
| -2.1           | -0.1             | 0.4                  | -2.0    | 67    | 10           | 23          | 11                     | 12                             | 2.4               | 2.5            |
| 1.2            | -0.1             | -0.2                 | 0.4     | 66    | 26           | 8           | 5                      | 3                              | 1.8               | 0.5            |
| 1.8            | 0.0              | -0.2                 | 0.8     | 57    | 28           | 15          | 7                      | 9                              | 1.9               | 0.4            |
| -0.1           | 0.0              | -0.4                 | -0.6    | 55    | 26           | 19          | 6                      | 12                             | 2.1               | 0.5            |
| 2.6            | 0.0              | -0.2                 | 1.4     | 65    | 33           | 2           | 0                      | 2                              | 3.2               | 0.6            |
| 1.7            | 0.0              | -0.2                 | 0.6     | 70    | 26           | 4           | 2                      | 2                              | 1.5               | 0.8            |
| 2.2            | 0.0              | -0.4                 | 2.9     | 45    | 49           | 7           | 7                      | 0                              | -1.0              | 0.0            |
| 0.2            | 0.0              | -0.2                 | -0.7    | 78    | 17           | 5           | 1                      | 4                              | 3.9               | 0.2            |
| 0.9            | 0.0              | -0.2                 | 0.1     | 62    | 25           | 14          | 12                     | 2                              | 1.9               | 0.5            |
| 3.1            | 0.0              | 0.1                  | 0.9     | 31    | 17           | 52          | 27                     | 25                             | 1.7               | 0.4            |
| -0.1           | 0.0              | -0.2                 | -1.1    | 67    | 31           | 3           | 2                      | 1                              | 4.3               | 0.5            |
| 6.1            | 0.0              | -0.2                 | 3.7     | 38    | 28           | 34          | 2                      | 32                             | -0.5              | 0.1            |
| 0.8            | -0.1             | -0.1                 | 0.1     | 67    | 24           | 9           | 8                      | 1                              | 2.0               | 0.6            |
| 0.6            | -0.2             | 0.0                  | 0.3     | 73    | 25           | 1           | 0                      | 1                              | 1.2               | 0.5            |
| 2.1            | 0.0              | -0.2                 | 0.7     | 65    | 24           | 11          | 11                     | 0                              | 4.0               | 0.7            |

 HI
 High income
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 Upper middle income

 LMI
 Lower middle income
 LI
 Low income

## Appendix 1: Key statistics

Data as averages containing measurements from 1990–2010\*

|                                 |      |                 | ₩ per<br>Capita | Popu-<br>lation |                | XA  | /            |                         |
|---------------------------------|------|-----------------|-----------------|-----------------|----------------|-----|--------------|-------------------------|
| Country                         | Code | Income<br>level | Growth<br>rate  | Growth<br>rate  | Growth<br>rate |     | Contri<br>ca | bution by<br>pital form |
|                                 |      |                 |                 |                 |                | нс  | PC           | NC                      |
| Finland                         | FIN  | HI              | 0.7             | 0.4             | 1.1            | 0.6 | 0.5          | 0.0                     |
| United Kingdom                  | GBR  | HI              | 0.9             | 0.4             | 1.3            | 0.7 | 0.6          | -0.1                    |
| Ireland                         | IRL  | HI              | 1.5             | 1.2             | 2.7            | 1.5 | 1.2          | 0.0                     |
| Iceland                         | ISL  | HI              | 0.0             | 1.1             | 1.2            | 0.9 | 0.6          | -0.3                    |
| Lithuania                       | LTU  | UMI             | 1.3             | -0.6            | 0.7            | 0.2 | 0.5          | 0.0                     |
| Latvia                          | LVA  | UMI             | 1.7             | -0.8            | 0.9            | 0.1 | 0.6          | 0.1                     |
| Norway                          | NOR  | HI              | 0.3             | 0.7             | 1.0            | 0.8 | 0.6          | -0.4                    |
| Sweden                          | SWE  | HI              | 0.6             | 0.5             | 1.1            | 0.6 | 0.5          | 0.0                     |
| Southern Europe                 |      |                 | 1.1             | 0.3             | 1.4            | 0.8 | 0.7          | 0.0                     |
| Albania                         | ALB  | LMI             | 0.8             | -0.1            | 0.6            | 0.2 | 0.4          | 0.0                     |
| Spain                           | ESP  | HI              | 1.9             | 0.9             | 2.8            | 1.8 | 1.0          | 0.0                     |
| Greece                          | GRC  | HI              | 1.0             | 0.6             | 1.5            | 0.9 | 0.8          | -0.1                    |
| Croatia                         | HRV  | HI              | 0.9             | -0.2            | 0.8            | 0.2 | 0.5          | 0.0                     |
| Italy                           | ITA  | HI              | 0.8             | 0.3             | 1.1            | 0.5 | 0.7          | 0.0                     |
| Malta                           | MLT  | HI              | 1.6             | 0.6             | 2.3            | 1.4 | 0.9          | 0.0                     |
| Portugal                        | PRT  | HI              | 1.0             | 0.4             | 1.4            | 0.7 | 0.7          | 0.0                     |
| Serbia                          | SRB  | UMI             | 0.7             | 0.2             | 0.9            | 0.6 | 0.2          | 0.1                     |
| Slovenia                        | SVN  | HI              | 1.4             | 0.3             | 1.6            | 0.8 | 0.8          | 0.0                     |
| Western Europe                  |      |                 | 1.1             | 0.6             | 1.7            | 0.9 | 0.8          | 0.0                     |
| Austria                         | AUT  | HI              | 1.1             | 0.5             | 1.6            | 0.8 | 0.8          | 0.0                     |
| Belgium                         | BEL  | HI              | 1.0             | 0.4             | 1.4            | 0.7 | 0.7          | 0.0                     |
| Switzerland                     | CHE  | HI              | 0.4             | 0.7             | 1.1            | 0.6 | 0.6          | 0.0                     |
| Germany                         | DEU  | HI              | 1.5             | 0.2             | 1.7            | 1.2 | 0.5          | 0.0                     |
| France                          | FRA  | HI              | 1.1             | 0.5             | 1.6            | 1.0 | 0.6          | 0.0                     |
| Luxembourg                      | LUX  | HI              | 1.5             | 1.4             | 3.0            | 1.5 | 1.5          | 0.0                     |
| Netherlands                     | NLD  | HI              | 0.9             | 0.5             | 1.5            | 0.7 | 0.8          | 0.0                     |
| Latin America and the Caribbean |      |                 | 0.0             | 1.4             | 1.5            | 1.2 | 0.5          | -0.2                    |
| Caribbean                       |      |                 | 0.5             | 0.9             | 1.4            | 1.1 | 0.4          | -0.1                    |
| Barbados                        | BRB  | HI              | 0.6             | 0.3             | 0.9            | 0.6 | 0.3          | 0.0                     |
| Cuba                            | CUB  | UMI             | 0.1             | 0.3             | 0.4            | 0.4 | -0.1         | 0.1                     |
| Dominican Republic              | DOM  | UMI             | 1.0             | 1.6             | 2.6            | 1.8 | 0.9          | 0.0                     |
| Haiti                           | HTI  | LI              | 0.9             | 1.7             | 2.6            | 2.3 | 0.3          | 0.0                     |
| Jamaica                         | JAM  | UMI             | 0.7             | 0.7             | 1.4            | 0.9 | 0.6          | -0.1                    |

\* For Croatia, Kazakhstan, Kyrgyzstan, Lithuania, Russian Federation, Slovenia, Tajikistan, and Ukraine the wealth account starts from 1991; for Czech Republic and Slovakia from 1992. \*\* GDP and HDI for Afghanistan, Cambodia, Czech Republic, and Estonia are averaged for the period 2000–2010 and Iraq, Nigeria, and United Arab Emirates for 2005–2010.

| ₩ adjusted per capita |                  | Perce                | ntage of ir | al form | Other indicators** |     |                     |                                |                   |                |
|-----------------------|------------------|----------------------|-------------|---------|--------------------|-----|---------------------|--------------------------------|-------------------|----------------|
| Growth<br>rate        | Contri           | bution by adj        | justment    | нс      | PC                 | NC  | Brea                | akdown NC                      | GDP per<br>capita | HDI            |
|                       | Carbon<br>Damage | Oil Capital<br>Gains | TFP         |         |                    |     | Renewable resources | Non-<br>renewable<br>resources | Growth<br>rate    | Growth<br>rate |
| -0.1                  | -0.1             | -0.1                 | -0.5        | 65      | 27                 | 8   | 8                   | 0                              | 1.7               | 0.5            |
| 0.1                   | -0.2             | -0.1                 | -0.5        | 78      | 21                 | 1   | 0                   | 1                              | 1.8               | 0.5            |
| 0.5                   | -0.2             | -0.1                 | -0.7        | 69      | 29                 | 2   | 2                   | 0                              | 3.4               | 0.7            |
| -0.4                  | -0.1             | -0.1                 | -0.2        | 48      | 17                 | 35  | 35                  | 0                              | 1.3               | 0.5            |
| 2.2                   | 0.0              | -0.3                 | 1.3         | 68      | 27                 | 5   | 5                   | 0                              | 1.8               | 0.5            |
| 3.9                   | 0.0              | -0.3                 | 2.4         | 67      | 21                 | 11  | 11                  | 0                              | 1.5               | 0.7            |
| -0.5                  | -0.2             | 0.3                  | -1.0        | 61      | 27                 | 12  | 2                   | 9                              | 1.9               | 0.6            |
| 0.0                   | -0.2             | -0.1                 | -0.3        | 70      | 24                 | 6   | 5                   | 0                              | 1.6               | 0.5            |
| 1.4                   | -0.1             | -0.2                 | 0.6         | 70      | 24                 | 6   | 4                   | 2                              | 1.8               | 0.6            |
| -2.0                  | -0.1             | -0.1                 | -2.5        | 38      | 35                 | 27  | 13                  | 14                             | 4.1               | 0.6            |
| 2.5                   | -0.1             | -0.2                 | 0.9         | 71      | 27                 | 2   | 2                   | 0                              | 1.6               | 0.8            |
| -0.1                  | -0.2             | -0.3                 | -0.7        | 59      | 31                 | 10  | 7                   | 3                              | 1.7               | 0.6            |
| 2.7                   | 0.0              | -0.2                 | 2.0         | 80      | 17                 | 3   | 3                   | 0                              | 2.2               | 0.6            |
| 0.6                   | -0.2             | -0.1                 | 0.1         | 67      | 31                 | 2   | 2                   | 0                              | 0.6               | 0.7            |
| 1.8                   | -0.1             | -0.4                 | 0.7         | 80      | 20                 | 0   | 0                   | 0                              | 2.8               | 0.5            |
| 1.4                   | -0.1             | -0.2                 | 0.7         | 77      | 22                 | 1   | 1                   | 0                              | 1.5               | 0.7            |
| 4.2                   | 0.0              | -0.1                 | 3.5         | 83      | 12                 | 5   | 5                   | 0                              | -0.7              | 0.3            |
| 1.7                   | 0.0              | -0.2                 | 0.5         | 72      | 24                 | 5   | 4                   | 1                              | 2.7               | 0.3            |
| 0.8                   | -0.2             | -0.2                 | 0.1         | 70.1    | 28.1               | 1.7 | 0.9                 | 0.8                            | 1.5               | 0.5            |
| 0.4                   | -0.2             | -0.1                 | -0.4        | 68      | 30                 | 2   | 2                   | 0                              | 1.6               | 0.6            |
| 1.1                   | -0.2             | -0.2                 | 0.5         | 73      | 27                 | 0   | 0                   | 0                              | 1.4               | 0.5            |
| 0.5                   | -0.2             | -0.1                 | 0.3         | 71      | 27                 | 2   | 2                   | 0                              | 0.7               | 0.4            |
| 1.6                   | -0.2             | -0.1                 | 0.4         | 69      | 26                 | 5   | 0                   | 4                              | 1.3               | 0.7            |
| 1.3                   | -0.2             | -0.1                 | 0.5         | 73      | 25                 | 1   | 1                   | 0                              | 1.0               | 0.6            |
| 0.3                   | -0.3             | -0.3                 | -0.7        | 68      | 32                 | 1   | 1                   | 0                              | 2.5               | 0.5            |
| 0.4                   | -0.2             | -0.2                 | -0.1        | 70      | 29                 | 2   | 0                   | 1                              | 1.7               | 0.4            |
| -1.0                  | -0.1             | -0.1                 | -0.8        | 52      | 17                 | 31  | 27                  | 4                              | 2.0               | 0.8            |
| -0.4                  | -0.1             | -0.3                 | -0.5        | 68      | 22                 | 10  | 6                   | 5                              | 1.6               | 0.6            |
| 0.4                   | -0.1             | -0.2                 | 0.0         | 78      | 21                 | 1   | 1                   | 0                              | 0.8               | 0.4            |
| -2.0                  | -0.1             | -0.6                 | -1.4        | 66      | 26                 | 8   | 5                   | 3                              | 1.8               | 0.7            |
| -0.9                  | -0.1             | -0.4                 | -1.4        | 75      | 16                 | 9   | 9                   | 0                              | 4.0               | 0.9            |
| 2.9                   | -0.2             | -0.3                 | 2.5         | 82      | 17                 | 1   | 1                   | 0                              | -2.0              | 0.6            |
| 0.9                   | -0.1             | -0.6                 | 0.9         | 57      | 25                 | 18  | 16                  | 3                              | 0.3               | 0.6            |

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## Appendix 1: Key statistics

Data as averages containing measurements from 1990–2010\*

|                                    |      |                 | ₩ per<br>Capita | Popu-<br>lation |                | XA  | V           |                          |
|------------------------------------|------|-----------------|-----------------|-----------------|----------------|-----|-------------|--------------------------|
| Country                            | Code | Income<br>level | Growth<br>rate  | Growth<br>rate  | Growth<br>rate |     | Contr<br>ca | ibution by<br>pital form |
|                                    |      |                 |                 |                 |                | нс  | PC          | NC                       |
| Trinidad and Tobago                | тто  | н               | -0.2            | 0.5             | 0.3            | 0.9 | 0.1         | -0.7                     |
| Central America                    |      |                 | 0.2             | 1.9             | 2.1            | 1.7 | 0.7         | -0.2                     |
| Belize                             | BLZ  | LMI             | -1.4            | 2.5             | 1.1            | 1.0 | 0.5         | -0.3                     |
| Costa Rica                         | CRI  | UMI             | 0.8             | 2.1             | 2.9            | 2.3 | 0.7         | -0.1                     |
| Guatemala                          | GTM  | LMI             | 0.3             | 2.4             | 2.7            | 2.3 | 0.6         | -0.1                     |
| Honduras                           | HND  | LMI             | -0.6            | 2.2             | 1.6            | 1.7 | 0.6         | -0.7                     |
| Mexico                             | MEX  | UMI             | 0.9             | 1.5             | 2.4            | 1.6 | 1.0         | -0.1                     |
| Nicaragua                          | NIC  | LMI             | -0.6            | 1.7             | 1.1            | 1.1 | 0.3         | -0.4                     |
| Panama                             | PAN  | UMI             | 0.6             | 1.9             | 2.5            | 1.8 | 0.9         | -0.1                     |
| El Salvador                        | SLV  | LMI             | 1.6             | 0.8             | 2.3            | 1.7 | 0.7         | 0.0                      |
| South America                      |      |                 | -0.3            | 1.4             | 1.0            | 0.9 | 0.4         | -0.2                     |
| Argentina                          | ARG  | UMI             | 0.3             | 1.1             | 1.4            | 1.0 | 0.5         | -0.1                     |
| Bolivia (Plurinational State of)   | BOL  | LMI             | -2.1            | 2.0             | -0.1           | 0.3 | 0.1         | -0.4                     |
| Brazil                             | BRA  | UMI             | 0.1             | 1.3             | 1.4            | 1.3 | 0.3         | -0.2                     |
| Chile                              | CHL  | UMI             | 1.1             | 1.3             | 2.4            | 1.6 | 0.9         | -0.1                     |
| Colombia                           | COL  | UMI             | -0.1            | 1.7             | 1.6            | 1.3 | 0.4         | -0.2                     |
| Ecuador                            | ECU  | UMI             | -0.9            | 1.7             | 0.8            | 0.9 | 0.6         | -0.7                     |
| Guyana                             | GUY  | LMI             | -0.1            | 0.2             | 0.1            | 0.0 | 0.1         | 0.0                      |
| Peru                               | PER  | UMI             | -0.6            | 1.5             | 0.9            | 0.7 | 0.4         | -0.2                     |
| Paraguay                           | PRY  | LMI             | -1.1            | 2.1             | 1.0            | 1.0 | 0.3         | -0.3                     |
| Uruguay                            | URY  | UMI             | 0.7             | 0.4             | 1.1            | 0.7 | 0.5         | 0.0                      |
| Venezuela (Bolivarian Republic of) | VEN  | UMI             | -1.1            | 2.0             | 0.8            | 1.0 | 0.2         | -0.4                     |
| Northern America                   |      |                 | 0.4             | 1.0             | 1.5            | 0.9 | 0.7         | -0.1                     |
| Northern America                   |      |                 | 0.4             | 1.0             | 1.5            | 0.9 | 0.7         | -0.1                     |
| Canada                             | CAN  | н               | 0.3             | 1.0             | 1.3            | 0.9 | 0.6         | -0.1                     |
| United States of America           | USA  | HI              | 0.6             | 1.0             | 1.6            | 0.9 | 0.8         | 0.0                      |
| Oceania                            |      |                 | -0.3            | 1.5             | 1.2            | 0.8 | 0.6         | -0.2                     |
| Australia/New Zealand              |      |                 | 0.4             | 1.3             | 1.7            | 1.0 | 0.9         | -0.2                     |
| Australia                          | AUS  | н               | 0.3             | 1.3             | 1.7            | 0.9 | 0.9         | -0.1                     |
| New Zealand                        | NZL  | HI              | 0.5             | 1.3             | 1.7            | 1.1 | 0.8         | -0.2                     |
| Melanesia                          |      |                 | -1.0            | 1.7             | 0.7            | 0.7 | 0.2         | -0.2                     |
| Fiji                               | FJI  | LMI             | 0.7             | 0.8             | 1.5            | 1.2 | 0.4         | 0.0                      |
| Papua New Guinea                   | PNG  | LMI             | -2.7            | 2.5             | -0.2           | 0.2 | 0.1         | -0.5                     |
| World                              |      |                 | 0.0             | 1.5             | 1.6            | 1.2 | 0.6         | -0.2                     |

\*For Croatia, Kazakhstan, Kyrgyzstan, Lithuania, Russian Federation, Slovenia, Tajikistan, and Ukraine the wealth account starts from 1991; for Czech Republic and Slovakia from 1992.

| ₩ adjusted per capita |                  |                      | Perce   | ntage of ir | al form | Other indicators** |                     |                                |                   |                |
|-----------------------|------------------|----------------------|---------|-------------|---------|--------------------|---------------------|--------------------------------|-------------------|----------------|
| Growth<br>rate        | Contri           | bution by adj        | ustment | нс          | PC      | NC                 | Bre                 | akdown NC                      | GDP per<br>capita | HDI            |
| -                     | Carbon<br>Damage | Oil Capital<br>Gains | TFP     |             |         |                    | Renewable resources | Non-<br>renewable<br>resources | Growth<br>rate    | Growth<br>rate |
| -3.6                  | -0.1             | 0.3                  | -3.6    | 48          | 27      | 25                 | 3                   | 22                             | 4.7               | 0.5            |
| -1.0                  | -0.1             | -0.3                 | -0.7    | 58          | 18      | 25                 | 24                  | 0                              | 2.0               | 0.9            |
| -3.3                  | 0.0              | -0.3                 | -1.6    | 28          | 11      | 61                 | 61                  | 0                              | 2.0               | 0.3            |
| -0.2                  | -0.1             | -0.2                 | -0.7    | 73          | 14      | 13                 | 13                  | 0                              | 2.6               | 0.7            |
| -0.9                  | -0.1             | -0.2                 | -0.8    | 75          | 16      | 9                  | 9                   | 0                              | 1.3               | 1.1            |
| -1.4                  | -0.1             | -0.2                 | -0.4    | 50          | 13      | 37                 | 37                  | 0                              | 1.4               | 1.0            |
| 0.9                   | -0.1             | -0.3                 | 0.4     | 60          | 27      | 13                 | 11                  | 2                              | 1.2               | 0.8            |
| -2.1                  | -0.1             | -0.4                 | -1.0    | 40          | 23      | 37                 | 37                  | 0                              | 1.4               | 1.1            |
| -1.5                  | -0.1             | -0.7                 | -1.4    | 61          | 17      | 22                 | 22                  | 0                              | 3.7               | 0.7            |
| 0.8                   | -0.1             | -0.3                 | -0.4    | 74          | 21      | 5                  | 5                   | 0                              | 2.5               | 1.3            |
| -1.3                  | -0.1             | 0.2                  | -1.1    | 39          | 13      | 48                 | 41                  | 7                              | 2.3               | 0.8            |
| -1.9                  | -0.1             | -0.1                 | -2.0    | 54          | 21      | 25                 | 21                  | 3                              | 3.3               | 0.7            |
| -2.6                  | 0.0              | 0.0                  | -0.5    | 9           | 2       | 90                 | 88                  | 2                              | 1.8               | 0.9            |
| -1.0                  | -0.1             | -0.1                 | -0.9    | 50          | 16      | 34                 | 31                  | 2                              | 1.7               | 1.0            |
| 0.3                   | -0.1             | -0.3                 | -0.5    | 65          | 14      | 21                 | 16                  | 4                              | 3.7               | 0.7            |
| -0.8                  | -0.1             | 0.0                  | -0.7    | 47          | 13      | 40                 | 34                  | 6                              | 1.7               | 0.9            |
| -1.5                  | -0.1             | 1.1                  | -1.5    | 35          | 23      | 42                 | 24                  | 19                             | 1.6               | 0.6            |
| -1.4                  | 0.0              | -0.1                 | -1.2    | 5           | 3       | 93                 | 92                  | 1                              | 3.5               | 1.1            |
| -3.1                  | 0.0              | -0.1                 | -2.4    | 27          | 11      | 61                 | 58                  | 3                              | 3.4               | 0.9            |
| -2.2                  | 0.0              | -0.2                 | -0.9    | 31          | 11      | 57                 | 57                  | 0                              | 0.7               | 0.7            |
| -0.6                  | -0.1             | -0.2                 | -1.0    | 68          | 18      | 14                 | 14                  | 0                              | 2.8               | 0.6            |
| 0.6                   | 0.0              | 2.4                  | -0.6    | 35          | 15      | 50                 | 14                  | 36                             | 0.8               | 0.8            |
| 0.8                   | 0.0              | 0.3                  | 0.1     | 62          | 19      | 19                 | 11                  | 8                              | 1.4               | 0.3            |
| 0.8                   | 0.0              | 0.3                  | 0.1     | 62          | 19      | 19                 | 11                  | 8                              | 1.4               | 0.3            |
| 1.0                   | 0.0              | 0.8                  | 0.0     | 51          | 17      | 31                 | 18                  | 13                             | 1.4               | 0.2            |
| 0.6                   | 0.0              | -0.2                 | 0.2     | 72          | 21      | 7                  | 4                   | 4                              | 1.5               | 0.3            |
| -1.0                  | 0.0              | -0.2                 | -0.5    | 45          | 17      | 38                 | 33                  | 5                              | 1.6               | 0.6            |
| -0.3                  | 0.0              | -0.1                 | -0.6    | 53          | 24      | 23                 | 15                  | 8                              | 1.8               | 0.4            |
| -0.3                  | 0.0              | -0.1                 | -0.6    | 48          | 21      | 31                 | 17                  | 14                             | 2.0               | 0.3            |
| -0.2                  | 0.0              | -0.2                 | -0.5    | 58          | 27      | 15                 | 14                  | 1                              | 1.5               | 0.5            |
| -1.8                  | -0.1             | -0.2                 | -0.4    | 37          | 9       | 53                 | 52                  | 2                              | 1.4               | 0.9            |
| 0.4                   | -0.1             | -0.4                 | 0.3     | 68          | 16      | 16                 | 16                  | 0                              | 0.9               | 0.7            |
| -4.0                  | 0.0              | 0.0                  | -1.1    | 6           | 3       | 91                 | 87                  | 4                              | 1.8               | 1.1            |
| -0.2                  | -0.1             | 0.1                  | -0.3    | 54          | 18      | 28                 | 21                  | 7                              | 2.0               | 0.8            |

 HI
 High income
 UMI
 Upper middle income

 LMI
 Lower middle income
 LI
 Low income









**Appendix 3**: Decomposition analysis of the three terms accounted in for human capital and their contribution to the changes in human capital

| Country                          | Percen | itage chang<br>199 | e with respe<br>0 | ect to | Relative contribution in percentage to human capital growth of each term |         |          |       |
|----------------------------------|--------|--------------------|-------------------|--------|--|---------|----------|-------|
|                                  | Term I | Term II            | Term III          | total  | Term I   | Term II | Term III | total |
| Afghanistan                      | 40     | 138                | 4                 | 182    | 22   | 76      | 2        | 100   |
| Albania                          | 10     | 9                  | -5                | 13     | 43   | 35      | 22       | 100   |
| Algeria                          | 35     | 57                 | -8                | 84     | 35   | 56      | 8        | 100   |
| Argentina                        | 10     | 31                 | 4                 | 45     | 23   | 68      | 9        | 100   |
| Armenia                          | 0      | -1                 | -5                | -6     | 5  | 19      | 76       | 100   |
| Australia                        | 5      | 36                 | 2                 | 44     | 12   | 83      | 5        | 100   |
| Austria                          | 16     | 11                 | 1                 | 28     | 56   | 39      | 5        | 100   |
| Bahrain                          | 39     | 219                | -2                | 256    | 15   | 84      | 1        | 100   |
| Bangladesh                       | 31     | 47                 | -4                | 74     | 38   | 58      | 5        | 100   |
| Barbados                         | 9      | 12                 | -4                | 17     | 37   | 47      | 17       | 100   |
| Belgium                          | 9      | 8                  | 5                 | 22     | 40   | 38      | 23       | 100   |
| Belize                           | 10     | 85                 | 1                 | 96     | 11   | 88      | 1        | 100   |
| Benin                            | 26     | 84                 | 0                 | 110    | 24   | 76      | 0        | 100   |
| Bolivia (Plurinational State of) | 31     | 50                 | 2                 | 84     | 38   | 60      | 3        | 100   |
| Botswana                         | 31     | 60                 | -2                | 89     | 33   | 65      | 2        | 100   |
| Brazil                           | 34     | 34                 | -1                | 67     | 49   | 49      | 2        | 100   |
| Bulgaria                         | 9      | -10                | -4                | -6     | 38   | 46      | 17       | 100   |
| Burundi                          | 17     | 58                 | -1                | 73     | 22   | 76      | 2        | 100   |
| Cambodia                         | 7      | 88                 | 2                 | 96     | 7  | 91      | 2        | 100   |
| Cameroon                         | 19     | 61                 | -1                | 79     | 24   | 75      | 1        | 100   |
| Canada                           | 16     | 27                 | -4                | 39     | 35   | 57      | 8        | 100   |
| Central African Republic         | 10     | 51                 | -1                | 61     | 16   | 82      | 1        | 100   |
| Chile                            | 19     | 39                 | 3                 | 62     | 32   | 64      | 4        | 100   |
| China                            | 26     | 26                 | -6                | 46     | 45   | 44      | 11       | 100   |
| Colombia                         | 20     | 48                 | 8                 | 76     | 26   | 63      | 11       | 100   |
| Congo                            | 6      | 70                 | 3                 | 79     | 8  | 89      | 3        | 100   |
| Costa Rica                       | 18     | 69                 | 0                 | 87     | 20   | 79      | 0        | 100   |
| Côte d'Ivoire                    | 21     | 56                 | -2                | 75     | 26   | 71      | 3        | 100   |
| Croatia                          | 11     | 2                  | -6                | 6      | 57   | 9       | 34       | 100   |
| Cuba                             | 12     | 10                 | -8                | 14     | 39   | 34      | 27       | 100   |
| Cyprus                           | 11     | 58                 | 4                 | 72     | 15   | 80      | 5        | 100   |
| Czech Republic                   | 12     | 11                 | -1                | 22     | 49   | 46      | 5        | 100   |
| Democratic Republic of the Congo | 5      | 88                 | 2                 | 96     | 5  | 92      | 3        | 100   |
| Denmark                          | 5      | 5                  | -5                | 6      | 35   | 33      | 32       | 100   |
| Dominican Republic               | 16     | 47                 | -2                | 61     | 24   | 72      | 4        | 100   |

Note: Term I = human capital embodied per person; Term II = adults who reached the average education level; and Term III = capitalized labor compensation per unit of human capital. Negative changes in Term III are considered in absolute numbers to estimate the relative contribution among Term I, II, and III.

| Country                          | Percentage change with respect to 1990 |         |          |       | Relative contribution in percentage to human capital growth of each term |         |          |       |
|----------------------------------|--|---------|----------|-------|--|---------|----------|-------|
|                                  | Term I                                 | Term II | Term III | total | Term I   | Term II | Term III | total |
| Ecuador                          | 11                                     | 53      | 2        | 67    | 17   | 79      | 4        | 100   |
| Egypt                            | 29                                     | 46      | -6       | 69    | 36   | 56      | 8        | 100   |
| El Salvador                      | 35                                     | 21      | 0        | 56    | 63   | 37      | 1        | 100   |
| Estonia                          | 22                                     | -12     | -5       | 4     | 56   | 30      | 14       | 100   |
| Fiji                             | 9                                      | 32      | 0        | 41    | 21   | 78      | 0        | 100   |
| Finland                          | 16                                     | 8       | -5       | 19    | 56   | 27      | 17       | 100   |
| France                           | 29                                     | 7       | -5       | 31    | 71   | 17      | 12       | 100   |
| Gabon                            | 38                                     | 69      | -2       | 105   | 35   | 63      | 2        | 100   |
| Gambia                           | 21                                     | 78      | 1        | 99    | 21   | 78      | 1        | 100   |
| Germany                          | 39                                     | 2       | -1       | 40    | 92   | 6       | 3        | 100   |
| Ghana                            | 14                                     | 72      | -4       | 83    | 16   | 80      | 4        | 100   |
| Greece                           | 22                                     | 14      | -2       | 34    | 57   | 37      | 6        | 100   |
| Guatemala                        | 15                                     | 64      | 3        | 82    | 19   | 78      | 3        | 100   |
| Guyana                           | 17                                     | 0       | -1       | 16    | 94   | 2       | 4        | 100   |
| Haiti                            | 19                                     | 51      | 5        | 75    | 25   | 68      | 7        | 100   |
| Honduras                         | 31                                     | 68      | 1        | 99    | 31   | 69      | 1        | 100   |
| Hungary                          | 27                                     | -3      | -3       | 20    | 81   | 9       | 10       | 100   |
| Iceland                          | 21                                     | 29      | -6       | 44    | 38   | 52      | 10       | 100   |
| India                            | 18                                     | 43      | -5       | 57    | 28   | 65      | 7        | 100   |
| Indonesia                        | 21                                     | 35      | -4       | 52    | 35   | 57      | 7        | 100   |
| Iran (Islamic Republic of)       | 46                                     | 64      | -10      | 100   | 39   | 53      | 9        | 100   |
| Iraq                             | 18                                     | 77      | -3       | 91    | 18   | 79      | 3        | 100   |
| Ireland                          | 12                                     | 37      | 5        | 54    | 22   | 69      | 9        | 100   |
| Israel                           | 8                                      | 78      | 3        | 88    | 9  | 88      | 3        | 100   |
| Italy                            | 16                                     | 6       | -7       | 15    | 54   | 21      | 25       | 100   |
| Jamaica                          | 27                                     | 17      | -10      | 34    | 50   | 32      | 18       | 100   |
| Japan                            | 15                                     | 7       | -10      | 11    | 46   | 22      | 32       | 100   |
| Jordan                           | 37                                     | 99      | -9       | 127   | 26   | 68      | 6        | 100   |
| Kazakhstan                       | 25                                     | 1       | 0        | 26    | 94   | 5       | 1        | 100   |
| Kenya                            | 20                                     | 84      | -4       | 101   | 19   | 78      | 3        | 100   |
| Kuwait                           | 4                                      | 42      | 2        | 49    | 9  | 87      | 4        | 100   |
| Kyrgyzstan                       | 6                                      | 32      | 1        | 38    | 15   | 82      | 3        | 100   |
| Lao People's Democratic Republic | 18                                     | 63      | 2        | 83    | 22   | 76      | 2        | 100   |
| Latvia                           | 22                                     | -12     | -7       | 3     | 53   | 30      | 17       | 100   |
| Lesotho                          | 21                                     | 34      | -11      | 44    | 31   | 52      | 17       | 100   |

**Appendix 3**: Decomposition analysis of the three terms accounted in for human capital and their contribution to the changes in human capital

| Country             | Percentage change with respect to 1990 |         |          |       | Relative contribution in percentage to human capital growth of each term |         |          |       |
|---------------------|--|---------|----------|-------|--|---------|----------|-------|
|                     | Term I                                 | Term II | Term III | total | Term I   | Term II | Term III | total |
| Liberia             | 37                                     | 90      | 8        | 135   | 27   | 67      | 6        | 100   |
| Lithuania           | 18                                     | -5      | -6       | 6     | 61   | 17      | 22       | 100   |
| Luxembourg          | 12                                     | 34      | 9        | 55    | 21   | 62      | 17       | 100   |
| Malawi              | 19                                     | 49      | -2       | 66    | 28   | 70      | 3        | 100   |
| Malaysia            | 37                                     | 65      | -7       | 94    | 34   | 60      | 7        | 100   |
| Maldives            | 27                                     | 73      | 14       | 114   | 24   | 64      | 12       | 100   |
| Mali                | 14                                     | 74      | 2        | 89    | 16   | 82      | 2        | 100   |
| Malta               | 17                                     | 25      | -1       | 40    | 40   | 59      | 2        | 100   |
| Mauritania          | 24                                     | 75      | 3        | 102   | 24   | 73      | 3        | 100   |
| Mauritius           | 14                                     | 26      | -5       | 35    | 32   | 58      | 10       | 100   |
| Mexico              | 30                                     | 40      | -3       | 67    | 42   | 55      | 4        | 100   |
| Mongolia            | 8                                      | 47      | -2       | 53    | 14   | 82      | 4        | 100   |
| Morocco             | 23                                     | 35      | -7       | 51    | 35   | 54      | 11       | 100   |
| Mozambique          | 10                                     | 69      | -2       | 77    | 13   | 85      | 3        | 100   |
| Myanmar             | 21                                     | 27      | -3       | 45    | 41   | 53      | 6        | 100   |
| Namibia             | -1                                     | 73      | 2        | 74    | 2  | 96      | 2        | 100   |
| Nepal               | 18                                     | 60      | 1        | 79    | 23   | 76      | 1        | 100   |
| Netherlands         | 8                                      | 11      | 4        | 23    | 35   | 48      | 17       | 100   |
| New Zealand         | 10                                     | 33      | 3        | 46    | 22   | 72      | 7        | 100   |
| Nicaragua           | 28                                     | 46      | 1        | 75    | 37   | 62      | 1        | 100   |
| Niger               | 11                                     | 99      | 8        | 118   | 9  | 84      | 7        | 100   |
| Nigeria             | 11                                     | 62      | 0        | 73    | 15   | 84      | 1        | 100   |
| Norway              | 16                                     | 15      | -2       | 29    | 49   | 45      | 6        | 100   |
| Pakistan            | 34                                     | 60      | 2        | 97    | 35   | 62      | 2        | 100   |
| Panama              | 22                                     | 54      | 1        | 77    | 28   | 70      | 1        | 100   |
| Papua New Guinea    | 11                                     | 66      | 0        | 77    | 15   | 85      | 0        | 100   |
| Paraguay            | 32                                     | 56      | -2       | 86    | 35   | 63      | 2        | 100   |
| Peru                | 20                                     | 44      | 6        | 70    | 29   | 63      | 9        | 100   |
| Philippines         | 17                                     | 56      | -3       | 69    | 22   | 74      | 4        | 100   |
| Poland              | 12                                     | 11      | -9       | 14    | 39   | 34      | 27       | 100   |
| Portugal            | 12                                     | 10      | -3       | 19    | 46   | 42      | 12       | 100   |
| Qatar               | 25                                     | 367     | 5        | 397   | 6  | 92      | 1        | 100   |
| Republic of Korea   | 24                                     | 20      | -10      | 34    | 45   | 37      | 18       | 100   |
| Republic of Moldova | 11                                     | -11     | -20      | -19   | 27   | 26      | 48       | 100   |
| Romania             | 10                                     | 0       | -3       | 6     | 72   | 3       | 24       | 100   |

| Country                            | Percentage change with respect to 1990 |         |          |       | Relative contribution in percentage to human capital growth of each term |         |          |       |
|------------------------------------|--|---------|----------|-------|--|---------|----------|-------|
|                                    | Term I                                 | Term II | Term III | total | Term I   | Term II | Term III | total |
| Russian Federation*                | 16                                     | 3       | -6       | 13    | 65   | 12      | 24       | 100   |
| Rwanda                             | 18                                     | 50      | 2        | 70    | 25   | 72      | 3        | 100   |
| Saudi Arabia                       | 34                                     | 94      | -11      | 118   | 24   | 68      | 8        | 100   |
| Senegal                            | 21                                     | 69      | 0        | 90    | 23   | 76      | 1        | 100   |
| Serbia                             | 11                                     | 7       | -3       | 16    | 53   | 34      | 13       | 100   |
| Sierra Leone                       | 15                                     | 41      | 4        | 60    | 25   | 68      | 7        | 100   |
| Singapore                          | 30                                     | 80      | 1        | 111   | 27   | 72      | 1        | 100   |
| Slovakia                           | 6                                      | 18      | -3       | 21    | 21   | 67      | 12       | 100   |
| Slovenia                           | 9                                      | 14      | 1        | 24    | 37   | 59      | 4        | 100   |
| South Africa                       | 19                                     | 46      | -8       | 56    | 25   | 63      | 12       | 100   |
| Spain                              | 39                                     | 22      | 2        | 64    | 61   | 35      | 4        | 100   |
| Sri Lanka                          | 22                                     | 27      | -6       | 43    | 40   | 50      | 10       | 100   |
| Sudan (former)                     | 13                                     | 68      | 0        | 81    | 16   | 84      | 1        | 100   |
| Swaziland                          | 20                                     | 48      | -11      | 57    | 26   | 61      | 14       | 100   |
| Sweden                             | 12                                     | 9       | -4       | 17    | 50   | 35      | 15       | 100   |
| Switzerland                        | 2                                      | 17      | -2       | 17    | 11   | 80      | 9        | 100   |
| Syrian Arab Republic               | 9                                      | 83      | -13      | 80    | 8  | 79      | 12       | 100   |
| Tajikistan                         | -1                                     | 44      | 1        | 44    | 2  | 96      | 2        | 100   |
| Thailand                           | 21                                     | 26      | -9       | 38    | 38   | 47      | 16       | 100   |
| Тодо                               | 22                                     | 69      | 1        | 93    | 24   | 75      | 1        | 100   |
| Trinidad and Tobago                | 16                                     | 26      | 2        | 44    | 37   | 59      | 4        | 100   |
| Tunisia                            | 33                                     | 37      | -9       | 60    | 41   | 47      | 11       | 100   |
| Turkey                             | 21                                     | 40      | -13      | 48    | 28   | 54      | 17       | 100   |
| Uganda                             | 23                                     | 79      | 0        | 102   | 22   | 77      | 0        | 100   |
| Ukraine                            | 14                                     | -6      | -6       | 2     | 53   | 23      | 24       | 100   |
| United Arab Emirates               | 48                                     | 455     | -1       | 502   | 9  | 90      | 0        | 100   |
| United Kingdom                     | 14                                     | 8       | -3       | 19    | 57   | 32      | 11       | 100   |
| United Republic of Tanzania        | 18                                     | 66      | -1       | 83    | 21   | 78      | 1        | 100   |
| United States of America           | 8                                      | 24      | -4       | 28    | 23   | 67      | 10       | 100   |
| Uruguay                            | 10                                     | 8       | 3        | 21    | 48   | 37      | 15       | 100   |
| Venezuela (Bolivarian Republic of) | 23                                     | 56      | 0        | 79    | 29   | 71      | 0        | 100   |
| Vietnam                            | 26                                     | 44      | -3       | 68    | 36   | 60      | 4        | 100   |
| Yemen                              | 37                                     | 102     | 2        | 141   | 26   | 72      | 2        | 100   |
| Zambia                             | 22                                     | 57      | 1        | 80    | 28   | 71      | 1        | 100   |
| Zimbabwe                           | 20                                     | 23      | -4       | 39    | 43   | 49      | 8        | 100   |

# The IWR and policy lessons

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## KEY MESSAGES

**Countries striving to improve their** citizens' well-being – and do so sustainably – should reorient economic policy planning and evaluation away from targeting GDP growth as a primary objective toward incorporating inclusive wealth accounting as part of a sustainable development agenda.

**Investments in human capital** – in particular education – would generate higher returns for IW growth, as compared to investments in other capital asset groups, in countries with high rates of population growth.

**Investments in natural capital, in** particular agricultural land and forest, can produce a twofold dividend: First, they can increase IW directly; second, they can improve agricultural resiliency and food security to accommodate anticipated population growth.

**Investments in renewable energy can** produce a triple dividend: First, they can increase IW directly by adding to natural and produced capital stocks; second, they improve energy security and reduce risk due to price fluctuations for oil-importing countries; third, they reduce global carbon emissions and thus carbon-related damages.

**Investments in research and development** to increase total factor productivity, which decreased in 65 percent of countries, can immediately contribute to growth in inclusive wealth in nearly every country.

**Countries should expand the asset** boundary of the present System of National Accounts (SNA), which currently captures only 18 percent of a country's productive base, to include human and natural capital, which are now measured only through satellite accounts, if at all.

#### 1. Introduction

In the majorities of countries, growth-led economic policy planning is incompatible with sustainable development. Policy-makers are concerned primarily with increasing market activity as a means to increase employment and improve quality of life. Success in this strategy is measured in gains in gross domestic product (GDP). This strategy, however, is facing increasing scrutiny, as evidence mounts that countries are improving GDP growth rates but seeing little or no gains in employment, stagnating incomes for the majority of citizens, and often little improvement in other measures of well-being, such as health outcomes, access to quality education, and economic security. The "trickle-down" effect from economic production (measured in GDP) to employment and wages, and quality of life, is no longer happening in modern economies (STIGLITZ ET AL. 2009, EASTERLIN 2003, DEATON AND KAHNENAM 2010, JACKSON 2009).

When we talk about growth-led economic policy planning, we are generally referring to macroeconomic fiscal and monetary policies, and within those, various levels of sector-specific investment and regulatory policy, in areas such as agriculture, industry, health, transportation, and energy.

At the same time, many governments have introduced efforts toward improving sustainability or sustainable development, most of which remain secondary to central policy planning priorities, and are usually focused solely on environmental protection. The mantra has always been: growth will come first, and growth will enable the time, effort, and money necessary to protect the environment.

As we are now seeing, the foundations of both sides of this equation are crumbling: growth in GDP will not assure nations of greater well-being; and there is far more to sustainability than environmental protection. To right the course, we must first broaden our understanding – and evaluation – of wellbeing and progress beyond GDP, and extend sustainable development beyond environment to encompass the full spectrum of well-being.

Results from Chapter I demonstrate how a focus on production growth alone has led many countries onto an unsustainable growth trajectory. The Inclusive Wealth Index (W) aims to help economists, policy-makers, and anyone interested in sustainability better understand what the basis of inclusive wealth (IW) is, and how countries are doing in assuring sustainable growth in wealth over the long term.

This chapter is not meant to be an exhaustive analysis of the results presented in Chapter I, nor does it present a complete range of associated policy prescriptions. Rather, it demonstrates how the Inclusive Wealth Report (IWR) can be used by countries to inform and guide policy-making today.

The chapter begins with some general policy lessons relating to human capital that can be taken from the results presented in Chapter I. Although human capital comprises both health and education, this report excludes health from the computation of human capital due to inconsistencies in methodology and lack of sufficient data. Therefore, our human capital figures essentially represent education, and the concrete returns for human well-being of education. However, some key policy considerations relating to health are shown in Box I.

Following sections cover in more detail two specific social problems countries face today – food and energy security – and explain how the IWR can be used to guide policy design in order to address them. Specific country results used to demonstrate significant changes in the respective capitals pertaining to the discussion are used purely for illustration, and should not be regarded as criticizing or endorsing the policies of any specific country.

In addition, this chapter includes a section on national accounts and policy suggestions for revising, modifying, and developing national accounts. The section provides arguments for including human capital and natural capital within the asset boundaries of core

#### BOX 1 Health policy and the IWR

Health plays a key role in both sustainable development, as well as the inclusive wealth of nations, as demonstrated quantitatively in this report (see Chapter 5). Quantifying health as a form of human capital is an important advancement, one that opens up new opportunities for policy in sustainable development. Yet, it is important to recall that health only recently became a "center stage" priority, with the formulation of the Millennium Development Goals (MDGs).

However, health development has sometimes been seen as expenditure, the magnitude of which has occasionally been a hindrance to increased action and investment. We have been slow to adopt the language of investment in health. This is changing, however; and with the formulation of the post-2015 Sustainable Development Goals (SDGs), the change is timely.

Nevertheless, making the case for health as a vital component of sustainable development is not yet a question of settled science, and it is one that is being answered in various ways by diverse developments in an emerging field. Within the UN system, the IWR and the Human Development Index (HDI) provide two examples. In the context of the broader international community, two other notable efforts stand out: the Human Capital Project of the World Economic Forum, and the Social Progress Index of the Social Progress Imperative.

While each of these methods arguably has its advantages and disadvantages, the benefit of the IWR is that it integrates health as an integral part of the inclusive wealth of nations, something to be invested in rather than spent on, and which yields return in national wealth and well-being.

It also gives a clear conceptual framework for clarifying what is meant by "the investment case for health", which already contains within its proposition health as an asset in the wealth of a nation. These advances are useful, moreover, since they address governments and donors in terms adapted to their needs, such as demonstrating accountability in decision-making regarding the use of resources. Health technology assessment (HTA) is a growing institutional focus for such concerns. In 2014, the World Health Assembly passed a resolution calling for increased use of health technology assessment in the context of a renewed focus on universal health coverage.

As countries scale up coverage with a range of health services, they need to decide which population groups should be covered first. Some choose to prioritize specific poor and vulnerable communities, offering selected services at minimal cost, while others choose to assure universal access to a limited set of services for a period of time. When new money becomes available, either through efficiency savings or the raising of new finance, additional choices must be made about whether to cover more people for the existing set of services, cover more services for the people already covered (and/or increase quality), or reduce out of pocket payments further for people currently covered with a set of services.

These decisions are difficult, and there is no single recipe for success. One of the roles of the IWR in these discussions is to provide a clear conceptual framework for understanding that, when health must compete for scarce resources, its benefits are not only intrinsic, but also instrumental to progress in many other domains.

economic accounts. It also demonstrates why these capital asset groups' current placement within satellite accounts misrepresents 82 percent of the assets that contribute to IW, while the mere 18 percent captured by the present system of national accounts (produced capital) might actually be providing false interpretations of progress.

#### 2. Policy lessons for education

As shown in Chapter 1, only 85 of 140 countries – 60 percent – were found to be on a sustainable growth trajectory in terms of inclusive wealth. The remaining 55 countries are unable to maintain a productive base to accommodate the present state of well-being, nor increases of

#### FIGURE 1

Average education levels attained across countries (in years)





well-being (see Chapter I). That is to say, the present consumption patterns of these countries are not sustainable.

Countries on unsustainable trajectories have essentially two available options: they can either increase investments to post positive gains in the rate of IW growth, or reduce consumption levels to levels which their productive base can maintain (ARROW ET AL. 20012).

Countries aiming to increase their productive base growth rate based on the results of Chapter I can invest in human and/or natural capital depending on the rate of return of these capital asset bases. It was found that investment in produced capital provides the lowest rate of return for the majority of countries.

Human capital, which in this IWR 2014 is primarily education, was found to be the greatest component of IW – nearly 54 percent of total inclusive wealth – in about 70 percent of countries. The largest contributions to IW – what we call inclusive investment – were also made into developing human capital. However, many countries were not investing proportionally in natural capital based on its contribution to IW.

Most investments were in the traditional produced capital category, consisting of infrastructure, roads, buildings, etc. This might be understandable as it is the only category recorded explicitly in the core economic accounts a country maintains (the System of National Accounts). The human and natural capital components, including education, are relegated to so-called satellite accounts, if at all recorded.

Within education, there are a number of target areas in which investments can be made. The education-related accounts within the IW refer to years of schooling, gender dimensions, demographic profiles, and wage returns on education. A large number of countries, as shown in Figure 1, have a low number of average years of schooling, and thus significant potential for returns on investment. In investing in education, countries can improve present wellbeing, future productivity and income, and

#### FIGURE 2

Growth rates of investment in education per capita across countries (in percentage), 2010/2009



higher levels of long-term inclusive wealth – and thus sustainability.

Figure I shows that most countries have not reached average education levels that include post-secondary schooling. Canada, Australia, the United States, and Norway have an average of 12 to 14 years of education. The majority of countries have achieved education levels that include secondary schooling (6 to 12 years) while most of the African and South Asian countries have only reached education levels commensurate with primary school levels (o to 6). The potential to increase human capital through further investment in education is therefore high.

However, there is clear indication of a slowing-down of investment in education, as shown in Figure 2. Six countries show a rate of growth in inclusive investment in education per capita greater than 2 percent. The majority of them fall in the -4 to 1 percent rate, which might explain the relatively low growth rate of IW in many of the countries. Several reasons can be assumed, but a key factor is likely the way education is presently factored in national accounts. Investments in education are considered expendi-

| Key        |         |
|------------|---------|
| > 2        | –4 to 1 |
| 1.5 to 2   | < -4    |
| 📃 1 to 1.5 | no data |

tures, since contributions to GDP are not direct (as with IW). Governments whose primary goal is GDP growth, particularly those lacking advanced educational monitoring and assessment facilities, will often focus on investments which are directly reflected in GDP, such as in produced capital.

Although produced capital is important, our results indicate that many countries place disproportional emphasis on the growth rate of produced capital compared to those of human and natural capital, if one takes into consideration the relative weight these capitals have towards overall well-being.

The recommendation of the IWR to increase investment in human capital resonates well with the proposed Sustainable Development Goals on education (UN-OWG 2014). The education inclusive wealth accounts go further than simply tracking literacy levels, providing information on the actual "value" generated by literacy rates (among other educational outcomes) for inter-temporal human well-being.

#### 3. Food security, agriculture, and land

Food security is a central policy priority for many countries today. While food security is a complex issue, one critical factor is availability of suitable arable land. It is estimated that an additional 3 to 5 million hectares of cropland might be required to feed the growing world population over the next 30 years (WIRSENIUS ET AL. 2010).

The competition for land to produce not only food grains for direct consumption, but also feed grain for livestock and horticultural food products, has been a continuous struggle, one recently compounded by the addition of a third competitive sector: biofuels.

The demand for agricultural land changes as the diet of populations changes. The trend seen across countries of late has been that as incomes grow, diets shift from grains to meat and vegetables (ALEXANDRATOS AND BRUINSMA 2012, TILMAN ET AL. 2011). This in turn suggests a different use of agricultural land, and the amount of land that needs to be made available.

At the same time, countries hoping to mitigate climate change and reduce (or compensate for) greenhouse gas emissions have increasingly resorted to biofuels. This has had the effect of shifting grains and feedstocks such as maize, palm oil, and sugarcane, among others, toward use as biofuels. For example, the United States shifted 27 percent of its maize production from use as feed to the production of ethanol (UNEP 2012). The long term impacts of these shifting trends on food supply are still not certain, but increasing demands for land for multiple uses, as well as for grains for multiple end uses, will certainly have an impact on food supply.

Further, increases in social conflict incidence recorded across much of the world have often been attributed to rising prices of food commodities, which in turn have been driven by a combination of droughts (drought frequency is expected to increase with climate change), shifts towards biofuels, and encroachment of arable crop land by urbanization (IPCC 2014).

In summary, key factors influencing food demand are:

- growing global population;
- changing demographics with shifting dietary-driven demands for grains, meat, and vegetables; and
- growing demand for biofuels.

The main variables determining supply are:

- decreasing supply of suitable arable land, as cropland is lost to urbanization and other alternate uses due to low valuation of agriculture in national accounting systems;
- increasing rate of land degradation caused by intensive unsustainable practices;
- diminishing water supplies;
- decreasing rate of technological innovations to boost productivity; and
- decreasing ecosystem services supporting food production from declining forest ecosystems.

Figure 3 shows the percentage changes in agriculture land (cropland) for some of the main food growing areas in the world. The lack of data for many African countries highlights a knowledge gap that needs to be closed quickly, as land on the continent is often converted to cropland for producing food for export to other major food demand centers.

An important trend seen at the global level is the reduction of the inclusive wealth of cropland in traditional major cereal-exporting countries like Canada, the United States, and Australia. The exceptions were mostly in South America, where cropland has increased, in particular in Argentina and Brazil.

Similar trends are observed in changes in pastureland as shown in Figure 4. Most regions around the globe have seen significant declines

#### FIGURE 3 Change in cropland wealth per capita (in percentage), 2010/1990



in the inclusive wealth of pasturelands<sup>1</sup>, even as the demand for meat increases. This might be explained by the transition from extensive gazing of livestock to significantly more-intensive livestock management, which requires less land. The impacts on the inclusive wealth are still unclear, as the increase in produced capital arising from intensification of livestock management cannot be isolated from the present inclusive wealth accounts. It is a potential area of further refinement for future reports to address.

Caution is warranted in interpreting data showing increased cropland, since such increases might be the result from land cleared for biofuels. Figure 5 shows the changes in cropland wealth per capita for USA, Brazil, China, India, and Canada. Unfortunately, the data available does not make the distinction between agricultural land for food crops and agricultural land for biofuels. Furthermore, when using IW to analyze food policy, we must not only pay attention to land availability, but also to land productivity. Here, linking total factor productivity (TFP) into the agricultural



component with IW would provide valuable and insightful findings for food policy. This can be done, for example, in a country such as India, through investment in increasing the value of per unit cropland by investing in irrigation, fertilizers, soil management, and crop yield technologies (BAGLA 2014).

A key value-added of using inclusive wealth accounts is in tracking total land change by observing changes in other categories of land. For example, in the case of Brazil and China, the inclusive wealth of forestland, which includes timber and non-timber forest products per capita, has decreased and increased, respectively.

Expansion might also entail replacing forests with cropland; this information can be derived from the IWR, too. The trade-off between cropland and forest can be observed in many

<sup>1</sup> Inclusive wealth of pastureland implies the change in physical land under pasture multiplied by the value of the pastureland. These are on a per capita basis.

#### FIGURE 4

Changes in pastureland wealth per capita (in percentage), 2010/1990



| кеу      |           |
|----------|-----------|
| > 5      | -28 to 10 |
| 📕 0 to 5 | < -28     |
| -10 to 0 | 🗌 no data |

countries. In fact, the most important driver of land conversion into agriculture in the tropics has been the conversion of forests to agriculture (LAMBIN AND MEYFROIDT 2011).

Information on land use as provided by the IWR, together with data on actual food production and exports, can be extremely useful in determining what trade-offs a country has been managing, and where future investment is needed. Such investment could be in restoring degraded lands, bringing irrigation to drought-affected areas, or elsewhere.

Similarly, the data provided by the IWR, together with data on agricultural production, can be used to gather information on productivity and henceforth options to increase food production. In fact results show that TFP has been decreasing in 87 countries. Increasing investments in areas such as technology and efficiency measures – thought to make up the bulk of TFP – might be used to increase agricultural productivity and thus the IW of agricultural-based economies, particularly those for which natural capital constitutes at least one third of the total inclusive wealth base. Investments in agricultural technologies, such as drought resistant seed varieties, soil improving technologies, and solar energy sources, are some options that might increase the productivity of the agricultural sector. This could help mitigate shifts from agriculture to manufacturing (based on the proportion of these sectors to GDP) that are underway in most countries. The picture changes when the values of these sectors are compared against IW.

The IWR also includes data on pastureland, which provides information on livestock production-related impacts on grain production. This also has implications for policies to ensure food security, as well as nutrition improvement efforts, which are gaining importance in developing countries.

The illustrations above show that the IWR can be a useful policy tool by itself for illustrating changes in the underlying



asset base necessary for food production. In tandem with other sources of information, it is even more powerful, providing information on issues such as productivity potentials, or optimal land use options that include both market-driven and non-market opportunities.

The IWR also provides insights into trade-offs across competing uses of land, and where future investments are needed if a primary policy objective is food security versus, for example, forest preservation. The ability to highlight trade-offs among the various assets comprising the productive base of an economy, in the pursuit of social, environmental, and economic objectives, is one of the IWR's key strengths.

# 4. Energy policy (non–renewable and renewable): produced and natural capital investment

A sustainable and secure supply of energy is essential for any country's development aspirations. Energy is not only important for supporting activities that lead to economic growth, but also for key aspects regarding human development and social transformations. Energy is central to improving life quality for a population, as it is a primary and required factor in providing essential humans needs, such as education, health, and infrastructure.

The link between energy and wealth is straightforward. In the IWR, natural capital captures changes in fossil fuels, which currently account for approximately 81 percent of energy supply. Increase in demand for energy translate to a direct negative impact on natural capital through the depletion of finite fossil fuel deposits, and thus negative impact on the overall inclusive wealth of a country.



#### FIGURE 6

Energy projections by energy source (in quadrillion Btu) (Source: EIA 2013)

However, the effects of energy on the other forms of capital must be considered as well. Looking at elasticity,

which measures how responsive (or sensitive) one variable is to a change in another, it is possible to discern the relationship between energy and wealth, as well as with its components.

For the IWR 2014 we ran simple regression analyses using our sample of 140 countries to estimate elasticities. Although these preliminary results are rough, and thus seen as first approximations, they are nonetheless illuminating. First, they suggest that an increase of 1 percent in energy consumption is associated with a positive variation of 0.64 percent in produced capital.<sup>2</sup> Part of this result can in practice be explained by the impact that energy consumption has on the production of produced capital. When it comes to human capital, a I percent increase in energy use is associated with a positive variation of 0.4 percent in human capital. This result could be practically explained with direct benefits of electricity use, for instance, to educational outcomes thanks to increased hours and flexibility for studying, as well as greater numbers of schools. Clearly other variables also contribute to this, but there certainly seems to be a degree of direct correlation between energy use and growth in different forms of capital.

The decision to switch from fossil fuels to alternative energy sources, which lead to reduced detrimental impact on natural capital, will be made when the marginal change in natural capital stock is greater than, or equal to, the marginal changes in produced and human capital, accruing from the energy used to produce these capitals, resulting in an overall negative impact on wealth (all other things being held constant). However, the equation becomes a bit more complicated when renewable energy sources are introduced in the analysis, also contributing to changes in natural and produced capital.

Kev

Oil and other liquids

Coal

Natural gas

Renewables Nuclear

<sup>2</sup> Estimations based on Granger causality tests also indicated bi-directional Granger causality between energy and produced capital. Results suggest that an increase of 1 percent in produced capital is associated with a positive variation of 0.62 percent in energy consumption.



#### FIGURE 7

Energy demands until 2040 (in quadrillon Btu) (Source: EIA 2013)

#### 5. Scarcity and the need for change

Over the past two decades, the renewable energy sector has gained momentum. However despite these improvements, approximately 81 percent of the current global energy supply comes from fossil fuels<sup>3</sup>. It is still the fastest-growing energy source in the world, increasing 2.5 percent per year, together with nuclear power. Despite the recent increases in renewable energy use worldwide, fossil fuels are still expected to supply about 80 percent of the total energy through 2040, when the energy consumption is projected to increase 56 percent compared to 2010.

A projection of the global demand for energy sources is given in Figure 6. Demand for fossil fuels (oil and other liquids, coal, and gas) is expected to dominate the energy sector for the next 50 years. This will represent an increase of 46 percent

in energy-related carbon dioxide emissions in the same period (IEA 2013).

Kev

World

China India

More than 85 percent of the energy demand from 2010 and 2040 comes from developing countries, reflecting growing populations and industrial expansion. Although there has been some degree of decoupling of energy use from economic growth, many rapidly-developing countries such as China and India can expect to see continued increase in demand for energy, as shown in Figure 7.

The main sources of energy, including oil, coal, and gas, are captured within the natural capital category. They supply 32 percent, 29 percent, and 21 percent of the world energy mix, respectively. Figure 6 shows increases in the demand for the various categories of fuels, indicating decreasing stocks of fossil fuels and subsequently decreasing natural capital.

According to the BP Statistical Review of World Energy (2014), if the world production of coal, natural gas, and crude oil continues at

<sup>3</sup> According to the International Energy Agency, in 2011, 31.5 percent of the world's energy supply was from oil, 21.3 percent from natural gas, 28.8 percent from coal and peat, 10 percent from biofuels and waste, 5.1 percent from nuclear sources, 2.3 percent from hydro sources, and 1 percent from others.



the rate of 2013, the remaining time for these reserves to be depleted are 113 years in the case of oil, 55 years for coal, and 53 years for gas as shown in Figure 8.<sup>4</sup>

Besides being the energy source that risks suffering the most rapid scarcity among the three main types, oil is responsible for about 36 percent of carbon dioxide emissions from fuel combustion.<sup>5</sup> Another aspect of relevance for policy-makers, especially from countries without oil reserves, concerns its price volatility. As shown in the Figure 9, in the last years, crude oil now costs more than it has in 100 years (in dollars of 2013), surpassing even the shocks of the late 1970's. The declining rates of oil per capita in constant U.S. dollars are also alarming. The figure 10 shows that the change rates from 2010, compared to 1990, have varied from -15 percent to more than -66 percent worldwide.

#### 6. The IW and renewable energy

As mentioned earlier, energy, while being an essential factor for improving well-being, imposes a policy trade-off by causing externalities, mainly to the environment. The key driving force in declining wealth per capita followed by population growth is the depreciation of natural capital. The Adjusted Inclusive Wealth<sup>6</sup>

<sup>4</sup> The reserves-to-production rate is calculated by dividing remaining reserves at the end of the year by the production in that year.

<sup>5</sup> Estimation from the International Energy Agency 2012.

<sup>6</sup> IW adjusted shows the results when climate change, oil capital gains, and TFP are included in the index. The results are available in Chapter 1, sections 3.4.

### FIGURE 10 Change in oil wealth per capita (in percentage), 2010/1990



confirms that most countries (134 countries out of 140) would suffer to some degree from climate change.

The Inclusive Wealth Index (*W*) also shows that for technological change, total factor productivity (TFP)<sup>7</sup> has been negative in 65 percent of the countries – TFP being the most influential component in affecting countries' inclusive wealth in a negative way.

Energy invariably requires the use of natural resources. Holding everything else constant, the increasing extraction of fossil fuels results in the decline of natural capital, which leads to an overall decline in IW. In the face of increasing scarcity, the natural direction should thus be toward investments in renewable resources – meeting energy production requirements through renewable energy, such as solar, biomass, wind, and others. Future IW accounts will have to be revised to reflect increases in these sectors, either within the natural capital or produced capital components.

| Кеу             |            |
|-----------------|------------|
| <b>&gt;</b> -15 | -66 to -49 |
| ■ -32 to -15    | < -66      |
| -49 to -32      | 🗌 no data  |

New investments in technology should be allocated in renewable energy such that the positive marginal benefits to IW from produced capital and components of natural capital related to renewable energy capital offset the marginal costs from natural capital declines driven by fossil fuels extraction, as illustrated by Figure 13. These results are underscore the urgent need for investments in new technologies in order to improve the efficiency and alternatives for energy.

Considering this framework, allocating resources to improve energy efficiency will result in overall improvement of wealth in every country. Chapter I clearly illustrates the decline in natural capital in the majority of countries, much of which can be attributed to the decline in fossil fuels. Because energy is a vital part of

<sup>7</sup> We used therefore total factor productivity (TFP) as a proxy variable of technological change. TFP growth measures the change in aggregated output (GDP) that cannot be explained by the growth rate of observable inputs or capital assets.



#### FIGURE 11

Germany: natural capital per capita - annual growth rates



#### BOX 2 Case: Germany's energy transformation

Germany is considered the world's first major renewable energy economy, having one of the most innovative renewable energy sectors in the world. Figure 12 shows the share of electricity produced from renewable energy has increased from 6.3 percent in 2000 to 31 percent in the first half of 2014, mainly from solar, wind, and biogas. The federal government designated a significant change in energy policy in 2011, known as "Energiewende", or energy transformation, focusing particularly on offshore wind farms and distributed generation rather than centralized energy sources.

Natural capital has generally been improving since 2001, although

#### **FIGURE 12**

Germany: consumption of solar energy (terawatts-hours). Source: BP Statistical Review of Energy 2014

development, it is not possible to reduce energy demand significantly.

Currently, renewable energy supplies only a fraction of that energy demand. Therefore, investment in renewables offers a triple dividend: First, investment in renewables, particularly in biomass, would mitigate downward trends in natural capital, thereby increasing inclusive wealth; second, investment in most forms of renewable energy, such as solar energy, will increase produced capital, thereby increasing inclusive wealth; third, greenhouse


# FIGURE 13

The fossil fuel, natural capital, renewable energy, inclusive wealth cycle

gas emissions are reduced, reducing carbonrelated damages and thus increasing inclusive wealth.

# 7. The System of National Accounts: time for change

National income accounts (NIA) are central macroeconomic statistics, showing the level and performance of economic activities in the economy. The System of National Accounts (SNA) of the United Nations attempts to provide a benchmarked framework for measuring and summarizing national income data across all activities within an economy and to facilitate comparing such data across countries (UNITED NATIONS ET AL. 2012). A crucial component of the SNA is the estimation of GDP, where, at given market prices, the gross value of all the goods and services produced within an economy is estimated.

The basic reason countries construct standard national accounts is to use them as

macroeconomic planning tools. With the data a country obtains from implementing the standard SNA, it can quantitatively answer questions on specific policy planning issues. For instance, if a country is looking to bring about a 5 percent increase in transportation sector output over the next five years, how would production in other sectors be reorganized to meet this demand?

A country is able to use the data in a general equilibrium model, computing the Leontief input-output coefficients to derive knowledge on optimal resource allocations (how much investment, labor, etc. resources must be pumped into each interdependent sectors?) to meet this planning target. The focus is on production.

Economists are now increasingly in agreement that instead of measuring growth in GDP or income – which are flow accounts – a more meaningful and correct approach would be to comprehensively measure growth in wealth: not just income wealth, but rather the entire stock of wealth that includes the value of natural and human capital, in addition to produced capital (DASGUPTA AND MÄLER 2000, ARROW ET AL. 2004, DASGUPTA 2009).

In measuring GDP, various contributions of ecosystem services, such as bioremediation by wetlands, storm and flood protection by mangroves, or the prevention of soil erosion by forests, are ignored. This is due both to the relatively low perceived value of ecosystem services, as well as the fact that these resources and their contributions usually fall outside the domain of the market and production boundary of the SNA, and hence remain un-priced.

Similarly, investments in education are absent from the SNA. Education is only reflected in accounts as an expenditure flow, while no references are made to the notion of human capital in national accounts (BOS 2011). There has been some discussion on developing satellite accounts for monitoring and keeping track of the changes in human capital, but these remain at an early stage.

Considering the enormous contribution human capital – in particular education – plays in determining the size of an economy's productive base, as we see in this report, this is a serious oversight of national accounting.

A country investing in constructing true inclusive wealth accounts will seek to gather and interpret information on the following:

- natural resource asset and stock accounts: the physical and economic value of stocks of natural resources, such as forests, water, fisheries, agriculture land, range land, etc.;
- flow accounts from management and use of the resource stocks: material flows, environmental services, and pollution by sector linked to input-output tables;
- environmental protection and resource management expenditure accounts: resource user fees, subsidies, expenditures by government to manage resources;
- information on how human capital or education, in this case, is utilized in the economy; and
- private and social rates of return and the relative prices of different types of

labor, based on education levels and demographics

This is information that can feed into the macroeconomic policy planning process, just as the SNA feeds into policy processes, while add-ing crucial input on sustainability:

A country that generates inclusive wealth accounting data as outlined above will be able to quantitatively answer critical questions, such as:

- How much of the stock of a particular asset, required to produce a sustainable flow of consumption, has changed over the previous x years?
- 2. Can the present stock of assets, i.e., the productive base of an economy, be sufficient to maintaining the present level of consumption in the future? For how long?
- 3. Where should investment be directed to ensure that change in the overall asset base of an economy remains positive?
- 4. How much of the productive base of an economy is lost due to climate change? How much is needed to invest in reducing and/or adapting to climate change?

Countries have long kept a close watch on national income accounts to evaluate economic performance and assess the effectiveness of development policies and plans. Yet, conventional indicators based on national income accounts, such as GDP, can be misleading because they say nothing about whether growth is sustainable.

Countries can grow in the short term by running down their assets, including natural capital assets such as forests, minerals, and water. But such growth cannot be maintained over the long term. Wealth accounting serves as an indicator for the sustainability of growth.

A key element of wealth accounting systems is the conversion of physical stocks of different attributes to a single common denominator – generally monetary – through prices.

In many cases, market prices are sufficient. However, there are various instances in which market prices for particular stocks either do not exist, or do not reflect the true value of the assets. In such cases, it is imperative to determine the shadow price (or social price) of these assets using different economic and noneconomic techniques. This is an area for which there is still a lack of a concerted action by governments to acquire these prices. In most cases, isolated valuation case studies are used to determine shadow prices.

Countries should undertake scoping studies to evaluate methodologies and strategies for determining the social prices of these nonmarket assets, and put in place systematic efforts to collect the information required to determine them. This is of paramount importance for accurately valuating natural and human capital assets.

8. Lessons and conclusions

The key information the IWR offers policymakers is an overview of changes in the productive base of a country. It provides insights into trends within the individual capital asset groups, particularly human and natural capital – central pillars of inclusive wealth that remain underserved by current statistical collection, analysis, and economic policy-making. For natural capital in particular, the IWR gives insight into how subcomponents have changed, for instance changes in land use and cover, or fossil fuel reserves.

The Inclusive Wealth Index (W) is, after two years, still in an early – essentially experimental – stage. However, it fills a crucial gap among the indicators of well-being and sustainability available today, giving information on the state of the productive base of an economy, as well as insights into relationships and trade-offs among the capital components of the productive base.

The IWR by itself provides information only on changes in asset bases. However, its utility vastly increases when used in conjunction with other information about an economy. For instance, we showed that changes in land use gleaned from the IWR, together with information on crop yields, can be utilized to inform investment decisions on how to direct investments to increase productivity. This type of composite analysis is possible for all the forms of capital comprising inclusive wealth.

Another potential area of use for the IWR is in education. The wealth of information in the report as well as the potential areas for research it identifies makes it a unique tool within the education spectrum. It offers a broader understanding of economics and its contribution to human well-being. Moreover, the multidisciplinary nature of the exercise makes it fertile ground for multidisciplinary collaboration bringing together economists, ecologists, demographers, health specialists, and educationalists, among others.

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# Part II: Human capital

# **CHAPTER 3**

# Human capital measurement: a bird's eye view

Gang Liu and Barbara Fraumeni

# KEY MESSAGES

**Measuring human capital can serve** many purposes: it can help one better understand what drives economic growth; assess the long-term sustainability of a country's development path; measure the output and productivity of the educational sector; and facilitate informed discussions on social progress and well-being. In spite of this, human capital has not yet been included within the asset boundary of the System of National Accounts (SNA).

The multifaceted nature of the concept of human capital creates substantial challenges for its measurement. By focusing on formal education and economic returns for individuals – rather than on human capital in general and all the benefits due to human capital investment – we can begin from an empirically manageable and practical point of departure.

All existing approaches to measuring human capital have both advantages and disadvantages. However, the monetary measures generated from the cost-based and the income-based approaches should arguably be designated a "core" status. One reason for this is to enable direct comparison of figures with those for traditional produced capital covered by the SNA, the construction of which is a primary task of national statistical offices. Drawing on country experiences and international initiatives in the field of human capital measurement, one may conclude that an international trend is emerging toward an incomebased approach, specifically the lifetime income approach. Estimates based on this approach can be used to assess the relative contribution of a range of factors (demographic, education, and labor market) to the evolution of human capital, and facilitate corresponding policy interventions.

Despite significant progress having been made, there remain considerable challenges regarding data availability and detailed methodological choices inherent in applying monetary measures. Further research should therefore be encouraged, including toward the compilation of quality data for use in international and inter-temporal comparisons; the construction of experimental satellite accounts, in order to better understand and reconcile the discrepancies between estimates based on the cost-based and the income-based approaches; and, eventually, toward incorporating human capital measures into the SNA in the future.

# 1. Introduction<sup>1</sup>

Human capital is, according to a number of empirical studies, by far the single most important component of national wealth (e.g., GREAKER ET AL. 2005, GU AND WONG 2008, WORLD BANK 2006, 2011). A robust measure of human capital, together with those for other assets, can therefore be used to gauge how well a country is managing its national wealth, and assess the long-term sustainability of its development path (UNECE 2009, ARROW ET AL. 2012).

Measuring human capital can also serve many other purposes, for instance, to facilitate a better understanding of the driving forces behind economic growth (e.g., LUCAS 1988, ROMER 1990A, BARRO AND SALA-I-MARTIN 1995, ARNOLD ET AL. 2007), or to measure the output and productivity performance of the educational sector (e.g., JORGENSON AND FRAUMENI 1992A, SIANESI AND VAN REENEN 2003, SCHREYER 2010, GU AND WONG 2010A). In addition, measuring human capital is crucial for the discussions on "beyond GDP" that have recently gained a resurgence among policy-makers, since the distribution of human capital across households and individuals, as well as the non-economic benefits due to human capital investment, are among the crucial determinants for "quality of life" and well-being (e.g., STIGLITZ ET AL. 2009, OECD 2011, and various EU initiatives<sup>2</sup>).

Despite this, human capital has not yet been incorporated into the asset boundary of the System of National Accounts (SNA). Empirical human capital models are usually based on various proxies of human capital measures. The diversity of approaches to measuring human capital makes it difficult to draw policy implications by comparing these estimates within and across countries.

The purpose of this chapter is to discuss the multifaceted nature of the concept of human capital, to review the various measuring approaches currently available, and to draw on the experiences from national studies and international activities in the field of human capital measurement in order to identify the most promising approach(es) for finally incorporating human capital measures into the SNA. Overall, Chapter 3 provides a framework for understanding general discussions presented regarding the conceptual and methodological issues related to human capital measurement; while Chapter 4 focuses on the implementation of several specific methodologies, with corresponding empirical results compared and discussed in detail.

In the *Inclusive Wealth Report 2012* human capital was considered to have the following four components: education, skills, tacit knowledge, and health (UNU-IHDP AND UNEP 2012). Compared with the first three, the health component is apparently of a distinct nature, although it is an essential part of human capital in a broad sense,<sup>3</sup> and without it, none of the other components can be utilized in a meaningful and proper way.

Within the section including empirical estimation of the *Inclusive Wealth Report 2012*, human capital was *de facto* referred to as the contribution of education to the productive base only, and the term health capital was deliberately reserved for the study of the progress

<sup>1</sup> This chapter builds on the work that was carried out when one of the authors (Gang Liu) led the human capital project at the OECD. The authors are very grateful for the valuable comments by three reviewers (Dale W. Jorgenson, Katharine G. Abraham, and Hui Wei), and for the important suggestions by all the participants in an Inclusive Wealth Report (IWR) Workshop held in Kuala Lumpur in May 2014. However, any remaining errors and omissions are the sole responsibility of the authors.

<sup>2</sup> For more information on these initiatives and a later European Commission Communication on "GDP and Beyond – Measuring progress in a changing world", please visit the following links: http://www.beyond-gdp.eu/ and http://www.eubusiness.com/topics/finance/beyond-gdp.

<sup>3</sup> For example, Gary S. Becker viewed education, on-the-job training, and *health* as components of human capital, all having consequences for earnings and economic productivity (Becker 1993).

made in the life expectancy of the population. For the 2014 report, we shall follow the same tradition and refer to human capital as comprising exclusively education, skills, and tacit knowledge.<sup>4</sup> In particular, unless otherwise stated, human capital from education remains the focus of Chapter 3 and Chapter 4, while health capital is separately dealt with in Chapter 5.

The rest of the chapter is structured as follows:

Section 2 explains why human capital has not yet been included within the asset boundary of the current SNA and illustrates the conventional practices dealing with the output of the educational sector in the SNA. Advances in researches attempting to incorporate human capital into the SNA are also briefly introduced in this section.

Section 3 provides comprehensive discussions about the concept and definitions of human capital, as well as the resulting implications for its measurement. The strengths and weaknesses of different measuring approaches currently available are then discussed. In particular, the purpose of Section 3 is to identify those promising approach(es) most likely to be employed to make human capital estimates for incorporation into the SNA in the future.

In Section 4, an overview on country experiences and international initiatives is given, based on an international questionnaire collected by the UNECE and the OECD, as well as on a literature survey of the leading activities in the field of human capital measurement.

Main issues and challenges remained are described in Section 5.

Section 6 concludes the chapter and includes recommendations for suggested future research.

# 2. Human capital and the SNA

The System of National Accounts (SNA) is an internationally agreed-upon standard statistical framework that provides a comprehensive, consistent, and flexible set of macroeconomic accounts suitable for measuring, monitoring, and analyzing the economy and its constituents, so as to assist national policy planning processes.

Human input is the primary input in most production processes. The value of that input is not only raw labor, but is to a large extent dependent on the knowledge, skills, and competencies human beings bring to production processes. Nowadays it is generally recognized that an educated population is vital to economic well-being in most countries. It is often proposed that expenditures on education and staff training should be classified as gross fixed capital formation - as a form of investment in human capital. Clearly the acquisition of knowledge, skills, and qualifications increases the productive potential of the individuals in concern and is a source of future economic benefit to them.

There exist repeated efforts to bring the concept of human capital, including its formation and utilization, into the framework of the SNA. There are essentially two arguments against these requests: One is attributed to the "production boundary" and the other to the "asset boundary", as stipulated by the SNA (e.g., SNA 2008).

First, human capital is usually acquired by learning, studying, and practicing. But these activities cannot be undertaken by anyone else on behalf of the person considered, and thus do not satisfy the "third party criterion" that delineates the production boundary of the SNA. As a consequence, the acquisition of knowledge and skills is not considered as a process of production, even if the provision of the services by educational institutions (schools, colleges, universities, etc.) is.

Second, human capital cannot be detached from the person in whom it is embodied, nor can it be transacted separately and in its own

<sup>4</sup> Human capital defined as such is closer to the OECD definition (see OECD 2001), which has gradually received wide acceptance. More on the OECD definition will be discussed in Section 3 of this chapter.

right in the market like conventional fixed capital. Simply because it is practically difficult, if not impossible, to envisage an "ownership right" in connection with people, human capital is not treated by the SNA as an asset.

Within the framework of the current SNA, education services produced by educational institutions are treated as being consumed by students in the process of their acquiring knowledge and skills. In other words, this type of education is treated as final consumption, rather than as input used in the production of human capital by the students. When training is given by an employer to enhance the effectiveness of staff, those costs are treated as intermediate consumption. In both cases, the learning and practicing processes of the individuals in question are entirely ignored.

Sometimes the entire educational sector (instead of the students themselves) is regarded as the supplier of human capital, and the product of the educational sector treated as human capital investment. Because most educational expenditures are often financed by public funds that are allocated by the government and/or non-profit institutions, the output of the educational sector is currently measured merely based on the costs of the market inputs into this sector, i.e., teachers' wages and salaries, the consumption of fixed capital (e.g., school buildings), household expenditures for school fees and educational materials, etc.

This input-based approach to measuring the output of the educational sector (which is human capital investment) is, however, inadequate for productivity analysis, since it ignores changes in the efficiency with which various inputs are used in production. To support a proper analysis of the productivity performance of the educational sector, independent output-based measures of its economic production are therefore required, which subsequently necessitates a good measure of human capital due to education.

Despite the difficulties of compiling a complete system of accounts for investment in human capital, wealth in the form of human capital, and the services of human capital that are integrated with the SNA, research efforts along this line are ongoing. For instance, Jorgenson and Fraumeni (1989) present a complete system of national accounts that include investment in human capital. The flow of human capital services, cross-classified by age, gender, and educational attainment, provides a price and volume of market labor compensation that has been integrated with the national accounts in ten countries following the EU KLEMS project (O'MAHONY AND TIMMER 2009).

Several attempts to measure the output of the educational sector beyond the conventional input-based approach have also been carried out. For example, by considering the output of the educational sector to be investment in human capital through formal education, Jorgenson and Fraumeni (1992a) and Gu and Wong (2010a) measure the output of this sector in terms of increments to lifetime incomes of individuals taking this education.

# 3. Concept and measuring approaches

#### 3.1 Concept and definition of human capital

The origin of the human capital concept can be traced back to the work of Adam Smith in the 18<sup>th</sup> Century. Smith underlined the importance of "the acquired and useful abilities of all the inhabitants or members of the society"; while an individual will incur costs to obtain such abilities, once acquired, they stand as "a capital fixed and realised, as it were, in his person" (SMITH 1776).

The practical implications of treating an individual's abilities as a kind of capital were not widely recognized until the 1960s, when economists began to incorporate the notion into their works. This shift partly reflected the view that the concept of human capital could be employed to explain the large difference between the increase of the economic output of a country and that of the traditional trioinputs (land, labor, and produced capital). Some



#### **FIGURE 1**

Human capital: a sketch of its formation, composition, and benefits generated

Source: The OECD human capital project (see Boarini et al. 2012).

economists contended that investment in human capital was probably the major explanation for this difference (e.g., SCHULTZ 1961).

There are many definitions of human capital in the literature, but most of them stress the economic returns due to human capital investment. Schultz (1961), for example, defined human capital as "acquired skills and knowledge" to distinguish raw (unskilled) labor from skilled labor; likewise, the Penguin Dictionary of Economics (1984) defined human capital as "the skills, capacities and abilities possessed by an individual which permit him to earn income", a definition which emphasizes the improvement of people's economic situation due to human capital investment. The World Bank (2006) similarly defined human capital as the productive capacity embodied in individuals, with special focus on its contribution to economic production.

As today's economies become more knowledge-based and globalized, the economic importance of human capital to both an individual's competitive advantage and to a country's economic success has become more significant than ever. But human capital investment delivers many other non-economic benefits as well, such as improved health, enhanced personal well-being, and greater social cohesion. These broader benefits are viewed by many as being as important as, if not more important than, the economic benefits in the form of higher earnings and economic growth.

Acknowledging these broader benefits, the definition of human capital has been gradually extended. For instance, in an OECD report published in 1998, human capital was defined as "the knowledge, skills, competences and other attributes embodied in individuals *that are relevant to economic activity*" (OECD 1998). A later report, however, updated the definition of human capital as "the knowledge, skills, competencies and attributes embodied in individuals *that facilitate the creation of personal, social and economic well-being*" (OECD 2001).

The OECD definition of human capital according to the 2001 report has gradually obtained wide acceptance. Figure 1 displays a stylized drawing of the elements pertaining to this broad definition, comprising the essential constituents that make up human capital, as well as the numerous channels through which human capital is developed, and the diverse benefits that human capital delivers.

# 3.1.1 What is included in human capital?

The overarching OECD definition incorporates various skills and competencies that are acquired through learning and experience, but may also include innate abilities. Some aspects of motivation and behavior, as well as the physical, emotional, and mental health of individuals are also regarded as human capital within this broad definition (OECD 2001).

The components of human capital reflect its multifaceted nature. For instance, they include both general and work-specific skills, as well as tacit and explicit ones. They cover not only the cognitive skills that were conventionally recognized but also non-cognitive skills, such as intra- and inter-personal skills, which have assumed an increasingly important role in modern societies.

Distinct from traditional fixed capital, almost all types of knowledge, skills, competencies,

and attributes are invisible. While fixed capital wears out through use, human capital typically grows through use and experience, at least at first. However, both human and fixed capitals accumulate through investments and decline due to obsolescence, although not exactly in the same manner. For example, human capital depreciates due to lack of use, obsolescence of knowledge, and ageing.

### 3.1.2 How is human capital acquired?

The OECD definition also implies that human capital can be accumulated through many channels that may be characterized as both *lifelong*, in terms of learning from birth to death, and *life-wide*, in terms of learning at various occasions, including families (through parenting), schools (through formal and informal education), workplaces (through on-the-job training and work practice), and even daily life (through informal learning, anywhere and anytime).

Human capital development is normally involved with private and public resources, as well as market and non-market inputs. A distinct feature of human capital investment is that almost all types of such investment require learning by individuals themselves, an activity that is not only non-market, but also outside of the production boundary of the current SNA, because it fails to meet the "third party criterion" as mentioned above in Section 2.

How human capital is developed also hinges upon a range of environmental factors, such as cultural backgrounds, social relationships, as well as political, legal, and institutional arrangements. For instance, investment in skills takes place in many different stages of the lifecycle of individuals, during which social capital (i.e., networks and norms) plays a critical role in fostering a culture of learning within a society (COLEMAN 1990). There is considerable agreement drawn from the relevant research that the family, as well as social- and home-backgrounds, will shape school outcomes, although the relative importance of the various factors is not always clearly understood. Another issue pertinent to human capital measurement that raises practical difficulties is how to distinguish, within education-related expenditures, between purposes for consumption and investment. While the distinction between the two elements is conceptually clear, in practice it is not easy to verify which of the two perspectives is more relevant in any situation. In reality, most of the activities contributing to human capital accumulation are likely to include both consumption and investment elements, as in the case of household expenses for buying school uniforms for students.<sup>5</sup>

Things become even more complicated once the health aspect is taken into account. Health care is regarded as one type of human capital investments in Figure 1, because health condition is one of the vital attributes that are encompassed by the broad notion of human capital. Moreover, a better health status normally enhances an individual's learning abilities and also job market performance, so that his/ her embodied human capital can be utilized in a more efficient way. However, it is not always clear-cut whether activities related to health care are pursued for the purpose of investment rather than consumption. For instance, doing exercises may qualify as an investment, but expenses for buying tonic foods and beverages could serve both purposes.

As shown in Figure I, human capital investment in any given country may also take the form of migration, with the immigration of skilled people representing an addition to the stock of human capital for the country of destination, and a depletion of human capital for the country of origin.

# *3.1.3* What benefits stem from human capital investment?

As reflected by the broad OECD definition and as shown in Figure 1 as well, human capital investment can generate both economic and non-economic benefits. Economic benefits include enhanced employability and, if employed, improved earnings and career prospects for individuals undertaking the investment; non-economic benefits can take the form of productivity increases in performing non-market activities (e.g., household production) or of personal benefits that are not related to production (e.g., greater enjoyment of arts and culture, improved health, and subjective well-being).

The benefits can also spill over to other agents and even to society at large. For instance, at the firm level, higher productivity of some employees, due to higher education, may increase the performance of other workers and, hence, company profitability. At macro-economic level, recent evidence has highlighted the positive impact of human capital on economic growth. Further, these spill-overs are not limited to economic returns. For instance, education may make people better citizens and better parents, leading to greater social cohesion.

Finally, as illustrated by the dotted arrow in Figure 1, there also exist feedback effects, running from the benefits generated by human capital investment to the investment itself. For example, workers with higher education are more likely to benefit from, and thus be willing to pursue, further education and training. In addition, the feedback process may lead to a virtuous cycle wherein more education makes further learning even easier and faster, and thus more efficient. At the national level, there is a long-standing debate on the direction of causality between education and economic growth. Recent studies have demonstrated that the causality may operate in both directions, suggesting that a feedback loop may also operate at the macro level.

<sup>5</sup> Due in part to this reason, there appears an additional difficulty if human capital is measured solely from the input side, i.e., to sum all expenditures made to produce the capital goods. More discussions on this methodological issue are in subsection 3.2 of this chapter.

In summary, there are so many elements involved in and intertwined with the concept of human capital that getting a full picture of every single element covered by the broad OECD definition of human capital, of the causal links between each type of human capital investment and the corresponding benefits, and of the feedback loops among them, is quite complicated. This implies that encompassing all the elements of Figure I into a single measure of human capital is a daunting task, which could not be realistically accomplished in the foreseeable future. The most sensible approach is to address this task following a stepwise approach.

A sensible way is to focus on a narrower range of elements, starting from those aspects characterized as either involving lower conceptual challenges or enjoying greater data availability. For instance, most of the human capital definitions currently employed distinguish human beings themselves from the acquired abilities, and include only the latter within the domain.

Closely related to this choice, health status is often considered separately as another specific asset, i.e., health capital (e.g., ABRAHAM AND MACKIE 2005) because of its apparently distinct feature from the education aspect of human capital, and also due to practical difficulties, such as how to distinguish health-related expenditures between purposes for consumption and investment. However, treating health as a separate type of capital does not imply that health status is irrelevant for the measurement of the "educational" capital explored here. What it implies is that the measure of the human capital stock described here will only reflect the impact of health care activities in improving people's economic returns, which is considered as one of the two outputs of the health sector (ABRAHAM AND MACKIE 2005).<sup>6</sup>

In practice, one frequently-chosen option by many researchers and institutions is to focus on formal education as the main form of human capital investment; and on the accrued economic returns for individuals as the main benefits due to human capital investment,<sup>7</sup> even if the broad OECD definition is accepted as a useful reference point.<sup>8</sup>

#### 3.2 Measuring approaches

Currently there exist several approaches to measuring human capital. In Figure 2, a broad distinction is made between an indicatorsbased approach and monetary measures. The indicators-based approach relies on physical indicators, which might be further divided into quantitative indicators (e.g., educational attainment, average years of schooling) and qualitative ones (e.g., class size, test scores).

The monetary measures of human capital include estimates based on an indirect or residual approach (e.g., WORLD BANK 2006, 2011),<sup>9</sup> as well as direct estimates that utilize information on its various components. The two main types of direct measures are the cost-based approach (e.g., KENDRICK 1976) and the income-based approach (e.g., JORGENSON AND FRAUMENI 1989, 1992a, 1992b).<sup>10</sup>

flow of health consumption over the course of a person's remaining life (Abraham and Mackie 2005).

7 However, it does not imply that there are no other possible ways to move beyond this option.

8 More on this is in Section 4.

9 Here, the distinction of "direct" versus "indirect" has only relative meaning. Unlike physical capital, human capital is invisible. Therefore, in a strict sense, all methods trying to measure human capital can only be "indirect".

10 The above typology is not the only way to classify the various approaches. For instance, a distinction can also be made between parametric and non-parametric approaches to measuring human capital. The former involves econometric techniques (e.g., Kyriacou 1991, Mulligan and Sala-i-Martin 1995, Barro and Lee 2010), while the latter usually does not.

<sup>6</sup> The other output of the health sector is the value of health capital that can be defined as the expected



Different measuring approaches have both advantages and disadvantages, some of which are specific to one approach, while others are common across different approaches. This section reviews some of the most important advantages and disadvantages, with the view to find one or several approaches that are more capable of making human capital estimates that are in accordance with the SNA.

The *indicators-based approach* measures human capital through various educational characteristics of the population. Indicators often used as single proxies for human capital in the academia include adult literacy rates (e.g., AZARIADIS AND DRAZEN 1990, ROMER 1990B), school enrollment ratios (e.g., BARRO 1991, MANKIW ET AL. 1992, LEVINE AND RENELT 1992), average years of schooling, and other measures (e.g., BENHABIB AND SPIEGEL 1994, BARRO AND SALA-1-MARTIN 1995, GUNDLACH 1995, ISLAM 1995, O'NEILL 1995, TEMPLE 1999, BARRO 1997, 2001, KRUEGER AND LINDAHL 2001).

However, the use of a single physical indicator as a proxy for human capital, though appealing for its simplicity, cannot on its own adequately measure the various dimensions of skills and competences (OECD 2001), and sometimes poorly specifies the relationship between education and the stock of human capital (WÖSSMANN 2003, KOKKINEN 2010). Therefore, only a wider definition can provide useful clues about where investment is most needed and where benefits go.

Sometimes, dashboard-type indicators (e.g., EDUCATION AT A GLANCE, EDERER ET AL. 2007, 2011) are applied, but they rely on a number of statistics that, though rich in information, lack a common metric and, as a result, cannot be easily aggregated into an overall measure. This makes them less suitable for comprehensive comparisons of the stock of human capital across countries and over time.<sup>11</sup> Also, a set of indicators does not allow easy comparison of the relative importance of different types of capital, i.e., produced, natural, and human capitals (STROOMBERGEN ET AL. 2002).

Recently one type of indicators has attracted growing attention in the world, i.e., pencil and paper test scores of people's competencies. Examples are the OECD Programme

<sup>11</sup> For instance, under certain circumstances, information drawn from different statistics may conflict with each other.

for International Student Assessment (PISA), which tests 15 to 16 year olds students for cognitive skills in terms of reading, mathematics, science, and problem solving; and the OECD Programme for International Assessment of Adult Competencies (PIAAC), which tests adults for their competencies in terms of literacy, numeracy, and ability to solve problems in technology-rich environments.<sup>12</sup>

These programs provide vitally important information for policy-making, in particular for evaluating educational programs. However, like all surveys, they are subject to survey and test limitations (e.g., with respect to sample size, range of variables included, country coverage, etc.). More importantly, as these programs are resource-intensive in terms of both money and time required to implement, administer, process, analyze, and report, they are typically undertaken with low frequency. Finally, the information generated from these programs is not easily integrated into human capital accounts.<sup>13</sup>

Within the group of monetary measures, the *indirect (residual) approach* is applied by the World Bank through its comprehensive wealth accounting. This approach measures the stock of human capital as the difference between the total discounted value of each country's future consumption flows (taken as a proxy for

total wealth) and the sum of the tangible components of that wealth, i.e., produced capital and the market-component of natural capital (WORLD BANK 2006, 2011, RUTA AND HAMILTON 2007, FERREIRA AND HAMILTON 2010).<sup>14</sup> A similar approach has also been applied by Statistics Norway at the country level (GREAKER ET AL. 2005).

While this indirect approach can be applied to a large number of countries based on lessdemanding statistics,<sup>15</sup> it has certain limits. First, by taking as its starting point the discounted value of future consumption flows, it apparently ignores the non-market benefits of various capital stocks. Second, this measure is affected by measurement errors in all the terms entering the accounting identities, resulting in potential biases in the final estimates of human capital. Third, the approach cannot explain what drives the observed changes of the human capital over time, thus offering less valuable information for policy interventions.

Among direct measuring approaches, the *cost-based approach* measures human capital by looking at the stream of past investments undertaken by individuals, households, employers, and governments (e.g., SCHULTZ 1961, KENDRICK 1976, EISNER 1985). This approach relies on information on all the costs that are incurred when producing human capital. These costs include monetary outlays by each of the agents mentioned above, but can also incorporate non-market inputs (e.g., the imputed value of the time devoted to education by students, their parents, and volunteers).

The cost-based approach is relatively easy to apply, at least when limited to market inputs, because of the ready availability of data on both public and private expenditures in formal education. The approach can also be extended

<sup>12</sup> For more information on the PISA and PIAAC, please visit the following websites: http://www.pisa.oecd.org/pages/0,2987,en\_32252351\_32235731\_1\_1\_1\_1\_1\_00. html; http://www.oecd.org/document/35/0,3746,en\_2649\_201185\_40277475\_1\_1\_1\_00.html.

<sup>13</sup> The last point is also relevant when considering differences between parametric and non-parametric approaches to measuring human capital. Parametric approaches are frequently used in academic researches; however, since they rely on econometric techniques, different assumptions and model specifications, even if based on the same dataset, will typically lead to different estimates. On the contrary, non-parametric approaches avoid these problems and are more akin to the tools typically used by national statistical offices and other producers of human capital statistics.

<sup>14</sup> This difference is labelled by the World Bank as "intangible assets", of which human capital is found to be the most important component (World Bank 2006, 2011).

<sup>15</sup> The World Bank work of comprehensive wealth accounting covers more than 100 countries over the decade from 1995 to 2005 (World Bank 2011).

to account for expenditures undertaken for on-the-job training. However, the approach has been criticized on conceptual grounds, with some arguing the value of human capital should be regarded as determined by demand and supply rather than solely by production costs (LE ET AL. 2003)<sup>16</sup>.

In addition, following this approach, it is hard, if not impossible, to distinguish between expenditures for investment and consumption. Consequently, estimates by this approach usually rely on arbitrarily allocating spending between these two categories. For instance, during the period of taking education, part of household expenditures is used for paying students' food and clothes, which could serve both consumption and investment purposes. Challenges are also involved with the choice of the price index used to deflate historical expenditures related to human capital investment in order to construct a stock value based on the perpetual inventory method.

Moreover, the depreciation rate, which matters a great deal when constructing the stock of human capital based on the perpetual inventory method, is usually set arbitrarily in practice, although this treatment is not uncommon during the estimation of the stock of traditional fixed capital based on the same method. Overall, the cost-based approach ignores a fundamental feature of the process of education, i.e., the lengthy gestation period between current outlays for educational inputs and the resulting generation of human capital embodied by more competent people (JORGENSON AND FRAUMENI 1989, 1992A).

The *income-based approach* measures human capital by looking at the stream of future earnings that human capital investment generates. In contrast to the cost-based approach, which focuses on the input side, the income-based approach measures the stock of human capital by looking at the output side.<sup>17</sup> By focusing on the earning power of each person (usually proxied by wage), the income-based approach values human capital at market prices, under the assumption that these prices are good signals of the value of human capital services that result from the interaction of demand and supply in the labor market.

This approach has been used at least since the 1960s (e.g., WEISBROD 1961, GRAHAM AND WEBB 1979). However, it was the seminal works of Jorgenson and Fraumeni (1989, 1992a, 1992b) that spawned interests in measuring human capital by applying the lifetime income approach (also called the Jorgenson-Fraumeni approach). The lifetime income approach applies the neoclassical theory of investment (JORGENSON 1967) to human capital. According to this theory, the price of capital goods depends upon the discounted value of all future capital services derived from the investments. On a per capita basis, this means that the value of an individual's human capital can be determined from that person's discounted lifetime income.

The lifetime income approach has other advantages. In particular, the extension of this approach naturally leads to an accounting system that includes values, volumes, and prices as basic elements. This opens the way for the construction of a sequence of accounts similar to those used for produced capital within the framework of the SNA (FRAUMENI 2009).

However, the income-based approach is not immune from drawbacks. For instance, it has been found that the relative value of human capital of a country is positively correlated with that of its produced capital, i.e., an increase in the latter tends to enhance the marginal product of labor and hence the value of human capital (e.g., HAMILTON AND LIU 2014). Therefore, due caution should be taken when comparing

<sup>16</sup> Note that the production costs of human capital include forgone earnings that are usually determined by demand and supply in the labor market.

<sup>17</sup> While the outputs from human capital investment are of many types (i.e., monetary and non-monetary, private and public), the output measured by the income-based approach is commonly limited to the private monetary benefits that accrue to the person investing in human capital.

human capital across countries. In addition, in order to calculate expected future earnings, some subjective judgements must be made about the discount rate, real income growth rate, etc. Most crucially, there are reasons to believe that the labor market does not always function perfectly. As a result, the wage rate typically used as a proxy for earning power in the income-based approach is not always equal to the marginal value of a particular type of human capital. Moreover, differences in wages will not truly reflect differences in earning power in some cases where trade unions may command a premium wage for their members, or where real wages may fall in economic recessions.

To sum up, distinct from the indicatorsbased approach, the monetary measures, in particular, the cost-based and the income-based approaches, combine many different aspects that contribute to human capital in a single metric: money. In other terms, they reflect the different factors that contribute to human capital accumulation. For example, estimates based on the income-based approach allow comparisons between the relative importance of demographic (e.g., the age and gender structure of the population), educational (e.g., the number of people with different levels of educational attainment, enrollment rates), and labor market factors (e.g., the employment probabilities and earning differentials by educational characteristics). Similarly, human capital estimates based on the cost-based approach allow comparisons between the relative importance of the expenditures incurred by different sectors (e.g., public administration, households, and firms) and of market versus non-market inputs (e.g., time devoted to educational-related activities by students, parents, support staff, etc.).

It seems that all approaches to measuring human capital have strengths and weaknesses. Given the importance of the SNA in official statistics and economic analysis, monetary measures – in particular the cost-based and the income-based approaches – are most likely to be used to construct human capital measures based on an explicit accounting framework within the SNA.

Arguably, to address issues related to growth accounting, monitoring sustainability, and measuring the output and productivity performance of the educational sector, monetary measures of human capital will have a key role to play. Even for the purpose of assessing the impact of human capital on "quality of life" and well-being, monetary measures of human capital are also important.

# 4. Overview of national studies and international initiatives

# 4.1 Country experiences

This subsection provides an overview of national studies conducted either as part of the research activities of national statistical offices, or by independent researchers, with focus being placed on the purpose, concept, methodology, and data sources used for measuring human capital in different countries. The overview begins with a summary of the results of a recent questionnaire on national practices in measuring human capital that was responded by UNECE CES (Conference of European Statisticians) countries.<sup>18</sup> Given the popularity of the monetary measures in national studies, this subsection also presents some findings from representative country studies based on the cost-based and the income-based approaches.<sup>19</sup>

<sup>18</sup> The UNECE CES questionnaire was designed by the OECD, and collected by both the OECD and the UNECE, with the purpose of providing an overview of what countries have done, are doing, and are planning to do in the field of human capital measurement. The questionnaire and answers to it are summarized in Boarini et al. (2012).

<sup>19</sup> The indirect/residual approach will not be discussed here, because, at the national level, there are just a handful of countries applying this approach (e.g., Norway, the Netherlands).

# TABLE 1

Key findings about country practices on measuring human capital

| Questions  | Findings  |
|--|---|
| 1. Purpose of measurement                            | Most countries have multiple purposes, including for education-related policies, growth accounting/productivity analysis, national wealth account-<br>ing, satellite account construction, sustainability assessment, and measur-<br>ing well-being and social progress.  |
| 2. Concept of human capital                          | Although some countries refer to the broad OECD definition of human<br>capital, many countries prefer pragmatically to use definitions that have<br>narrower scope, tending to focus on the economic/productive dimension.  |
| 3. Data sources and availability                     | Data sources used are diverse, including survey data, administrative data, census data, and others. Almost all data needed for making measures of human capital are available within the statistical system of each country.  |
| 4. Status and frequency of human capital estimates   | Measuring human capital is carried out usually by either independent<br>researchers or statisticians working with national statistical offices.<br>Although most of the existing estimates are in the form of research results<br>instead of official statistics, many countries measure human capital on a<br>regular basis, most of them annually.  |
| 5. Satellite accounts                                | Only a few countries plan to construct satellite accounts for human<br>capital in general and for the education sector in particular. Likewise, very<br>few countries have assessed the possibility and potential implications of<br>incorporating measures of human capital into the SNA.  |
| 6. Measuring approaches                              | Both physical indicators and monetary measures are often applied. Most<br>countries choose only one approach, and the majority rely on monetary<br>measures only, while a few other countries choose multiple approaches.   |
| 6.1 Indicators-based approach                        | For many countries, conventional quantitative indicators are still used and<br>draw directly from education statistics, such as population distribution by<br>education category, average years of schooling, etc. Only a few countries<br>collect qualitative indicators such as those undertaken as part of the<br>OECD PISA and PIAAC programs.  |
| 6.2 Monetary measures                                | Among monetary measures, the income-based approach is predominant over the cost-based and residual approaches.  |
| 6.2.1 Reasons for applying the residual approach     | The main reason provided for utilizing the residual approach is its simplic-<br>ity, despite conceptual drawbacks.  |
| 6.2.2 Reasons for applying the cost-based approach   | The main reasons for utilizing the cost-based approach are data availabil-<br>ity, applicability in the SNA, and lack of necessity for making assumptions<br>about the future; while the main challenges are related to data availability.<br>Some countries included in their estimates not just the costs incurred by<br>educational institutions, but also expenditures by firms and private house-<br>holds. However, no countries have ever included non-market costs in their<br>estimates of human capital based on the cost-based approach.   |
| 6.2.3 Reasons for applying the income-based approach | The main reasons for using the income-based approach are its consis-<br>tency with economic theory and with the way in which other assets (such<br>as natural resources) are measured in the SNA. This approach is also con-<br>sidered to be well-established and widely-employed, and to be suitable for<br>constructing a fully-fledged human capital account with volumes, values,<br>and prices as basic elements. Issues related to the methodology and data<br>availability, rather than the concept itself, are regarded as the main chal-<br>lenges for applying this approach. Partly due to data limitations, almost all<br>countries having applied the income-based approach limit their estimates<br>to the working age population and to market activities only. |

# *4.1.1 Results of the UNECE CES questionnaire on measuring human capital*

Overall, out of the 70 CES countries, 46 answered the questionnaire, with 17 providing detailed answers. They are, among OECD countries, Austria, Canada, Finland, France, Germany, Israel, Italy, the Netherlands, New Zealand, Norway, Poland, Slovenia, the United Kingdom, and the United States; and, among CES non-OECD countries, Liechtenstein, Romania, and Ukraine. Highlights from countries' responses are presented in Table I.

# *4.1.2 Representative studies using the cost-based approach*

The cost-based approach to measuring human capital is practically similar to what has been conventionally applied to measuring the stock of fixed capital by the perpetual inventory method, i.e., the stock of human capital is measured as the accumulated value of all the expenditures occurring to its formation, which are considered as human capital investment.

The most well-known application of the cost-based approach is provided by Kendrick (1976) for the United States. Kendrick's estimates are more inclusive than most other applications of this approach, as they include the cost of child rearing, spending on education, and other expenditures considered to have educational value. In addition to these expenditures, Kendrick (1976) also includes the opportunity cost of student time, i.e., earnings forgone by students when studying. Following the same approach, Eisner (1978, 1985, 1988, 1989) estimated the value of the stock of human capital in the United States through a number of modifications to the U.S. national income accounts. Both Eisner and Kendrick included in their estimates of human capital formation the opportunity cost of students' time while in school, as well as the actual costs of education undertaken by both households (e.g., costs for tuition and educational materials) and governments (e.g., costs for salaries and investments

of educational institutions). However, unlike Kendrick, Eisner excluded the costs of childrearing from the investment in human capital.

As discussed in Section 3, applying the cost-based approach requires confronting several methodological challenges. One is how to distinguish between consumption and investment expenditures. Kendrick included in human capital investments all household expenditures related to child rearing to the age of 14, as well as half of household expenditures on health and safety, considering the other half as consumption.

Another challenge related to the implementation of the cost-based approach is choosing the depreciation rates when constructing the stock of human capital. Because of a lack of empirical evidence, Kendrick used for this purpose a modified double declining balance method, while Eisner used straight-line depreciation.

The cost-based approach to measuring human capital was also applied in Germany (EWERHART 2001, 2003), and the Netherlands used the approach to measure firm-specific human capital (ROOIJEN-HORSTEN ET AL. 2007, 2008). Within the framework of the SNA, the cost-based approach was also used by the Statistics Finland to measure human capital in an empirical analysis of the relation between human capital and economic growth (KOKKINEN 2008, 2010). In addition, Statistics Canada plans to apply the cost-based approach, together with the income-based approach that was already employed, in order to compare and reconcile the estimates from the two approaches.

# 4.1.3 Representative studies using the incomebased approach

One of the main conclusions from the CES questionnaire responses is that several countries are currently applying variants of the income-based approach. By bringing together the influence of a broad range of factors (demography, mortality, educational attainment, and labor market aspects), this approach allows comparing the

# TABLE 2

A list of selected national studies applying income-based approaches

| Examples<br>of national<br>studies                | Country           | Motivation   | Time range                | Main data sources  | Population covered  | Market/<br>non-market<br>activities |
|---|-------------------|--|---------------------------|--|---|-------------------------------------|
| Jorgenson and<br>Fraumeni (1989,<br>1992a, 1992b) | United States     | New system of<br>national accounts,<br>output of education<br>sector | 1948-1984,<br>1947-1987   | Rich data based<br>on decades of<br>research                             | Age 0-75  | Both                                |
| Ahlroth et al.<br>(1997)                          | Sweden            | Output of education sector   | 1967, 1973,<br>1980, 1990 | Level of living<br>surveys   | Age 0-75  | Both                                |
| Ervik et al. (2003)                               | Norway            | Output of higher education sector                                    | 1995                      | Register data  | Age 20-64   | Market only                         |
| Wei (2004, 2008)                                  | Australia         | Incorporating human<br>capital into the SNA<br>(stock/flow)          | 1981-2001                 | Census data  | Age 18 (25)–<br>65, labor<br>force/whole<br>population    | Market only                         |
| Le et al. (2006)                                  | New Zealand       | Measuring human<br>capital (stock)                                   | 1981-2001                 | Census data  | Age 18-64   | Market only                         |
| Gundimeda et al.<br>(2006)                        | India             | Accounting for<br>human capital<br>formation                         | 1993-2001                 | Surveys of<br>employment and<br>unemployment,<br>census of<br>population | Age 15-60   | Market only                         |
| Gu and Wong<br>(2008)                             | Canada            | Human capital<br>contribution to<br>national wealth<br>account       | 1970-2007                 | Census/labor force<br>survey   | Age 15-74   | Market only                         |
| Liu and Greaker<br>(2009)                         | Norway            | Measuring human<br>capital (stock)                                   | 2006                      | Register data  | Age 15(16)–<br>67(74), labor<br>force/whole<br>population | Market only                         |
| Christian (2010)                                  | United States     | Measuring human<br>capital (stock/<br>investment)                    | 1994-2006                 | Rich data  | Age 0-80  | Both                                |
| Coremberg<br>(2010)                               | Argentina         | Measuring human<br>capital (stock)/output<br>of education sector     | 1997, 2001,<br>2004       | Household perma-<br>nent survey  | Age 15-65   | Market only                         |
| Li et al. (2010)                                  | China             | Measuring human<br>capital (stock)                                   | 1985–2007                 | Household survey/<br>health and nutrition<br>survey                      | Urban/rural,<br>Age 0-60 (55<br>for female)               | Market only                         |
| Jones and<br>Chiripanhura<br>(2010)               | United<br>Kingdom | Measuring human<br>capital (stock)                                   | 2001-2009                 | Labor force survey   | Age 16-64   | Market only                         |
| lstat (2013)                                      | Italy             | Measuring human<br>capital (stock)                                   | 2008                      | Various surveys  | Age 15–64   | Both                                |

relative importance of these factors and drawing useful policy implications from the estimates.

Table 2 presents a list of national studies that have applied the income-based approach to measuring human capital. This list is meant to highlight the broad range of countries (13) for which these estimates exist, rather than offering an exhaustive list of the full range of studies based on the approach.

As shown in Table 2, data availability varies across national studies. For many countries, the data needed for applying the income-based approach are compiled by the researcher, with many assumptions made during the data construction process. In part due to this and differently from the original studies by Jorgenson and Fraumeni – most of the national studies listed in Table 2 focused on people of working age (typically based on exogenous age thresholds, e.g., 15 and 64) and market activities only.

These limitations reflect a pragmatic way to sidestep a number of conceptual and data issues that arise when applying the full Jorgenson-Fraumeni approach. Incorporating non-market activities into human capital estimates is bound to incur many imputations, and so attracts more controversy. Although it is data limitation that constrains the scope of estimation, focusing on working age population is sometimes considered as being more relevant for measuring a country's productive capacity (WEI 2004, GU AND WONG 2008, GREAKER AND LIU 2008).

Methodological modifications to the original Jorgenson-Fraumeni methodology were also made in some of the national studies. For example, to smooth the business cycle effects that affect the Jorgenson-Fraumeni approach (which relies exclusively on current crosssectional information), Wei (2008) applied a cohort-based estimation method to simulate future earnings. In order to apply the data of educational attainment by quality credentials (rather than by calendar years), some assumptions were also made in order to overcome the difficulties that incurred (e.g., WEI 2004, GU AND WONG 2008, GREAKER AND LIU 2008).<sup>20</sup>

Results from these national studies suggest that the estimated value of the stock of human capital is substantially larger than that of conventional produced capital, even if the measures of the former are restricted to market activities. Measures of the stock of human capital based on the income-based approach tend also to exceed those based on the cost-based approach, a pattern that may reflect the fact that the former approach implicitly attributes the impact of onthe-job training and work experiences to formal education.

When regarding the output of the educational sector as human capital investment, the value of such investment is high compared to the gross fixed capital formation traditionally measured in the SNA. Considering educational expenditures as investment rather than consumption would significantly change our understanding of the extent of capital formation in any given year.

Estimates of the value of the human capital stock based on the lifetime income approach are sensitive to choices on key parameters employed in this approach, namely the real annual growth of labor income that is assumed to prevail in the future, and the rate used to discount future earnings. Growth rates of the human capital stock, as well as its distribution across different groups of people, are less sensitive to the choice of these parameters.

Towards the construction of the human capital account, despite many challenges ahead, several attempts have been made to compile the flow and stock values of human capital in a systematic way by applying the lifetime income approach (e.g., WEI 2008, GU AND WONG 2010B).

<sup>20</sup> For more detailed discussions on the technical issues, besides the conceptual, methodological, and data issues, in national studies that applied the income-based, especially the lifetime income approach to measuring human capital, see Liu (2012).

# 4.2 International initiatives

Several researchers and international organizations have attempted to develop comparable measures of human capital. One example of current research in this field is represented by Barro and Lee (1993, 1996, 2001, 2010, 2013), for constructing an international dataset of educational attainment, school years, and schooling quality as proxies for human capital, based on census and survey information compiled by UNESCO and other sources.

Among international organizations, developing comparable measures of human capital has been a priority of the OECD. Much OECD work in this field has aimed at developing a better understanding of how teaching and learning outcomes can be improved in the classroom, and helping policy-makers to learn from each other's successes and failures. A large range of physical indicators is published in the OECD flagship publication Education at a Glance. Recently, the PISA has also attracted much international attention.

The OECD also has a long tradition in the field of measuring human capital beyond formal education. Earlier works include the investigation of further education and training and its impacts on the job market (e.g., OECD 1994). To deepen the understanding of the determinants of learning, attempts have been made to develop a framework for rethinking human capital information and decision-making; based on this framework, the OECD has analyzed obstacles to measurement, and suggested methods for further improvement in this area (OECD 1996).

In response to the growing interest in human capital, an OECD report in 1998 proposed an initial set of human capital investment indicators based on existing data. The report identified areas where significant gaps in internationally comparable data existed, and the cost of developing data collection for new measures and performance indicators (OECD 1998).

Building on the 1998 report, a subsequent report (OECD 2001) extended the OECD definition of human capital with a view to: 1) describe the latest evidence on investment in human capital and its impact on economic growth and wellbeing; 2) clarify the more novel concept of social capital; and 3) identify the roles of human and social capitals in realizing sustainable economic and social development. This report became an input into OECD projects on economic growth and sustainable development (OECD 2001).<sup>21</sup>

Since then, the OECD's work on human capital has continued along two lines. First, toward extending the measurement of students' competences in schools (PISA) to those of adults (PIAAC): In 2011, the PIAAC was launched with first results becoming available in 2013. The PIAAC program also links with the previous OECD work on the International Adult Literacy Survey (IALS) (see OECD AND STATISTICS CANADA 2000). Second, to identify the common methodology and data requirements for building human capital accounts: In cooperation with a number of national statistical agencies, a project was launched in 2009 by the OECD Statistics Directorate to compile monetary estimates of human capital for international and inter-temporal comparisons.

Liu (2011) summarizes the results from the OECD human capital project covering 15 OECD and one non-OECD countries over varying time periods from 1997 to 2007.<sup>22</sup> The results demonstrate the feasibility of applying the lifetime income approach to measuring human capital for comparative analysis, based on data currently available within the OECD statistics system. In addition, the estimated values of human capital

<sup>21</sup> To communicate the findings from OECD research to a wider audience, one book of the OECD Insights series summarized the work on human capital undertaken by the OECD in the message that "how what you know shapes your life" (Keeley 2007).

<sup>22</sup> The OECD countries are: Australia, Canada, Denmark, France, Israel, Italy, Korea, the Netherlands, New Zealand, Norway, Poland, Spain, the United Kingdom, and the United States, and the non-OECD country is Romania; later, estimates for another OECD member country (Japan) were added to the database (see Liu 2014). The detailed information on the country databases can be found at http://www.oecd.org/std/ publications documents/workingpapers/.

are in line with those reported by a number of national studies.

Beyond the OECD,<sup>23</sup> many other activities on measuring human capital in the international arena have taken place. These include the following:

- The UNECE/OECD/Eurostat Working Group on Statistics for Sustainable Development has worked to develop a broad conceptual framework for measuring sustainable development with the concept of capital at its core, and to identify a small set of indicators that might be used for international comparisons (UNECE 2009). The forthcoming report of a new UNECE/ OECD/Eurostat Task Force on measuring sustainable development will include a specific section on human capital measurement.
- The UNDP Human Development Index (HDI), which aims to illustrate the state of development of a society, is a composite index that combines measures of average achievements in a country in three basic dimensions of human development, i.e., health, education, and knowledge, and standards of living. The 2012 Human Development Report includes two measures of education and knowledge, namely school attainment, expressed in terms of the number of years of schooling, and school-life expectancy.<sup>24</sup>
- The EU KLEMS project has constructed a database (the EU KLEMS Growth and Pro-

ductivity Accounts) for empirical research of economic growth (see O'MAHONY AND TIMMER 2009). Although the primary aim of the EU KLEMS database is to generate comparative information on productivity trends, the data collected are also useful in other contexts. Thanks to its extensive country and industry coverage, potential applications of the database vary widely.

- The World Bank developed comprehensive wealth accounts including estimates of human capital for more than 120 countries, in order to answer the question "Where is the Wealth of Nations?" (WORLD BANK 2006). Beyond the snapshot of national wealth at a point in time, the World Bank extended the accounting of wealth over the decade from 1995 to 2005 and provided the first inter-temporal assessment of global, regional, and country performance in building comprehensive wealth and achieving sustainable development (WORLD BANK 2011).
- In 2012, the first UN "Inclusive Wealth Report", undertaken by the UN University - International Human Dimensions Programme on Global Environmental Change and the UN Environment Programme, presented estimates of inclusive wealth (the sum of manufactured, human, and natural capital) for 20 countries; in this approach, human capital is captured by measuring the population's educational attainment and the additional compensation over time of this training (UNU-IHDP AND UNEP 2012).<sup>25</sup>

<sup>23</sup> Other relevant streams of recent OECD works on human capital are the Social Outcomes of Learning project, the OECD Skills Strategy; work on intangible assets undertaken as part of the OECD work on New Sources of Growth; and the OECD Better Life Initiative. For more on these streams of work, please visit the following websites: http://www.oecd.org/document/9/0, 3746,en\_2649\_39263294\_33706505\_1\_1\_1\_1,00.html, http://www.oecd.org/dataoecd/58/28/47769132.pdf, http://www.oecd.org/dataoecd/60/40/46349020. pdf, http://www.oecd.org/document/0/0,3746 ,en\_2649\_201185\_47837376\_1\_1\_1\_00.html

<sup>24</sup> More information is available at http://hdr.undp. org/en/.

<sup>25</sup> Basically, the approach applied for measuring human capital in the *Inclusive Wealth Report 2012* is also income-based, but it is different from the original Jorgenson-Fraumeni approach. More discussions on the differences are given in Chapter 4 of this 2014 report.

# 5. Main issues and challenges ahead

Despite the emerging trend of countries to apply the income-based, and in particular the lifetime income, approach to measuring human capital, several issues and challenges remain. These challenges relate to both data availability and methodological issues, both of which are discussed below.

# 5.1 Data availability

The data necessary for the income-based approach are often currently not available for some countries, or are not in a form suitable for direct use. Based mainly on the OECD experience in compiling monetary estimates of the stock of human capital (LIU 2011), several issues stand out:

- First, the quality and sources of earn-• ings data cross-classified by different characteristics of workers vary significantly across countries. Data may refer to different earnings concepts (hourly and weekly earnings in most cases, annual and monthly earnings for some countries) and may include different elements of the remuneration packages of workers. In some cases, data on earnings refer only to the main job, while in other countries they may also cover secondary jobs and other remunerated activities. Finally, earnings data for different countries typically refer to different categories of educational attainment that may not be directly comparable, and may be collected as either point estimates or in the form of earnings brackets.
- Second, despite the great progress accomplished in collecting harmonized educational statistics, there remain issues with the quality of data on school enrollment and graduation rates, as definitions and classifications are not always comparable across countries.

This is often due to differences in educational systems and in ways of counting students (e.g., students who repeat the year, students who graduate for a second time, etc.).

• Third, human capital estimates would ideally require data on survival rates broken down by education. While some national estimates exist, and they highlight large mortality differentials by socio-economic characteristics, these breakdowns are not available for all countries, and they are rarely comparable across countries.<sup>26</sup> For instance, mortality statistics by educational level are not compiled through common standards across OECD countries, and in several countries simply do not exist (OECD ET AL. 2011).

More generally, constructing estimates of human capital based on the income-based approach requires that data from a range of sources – e.g., earnings statistics, population census, labor force surveys, mortality records – be integrated and harmonized to meet the requirements of human capital accounting, which is still practically difficult.

### 5.2 Methodological difficulties

Besides data issues, several methodological challenges need to be addressed. First, most human capital estimates currently available rely on the assumption that cross-sectional earnings data are good predictors of future cohorts' earnings. However, there is ample evidence that cohort effects are typically large. This suggests that it would be appropriate to use longitudinal earnings data that disentangles age and cohort effects, and which makes it possible to account for cohort-specific factors.

<sup>26</sup> The survival rates as constructed in e.g., Barro and Lee (2010) differ by educational attainment for some older age groups, and are differentiated by OECD vs. non–OECD countries.

Similarly, it would be important to separate wage premiums due to educational attainment from those due to adult-learning, on-the-job training, and other firms' characteristics, as failure to do so may lead to overestimates of the educational contribution to human capital.

With respect to labor market indicators (e.g., employment rates and earnings), it is also important to separate business cycle effects that distort comparisons (e.g., by depressing earnings or employment rates for different categories of workers during a recession).

A further difficulty when applying the lifetime income approach relates to the choice of some of the key parameters required by the method, such as the expected real growth of labor income in the future, the discount rate, and the price deflators used for temporal and country-comparisons.

While assumptions on these parameters are currently left to the discretion of researchers, their choice would ideally require further theoretical and empirical grounding. In other words, clear guidance in each of these fields remains needed. Similar challenges also confront the cost-based approach with respect to the choice of depreciation rates and price deflators.

Another challenge for developing monetary measures of the stock of human capital can be seen in the large discrepancies between estimates based on the income-based and the costs-based approaches. These discrepancies should be better understood and reconciled. One way to address this challenge would be to apply the two approaches simultaneously, which would offer an opportunity to identify the main factors accounting for the differences and to reconcile the two methods.<sup>27</sup> Satellite accounts could be used for such purpose, as they would allow for linking stock and flow measures of human capital in a fully-fledged accounting system consistent with the SNA.

As reflected by existing national studies and international activities, almost all currently available monetary measures effectively ignore the non-market benefits of human capital investments. This does not imply that non-market benefits are not significant or important, if compared with market benefits. It simply reflects the difficulties in developing estimations of them. Therefore, these current measures will tend to "under-estimate" the value of the human capital relative to an "ideal" norm that would include a monetary estimate of these non-market benefits.

# 6. Conclusions and recommendations

The concept of human capital has evolved over time, from a narrow scope focusing on cognitive knowledge, working skills, and the economic returns associated with them, to today's more comprehensive definition embracing a broader range of individual attributes and resulting benefits. The human capital concept defined by the OECD (2001) has come to be widely accepted.

However, implementing this comprehensive concept raises significant measurement challenges. The multi-faceted nature of human capital, the complex links between the various types of human capital investment, and the diverse benefits that it delivers make it impossible to find a one-size-fits-all measure of human capital. By necessity, the measurement of human capital must be undertaken step by step.

Currently, many countries use definitions of human capital that focus on the productive capacity of individuals. Even among the countries that refer to the broader OECD definition, most measurement initiatives focus on formal education and the economic returns for individuals, rather than to human capital in general and to all the benefits (economic and non-economic, private and collective) from

<sup>27</sup> An excellent discussion on the possible explanations about the large divergence of the human capital estimates between the cost-based and the incomebased approaches can be found in Abraham (2010).

human capital investment. Given the current state of knowledge, this seems to be a practical and reasonable point of departure.

Following from this narrower focus, measurement activities in this field have initially aimed to develop summary indicators providing simple proxies for human capital (e.g., average years of schooling, educational attainment). While the data requirements of such indicators are limited, so is the scope of these proxies. As a result, human capital measurement has gradually moved in the direction of quantifying the knowledge and cognitive skills of students and adults after they left school. In more recent years, the challenge of developing monetary measures of human capital in a systematic way has garnered increasing interest from independent researchers, national statistical offices, and international organizations.

All the approaches to measuring human capital reviewed in this chapter have both advantages and disadvantages. Depending on the purpose, different approaches may be applied individually or jointly to address different issues. However, the monetary measures generated from the cost-based and income-based approaches should arguably be designated a "core" status. One reason for the growing interest in monetary measures of human capital is that these measures can be compared with those for traditional produced capital covered by the SNA, the construction of which being a primary task of national statistical offices.

Even if limited in terms of the range of benefits considered, the policy implications of accounts with monetary measures for human capital are potentially great, as they imply that expenditures related to human capital formation should be considered as a form of investment rather than consumption, which is unfortunately not the case within the current framework of the SNA.

Based on country experiences and international initiatives in the field of human capital measurement, an international trend is emerging toward the income-based approach. Estimates based on this approach can be used to assess the relative contribution of a range of factors (demographic, education, and labor market) to the evolution of human capital, facilitating corresponding policy interventions.

Recent international experience also suggests the feasibility of producing this type of measures based on information already available within the national and international statistical systems (e.g., LIU 2011). However, the scope for improvements in terms of consistency and comparability of the underlying data, as well as the detailed methodological choices, remain significant.

Given the existence of large discrepancies between estimates based on the cost-based and the income-based approaches, it could be interesting to change perspective and consider the two approaches not as alternatives, but rather as complements of each other, viewed within a more comprehensive information system. Such a comprehensive system could be described by means of human capital (or educational) satellite accounts.

Constructing human capital satellite accounts linked to the SNA, rather than incorporating human capital measures directly into the core accounts of the SNA, is also justified by the following reasoning: human capital is presently out of the asset boundary, and extending the production and asset boundaries to incorporate it would fundamentally change the SNA (although steps in the direction of expanding the asset boundary have already been taken in recent years, for instance by treating research and development as a "produced asset").

The satellite accounts should be modeled along the lines proposed by e.g., Jorgenson and Fraumeni (1989). This would include direct links to the SNA through market labor services and output of the educational sector. At a later point, depending on the level of complexity, the choice of a more or less broad definition of human capital, and more or less exhaustive inclusion of the inputs and outputs associated to human capital investment, should be determined appropriately.<sup>28</sup>

Against this backdrop, the following recommendations can be made:

- Studies should be carried out to investigate in detail the discrepancies between estimates of human capital based on the cost-based and the income-based approaches, with the goal of better understanding, and eventually reconciling, such discrepancies.
- Initiatives should be undertaken to influence the type of data collected and harmonized internationally, so as to facilitate improving the quality of these monetary estimates of human capital for use in international and inter-temporal comparisons.
- Research should be encouraged to construct experimental satellite accounts for human capital as a mid-term goal, based on common methodologies and on agreement regarding the ambition of such accounts.
- Work should be pursued to estimate noneconomic returns to human capital, with the objective of incorporating these estimates in more sophisticated types of satellite accounts in the long-term, and to incorporate human capital measures into the SNA in the future.

<sup>28</sup> For more details on the rationale and feasibility of developing human capital satellite accounts with varying levels of complexity, please refer to e.g., Abraham and Mackie (2005) and Boarini et al. (2012).

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# Human capital: country estimates using alternative approaches

Barbara Fraumeni and Gang Liu

# KEY MESSAGES

**Human capital is critical to** individual and societal well-being.

The educational attainment of a country's younger cohort is frequently higher than the educational attainment of the older cohort; high levels of youth educational attainment correlate to high potential for improved wellbeing and economic growth in the future.

Human capital indicators which depend solely on educational attainment information fail to capture the full potential of a country's population.

Human capital measures including information on present and future demographic trends, education, and wage or income components are essential for appropriate policy formulation and analysis.

# 1. Introduction<sup>1</sup>

The previous chapter outlines how human capital can be conceptualized, both generally, as well as with specific reference to the System of National Accounts (SNA), and introduces a taxonomy of how it might be measured. The chapter also reviews how national statistical agencies and other researchers have implemented different approaches to measuring human capital, and briefly discusses challenges related to such implementation.

This chapter presents country estimates of human capital based on comparisons of Barro-Lee educational attainment estimates and Jorgenson-Fraumeni (J-F) lifetime income estimates for 18 countries. In a previous chapter, inclusive wealth estimates for all categories of wealth are presented and discussed for a large number of countries. Inclusive Wealth methodology for human capital is described in this chapter.

As part of the 18-country comparison, which is the focus of this chapter, a gap analysis is undertaken using both Barro-Lee and J-F approaches. This gap analysis illustrates significant disparities between the younger cohorts (aged 25 to 34) and the older cohorts (aged 55 to 64). This chapter concludes by briefly discussing how human capital estimates can shed light on specific policy issues.

An analysis of country-level human capital measures shows that such measures are critical indicators, both of individual, as well as nationallevel potentials for growth and future well-being.

# 2. Barro–Lee, Inclusive Wealth, and Jorgenson–Fraumeni methodologies to measure human capital

Three methodologies are described in this section: The first, Barro-Lee, is an indicatorbased measure, while the other two, Inclusive Wealth and Jorgenson-Fraumeni, are monetary measures.

### 2.1 Barro-Lee methodology

The most widely-used human capital proxy or indicator data set is that of BARRO AND LEE (2013A, 2013B). Average formal education attainment is reflected in this data set beginning at age 15, in five-year age increments, for the total population (for females, data is available for every five years from 1950 to 2010), for 146 countries. Population numbers are also available for each associated educational attainment estimate. Benchmark data is collected from census and/or survey information and compiled by UNESCO, Eurostat, national statistic agencies, and other sources.

Barro-Lee uses a variety of techniques to fill in gaps in observations and educational attainment subcategories, and to avoid misestimation of average years of schooling.

To fill in missing observations (as benchmarks are not available for all five-year periods) they begin by calculating the distribution of educational attainment among four broad categories: no formal education (hu), primary (hp), secondary (hs), and tertiary education (hh). Primary and tertiary are further divided into complete and incomplete; secondary is further divided into lower secondary and upper secondary.

Most missing observations are filled in with backward or forward extrapolation with an appropriate time lag. There are 12 5-year age groups (*ag*), from ag=1 (15-19) to ag=12 (70-74), plus one age group ag=13 (75 and over).

The forward extrapolation method assumes that the educational attainment distribution

<sup>1</sup> Many of the estimates in this chapter are from the work that was carried out when one of the authors (Gang Liu) led the human capital project at the OECD. Any remaining errors and omissions are the sole responsibility of the authors.

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of an age group ag at time t is identical to that of the age distribution that was five years younger at time t-5.

EQUATION 1

hj(ag, t) = hj(ag-1, t-5)

for *j*=*u*, *p*, *s*, or *h* and *ag*=3 (25-29) through *ag*=11 (65-69).

As those younger than 25 are potentially still in school, a different methodology is employed.

Similarly, backward extrapolation assumes that the educational attainment distribution of an age group ag at time t is identical to the age distribution that was five years older at time t+5.

### EQUATION 2

hj(ag, t) = hj(ag+1, t+5)

for *j*=*u*, *p*, *s*, or *h* and *ag*=3 (25-29) through *ag*=11 (65-69).

The net effect of this methodology is to hold an individual's educational attainment constant from age 25 through 64.

For older individuals, the probability of dying differs by educational attainment level. Accordingly, for the three oldest age groups: ag=11 (65-69), ag=12 (70-74), and ag=13 (75 and older), survival probabilities are estimated by educational attainment level. Highly educated individuals live, on average, longer than their less educated peers; this correction is necessary to ensure accurate estimations of average educational attainment for older age groups. For all younger age groups (ag=10 and below), it is assumed that survival rates do not differ by educational attainment.

The process for creating subcategories of educational attainment (complete and incomplete for primary and higher education; lower and upper for secondary school) depends upon the age level.

For primary school, Barro-Lee use country and age-specific completion ratio profiles to estimate the subcategories for ag=1 (15-19) and ag=2 (20-24). For ag=3 (25-29), the primary school completion rate is set equal to the ratio

of the number of individuals who completed primary school, but did not enter secondary school, to the number of individuals who entered primary school. Backward and forward extrapolation and other methods are used to fill in any missing observations for ag=3 (25-29) and above.

When there are missing observations, secondary-school enrollees for ag=1 (15-19) are assumed to be incompletely educated at the secondary level, and higher-school enrollees for ag=2 (20-24) are assumed to be incompletely educated at the higher level.

Other estimation problems arise because some countries do not report the proportion of the population who have no formal education, but do report on the proportion of the educated population who have achieved primary, secondary, or tertiary level of education. Alternatively, the proportion of the population with no formal education, or has achieved at most some level of primary education, is often reported as a single number. Barro-Lee uses illiteracy rate, primary enrollment ratio, and/or data from other census years to resolve such inconsistencies.

Finally, estimations are made for average number of years of schooling for those aged 15 and above, and separately for each of the 13 age categories. For those aged 15 and above:

# EQUATION 3

 $S(t) = \Sigma l(ag,t)s(ag,t)$ 

where the summation is over all age groups, l(ag,t) is the population share of the group aged ag in the total population aged 15 and above, and s(ag,t) is the average number of years of schooling for age group ag.

The average number of years of schooling by age group *ag* is:

# **EQUATION** 4

 $S(ag,t) = \Sigma hj(ag,t)Dur(j,ag,t)$ 

where the summation is over educational levels j (p, s (incomplete, complete), h (incomplete, complete)), hj(ag,t) is the fraction of the group aged ag with the educational level j, and Dur is the duration of school attendance in years.

# 2.2 Inclusive wealth methodology<sup>2</sup>

The inclusive wealth (IW) human capital methodology follows that of Arrow, Dasgupta, et al. (2012a, 2012b, AD) and Klenow and Rodríguez-Clare (2005). In this report, country aggregates are determined for 140 countries. The country aggregates, separated by gender, which enter into the calculation are: average formal education attainment, average wage, total number of employed, total adult population, and average expected remaining working years.

The first step is to estimate human capital per capita. Fc. lowing KLENOW AND RODRÍGUEZ-CLARE (2005), education is assumed to earn a market rate of interest,  $\rho$ , of 8.5 percent per annum. Human capital per person is:

EQUATION 5

# $h = e^{(Edu * \rho)}$

where *Edu* is the average number of years of educational attainment in a formal setting (from Barro-Lee). As expected with an exponential function, human capital per person rises at an increasing rate with increases in the average number of years of educational attainment. Note that the human capital of a person with eight years of education is estimated to be almost twice that of a person with no education.

As all adults have human capital, even if they do not work, h is multiplied by the total number of adults in the country to determine total human capital. The number of adults in the country is defined as the number of individuals of age (*Edu*+5). As *Edu* varies by country, the age of someone who is considered to be an adult varies significantly by country.

IW assumes that the labor market is sufficiently competitive such that the marginal productivity of human capital can be assumed to be equal to the real wage rate, *r*. The shadow price for a unit of human capital is then calculated as the real wage rate, *r*, discounted over the expected number of working years remaining, *T*, for the average adult:

EQUATION 6

$$P_{HC} = \int_{t=0}^{T} r \cdot e^{-\delta t} \, \mathrm{dt}$$

where  $\delta$ , the discount rate, is set equal to 8.5 percent per annum as before. IW uses World Health Organization life tables and U.S. Census Bureau demographic data by country and gender to calculate the average expected remaining working vears, T, across all individuals of working age. The real wage is computed as a country's average total wage bill divided by the total number of workers in the country, over the 1990 to 2010 time period. The expected number of working years remaining is estimated by contemporaneous (as opposed to expected future) age-gender participation and mortality rates. Labor market information, such as employment, wages, and labor force participation come from a variety of sources, including the International Labour Organization, the Conference Board, and the United Nations Statistics Division.

Country total human capital depends on the average educational attainment, real wage rate, and adult population. Total human capital is divided by the total population, regardless of age, to determine country level per capita human capital:

## EQUATION 7

$$\frac{Human Capital}{Total Population} = e^{Edu * \rho} \cdot \frac{Population_{(5 + Edu)}}{Total Population} \cdot \int_{t=0}^{T} r \cdot e^{-\delta \cdot t} \cdot dt$$

The first term on the right-hand side of the equation is the human capital per capita expression from Equation 5; the third term on the right-hand side of the Equation is the shadow price for a unit of human capital expression from Equation 6. The age composition of the population, entering through the second term on the right-hand side of the equation, clearly impacts the above human capital ratio. Population estimates are from the Population Division of the United Nations Department of Social and Economic Affairs.

<sup>2</sup> This section relies heavily on the appendix to Arrow et al. (2012b). Also see Arrow et al. (2012a).
## 2.3 Jorgenson-Fraumeni methodology

The J-F lifetime income approach applies the neoclassical theory of investment (JORGENSON 1967) to human capital. According to this theory, the price of capital goods depends upon the discounted value of all future capital services derived from the investments. On a per capital basis, this means that the value of the human capital of an individual can be determined from that person's discounted lifetime income.

The J-F methodology (1989, 1992a, 1992b) is modified, most notably by Liu (2011), to reduce estimation difficulty and time requirements; to deal with data availability constraints; and to reflect country-specific conditions. In addition, almost all country studies have estimated only market lifetime income because of the additional assumptions, time, and data needed to include nonmarket lifetime income as part of human capital.

The following sets of data for a J-F simplified approach (FRAUMENI 2008) as implemented by Liu (2011) are required, except as noted for ages 15 through 64 and gender: 1) working age population; 2) survival rates; 3) school enrollment rates for ages 15 through 29 by single year, ages 30-34 and 35-39 by five year categories, and 40 and above; 3) educational attainment; and 4) annual earnings.

The simplified approach identifies three life stages.<sup>3</sup> The characteristics of these stages are dictated by typical life stages and data availability. In the equations that follow, the following notation is used:

*mi:* Expected lifetime market income per capita, discounted to the present

*R*: The adjustment factor applied to lifetime income

= (I + real rate of growth on labor income)/(I + real discount rate)

*sr:* Survival rate *senr:* Formal school enrollment rate and *ymi:* Yearly market income per capita.

For subscripts: *a*: Age *e*: Highest level of education completed *enr*: Formal education enrollment level *older*: Equal to *a* + 1 *s*: Gender, and *school*: Equal to *e* +1

The nominal market value life stage equations are as follows:

## Stage 1: Work and school, ages 15 through 40 when an individual could be enrolled in school

For these ages, individuals can attend school and perform market work. It is assumed that dropouts do not later continue their education, that no grades are skipped or repeated, and that once enrolled, a student finishes that year of education. Market hours are valued at the average wage or income paid for the corresponding gender, age, and highest education level completed category. In stage I, individuals earn income in the current year, and if they survive for another year, can earn the lifetime income of someone who is a year older than the individual's current age. The individual's future lifetime income is dependent upon whether they are enrolled in school or not. Finally, as is true for both stage 1 and 2, the income sum is adjusted by *R*, the factor reflecting a future real rate of growth in labor income and discounts the income sum back to the present.

#### EQUATION 8

mi(s,a,e)=ymi(s,a,e)+[senr(s,a,enr)\*sr(s,older)\*m i(s,older,school)+(1-senr(s,a,enr))\*sr(s,older)\*mi (s,older,e)]\*R

Stage 2: Work only, ages 41 through 64 when it is assumed that an individual is not enrolled in school

For these ages, it is assumed that no one is enrolled in school, as insufficient data existed

<sup>3</sup> Jorgenson-Fraumeni estimated nonmarket lifetime income for two more stages to include those too young to be doing market work. Their first stage (ages 0 through 4) is a no market work, no school stage. Their second stage (ages 5 through 15) is a no market work, school stage.

on students above the age of 40. Human capital therefore depends only on a person's expected future market income and whether the individual survives for another year.

#### EQUATION 9

mi(s,a,e)=ymi(s,a,e)+sr(s,older)\* mi(s,older,e)\*R

#### Stage 3: Retirement, age 65 and over

When only market lifetime income is counted in a J-F computation, the human capital of retired persons is zero. Because of data constraints, it is assumed that everyone aged 65 or older is retired.

# EQUATION 10 mi(s,a,e)=0.

Calculations are done in a backwards recursive manner, starting from the oldest age group and continuing to the youngest age group. For example, for a particular year – say, 2000 – the computations start by setting the lifetime income of someone who is 65 equal to zero. If lifetime incomes are being computed by single year of age, the next calculation would be for a 64-year-old: Because the lifetime income of a 65-year-old is zero, the 64-year-old's lifetime income is equal to the income that person earns in 2000. All but the first term in Equation 9 drop out.

For someone who is 63 in 2000, there are two possible components to their lifetime income: income earned in the current year and, if they survive for another year, lifetime income of someone who is 64 in 2000. The 63-year-old's future lifetime income is adjusted for a one-year change in the real wage rate and discounted for the one year before the 63-year-old in 2001 receives the thusly adjusted income of a 64-year-old in 2000.

The sequence continues backwards, with each step reducing the age of the person for which the computation is made by one year. The future lifetime income of a 63-year-old, should they live until age 64, has already been adjusted for a one-year change in the real wage rate and discounted. Accordingly, for a 62-year-old, there is only a one-year real wage rate and discounting adjustment appearing in the equation.

It is assumed that the relative wage rates by educational attainment levels are determined by contemporaneous relative wage rates, survival rates, and enrollment rates. For example, the information regarding the probability that someone who is 20 in 2000 will enroll in school and survive until he is 21 in 2001, as well as the wage the person will earn in 2001 compared to someone who does not continue in school in 2000, is predicted by the information about someone of the same gender who is 21 in 2000, perhaps with one more year of school completed in 2000.<sup>4</sup>

Total nominal human capital is constructed by multiplying each stage's market income per capita by the population of the corresponding ages, which is then summed to determine total human capital across ages 15 to 64. For all stages except for stage 1, which involves possible school enrollment, a total stage population suffices. For stage 1, population by single year of age or by fiveyear categories for some ages is required.

Two different types of volume indices are constructed. Divisia (Tornqvist) temporal volume indices are constructed with a weighted growth rate to compare stocks of human capital over time. The weights are nominal human capital and the growth rates are population growth rates for the corresponding age/educational category. Spatial indices are derived by dividing nominal human capital by purchasing power parities (PPPs) to compare human capital in real terms between different countries at one point in time.<sup>5</sup>

#### 3. Barro-Lee, IW, and J-F comparisons

There are several similar data constructs in each of the three approaches outlined in this chapter. Barro-Lee, IW, and J-F all contain information by

<sup>4</sup> Because of the rapidly changing school enrollment rates in China, the probability that a Chinese student will enroll in school when they are one year older was allowed to increase. See, for example, Li (2012).

<sup>5</sup> Liu (2011).

gender on population, educational attainment, and survival. However, the amount of detail and/ or the level of aggregation differs.

For population, IW ultimately needs the least amount of detail as it uses only total population and adult population in its final equation (Equation 7). Barro-Lee publishes population by the five-year age categories it covers. J-F requires the most detailed population numbers for the first (school enrollment) stage, but its other categories require less population detail than Barro-Lee with a Liu (2011) simplified approach.<sup>6</sup> All three approaches also employ other population information to derive their final results.

Barro-Lee educational attainment data by five-year age groups is an input to the IW estimates. J-F educational attainment is interpolated between five-year age groups by Liu to obtain educational attainment by single year of age. Only J-F includes expected future education as an input to its calculations. All three approaches used similar data sources to determine survival rates.

IW does not need school enrollment figures to produce estimates of human capital. Barro-Lee use enrollment data to solve a number of missing data problems. J-F use enrollment data directly in its stage I work and school calculations. IW and J-F are distinctly different from Barro-Lee in that both require wage or income data.

The analytical power of IW and J-F as compared to Barro-Lee arguably comes from their inclusion of current and future expected income, as well as – in J-F – future expected education.<sup>7</sup> Both IW and J-F capture current income and future expected income based upon current age-gender labor participation, mortality rates, and wages or income. That J-F allows for future education, notably for younger individuals, is an important distinguishing feature compared to the other two approaches. Clearly, future investments in human capital through education, as well as the contribution of individuals to the economy through income generation, are relevant.

The next four figures illustrate comparisons between J-F and Barro-Lee for working age populations in 18 countries, to illustrate how different human capital perspectives can be depending upon what measure one uses. The comparisons are shown with J-F instead of IW because, in general, J-F will show larger variance with Barro-Lee than does IW, as J-F allows for future education. In addition, IW results are discussed in detail in another chapter.

First, in order to have a clear picture of the J-F ranking of countries, Figure I presents 2006 J-F human capital per capita estimates in thousands of nominal US dollars.<sup>8</sup> Per capita income, adjusted by a private consumption PPP (WORLD BANK 2014), shows a wide variation: from US\$71,000 for China to US\$641,000 for the United States.<sup>9</sup> <sup>10</sup> Table I sorts the countries into broad lifetime income per capita categories.

Figure 2 shows that there is frequently little ranking correspondence between the average Barro-Lee educational attainment of a country

<sup>6</sup> The original Jorgenson-Fraumeni estimates (1989, 1992a, 1992b) were done by single year of age through age 74, by single year of educational attainment, and enrollment through undergraduate college. The simplified approach is described in Fraumeni (2008) and implemented with modifications in Liu (2011).

<sup>7</sup> Others may use Barro-Lee to derive estimates of income and future educational attainment, but Barro-Lee do not present such estimates in their database.

<sup>8</sup> The J–F estimates are for all countries in 2006 except for Australia: 2001 and Denmark: 2002. As this is a spatial index, all countries' figures are adjusted by private consumption purchasing power parity from the World Bank (2014). Estimates for China are from Li (2012). Estimates for India are from Gundimeda et al. (2007). Estimates for all other countries are from the OECD Human Capital Project. The results for all project countries except for Japan are described in Liu (2011).

<sup>9</sup> Returns to human capital depend on a number of factors not discussed in this chapter, such as the availability and use of nonhuman capital.

<sup>10</sup> An alternative approach to adjusting country estimates by PPPs is to show comparisons based upon national currencies. Neither approach is perfect. As PPPs attempt to compare currencies based upon a representative market basket of goods, they are preferred. Results based on national currencies will show wide variations and differences, as do the PPP adjusted results.

#### FIGURE 1

International comparison of J–F human capital per capita, individuals aged 15 to 64, 2006

Abbreviations: CHN (China), IND (India), ROU (Romania), POL (Poland), ITA (Italy), ISR (Israel), NZL (New Zealand), ESP (Spain), NLD (the Netherlands, AUS (Australia), DNK (Denmark), FRA (France), JPN (Japan), KOR (South Korea), CAN (Canada), NOR (Norway, GBR (Great Britain), USA (United States).



and its nominal J-F human capital per capita. In addition, it shows that there is significantly more variation in human capital per capita than in average educational attainment.<sup>11</sup>

Reasons for this include differential returns to education and the age by education distribution of the working age population. The biggest pictorial "gap" between average educational attainments is for the countries in the low per capita lifetime income category. There are substantial gaps for other countries as well, e.g., for Israel and New Zealand, both of which are in the lower middle per capita income category. Great Britain is ranked second highest among the 18 countries in J-F human capital per capita, yet has the third-lowest average educational attainment. J-F databases provide a rich starting point basis to explore the reasons for these ranking and variation differences, as they include demographic information, as well as information on current income and expected future income.

Such an exploration is best augmented by an analysis of differences in a country's labor markets. It does not suffice to note simply that returns to education vary widely, or to know from J-F how relative wages differ across individuals with different levels of educational attainment or of different ages. A comparison of Barro-Lee and J-F results begs a host of interesting questions, the answers to which are critical for developing appropriate policy interventions. The comments that follow are speculative, and intended only to provide suggestions for further study.

Changes in educational attainment over time might provide a clue as to why aggregate measures of human capital differ. The next two figures illustrate changes in educational attainment between the younger working age population, who have most likely finished their

#### TABLE 1

Ranking of countries by J–F per capita lifetime income

| Low          | China, India, Poland, and<br>Romania                      |  |
|--------------|---|--|
| Lower middle | Israel, Italy, and New Zealand                            |  |
| Upper middle | Australia, Denmark, France, the<br>Netherlands, and Spain |  |
| Lower high   | Canada, Japan, and South Korea                            |  |
| Upper high   | Great Britain, Norway, and the<br>United States           |  |

<sup>11</sup> The United States is the ideal reference point for this graph as it has both the highest per capita income and the highest average educational attainment.



### FIGURE 2

Average years of education 2005 compared with human capital per capita 2006, individuals aged 15 to 64

## Key Human capital per capita (US\$ in thousands) Years of education

#### FIGURE 3

Comparison of education attainment, 2005. Between ages 25–34 and 55–64, with ages 15–64 average comparison





education – those aged 25-34 – and the older working aged population – those aged 55-64.

Figure 3 plots the average educational attainment for the whole working aged population - those aged 15-64 (shown with a line) - versus the younger group (shown with diamonds), and the older group (shown with squares). It is not surprising that for all but two countries the younger group on average has a higher level of educational attainment than the older group. The two exceptions are Denmark and Norway. However, the average educational attainment of the younger and older group is almost identical for a number of countries: Australia, Norway, Romania, and the United States. The higher educational attainment of the younger group is expected to result in their higher future income. This is explicitly recognized in J-F, but only implicitly recognized in Barro-Lee.

Figure 4 illustrates the younger-older gap story in a way that makes it easier to see the size of the gaps. The 18-country average is weighted by population average; clearly the lower educational attainment of the two most populous countries in the world skews the average towards the lower end. As these estimates are for 2005, they should be little if at all impacted by the onset of the recession and the financial crisis.

One is tempted to conclude that the gap is small for countries for which there is little differential return to education - perhaps due to taxation. But this does not seem to be consistently borne out by Figure 4. It is well known that incomes, even by education categories, differ relatively little in Denmark and Norway. In the United States, by contrast, there are very substantial returns to higher education, yet the educational attainment gap is very small. Between the younger and older groups, there are very large differences in educational attainment in emerging China and South Korea; there are also very large differences in educational attainment in developed France, Italy, and Spain.

The final table and figure combine the information on the younger versus older educational attainment graph and the J-F human capital per capita. Table 2 shows the results in a crosstabulation format; Figure 5 shows the results in a quadrant format. In Figure 5, the country markers are colored according to the quadrant in which they appear. The intersection of the gap axis and the human capital per capita axis occurs at a 1.25 years gap in educational attainment between the younger and the older individuals and a US\$435,000 lifetime income per capita. In both cases, these values are approximately at the mid-point of their categories.



|   |        | Level of human capital per capita |                   |                   |
|---|--------|-----------------------------------|-------------------|-------------------|
|   |        | Low                               | Medium            | High              |
| Size of<br>education<br>attainment<br>gap | Small  | ROU                               | AUS<br>DNK        | NOR<br>USA        |
|   | Medium | POL                               | NLD<br>ISR<br>NZL | CAN               |
|   | Large  | CHN<br>IND                        | ESP<br>FRA<br>ITA | KOR<br>JPN<br>GBR |

#### TABLE 2

Cross tabulations of younger vs. older educational attainment and human capital per capita The two former Soviet Union countries (Poland and Romania) are in the small gap and low per capita human capital category. This may reflect the relative scarcity of economic institutions and their weakness in actual and expected returns to higher education. By contrast, the emerging country power-houses China and India are in the low per capita income per capita category, but are in the medium or high gap categories. Expectations about continuing future growth with an increasing reward to education may explain the differences between China and India on the one side and Poland and Romania on the other.

As previously noted, it is difficult to easily explain why France, Italy, and Spain have large gaps without the detailed income wage and income information in J-F accounts and an understanding of country-specific labor market conditions, but it is fairly simple to postulate why China and South Korea are in the large gap category.

The high per capita income countries are all highly developed countries, with the exception of South Korea. Between the table and the figure, it is fairly easy to get a sense of the cross-categorizations of the 18 countries, but they alone cannot answer why countries appear in certain cross-categories without additional information.

There is a significant clustering of countries around the gap axis in Figure 5. In the lower two quadrants, it is easy to see that China, India, Poland, and Romania are outliers, as their J-F income per capita is at least US\$150,000 below that of other countries. In the upper two quadrants, Great Britain, Norway, and the United States could also be classified as outliers, but this is less clear, even though their J-F income per capita is at a least about US\$40,000 higher than that of other countries.

#### BOX 1

# Policy implications of human capital wealth analyses

Human capital wealth measures that combine demographic, education, and wage or income information provide a rich basis for policy analysis. Both the J-F and IW human capital measures include these three elements.

For instance, China is considering revising both its well known onechild, as well as retirement, policies. In fact, the Communist Party's Central Committee has already begun modifying the former (XINHUANET 2013). There is concern about the demographic pressures associated with a relatively small younger cohort together with a rapidly aging population, particularly in a country with a long tradition of the young supporting their elders.

At the 5th International Symposium on Human Capital and the Labor Market in December 2013 in Beijing, a discussion of both of these policies was central to presentations by two of the keynote speakers: Jiang Zhenghua and Li Haizheng. Mr. Jiang, who served as Vice-Chair of the Standing Committee of the 9th and 10th National People's Congress, is coordinating a large research project to consider changes in both policies. Although he referred to changes in the definition of elderly, rather than to changes in the retirement age, the connection between a change to the definition of elderly and a possible change in the retirement age is easy to make.<sup>12</sup>

12 In China, the normal mandated retirement age for a female is 55; for a male it is 60. A recent article in China Daily USA (HE 2014) noted than an increase in the retirement age would substantially decrease social security deficits. In the same article, a World Bank expert concluded that raising the retirement age could be justified given China's life expectancy has now risen to 73 years. Dr. Li, in his presentation at the symposium, looked at the implication of a change in the retirement age with a J–F human capital measure (LI 2013). Raising the retirement age would do nothing to the average educational attainment of those beyond school years, but it would significantly change the human capital measures for China, as individuals could be earning substantial income beyond the current retirement age.<sup>13</sup>

As the younger cohorts in China (and elsewhere) are very likely to be more highly educated than the older cohorts (see Figure 4), demographic changes can be very important to the future well-being of any country. Policies which affect on birth rates, retirement age, access to education, and health care all will be reflected in the J–F and AD human capital wealth measures through their income measures.

Other policies of relevance to human capital wealth are those impacting internal and external migration and emigration. Migration and emigration affect both the size and geographical distribution of the population, which in turn affect wages and total income. Policies which restrict or facilitate mobility accordingly are reflected in J–F and AD measures of human capital.

Examples of immigration restriction include the policies of the United States and, more recently, those

13 Currently individuals past retirement age in China may simply be switching jobs, although to a lower paying or more part-time position. of Switzerland; an example of immigration facilitation would be the European Union policies allowing migration between member countries.

The immigration reform debate in the United States is complex, but many analysts point to the importance of allowing more highly-skilled individuals, particularly so-called STEM (Science, Technology, Engineering, and Mathematics) workers, and those foreigners with advanced degrees, to immigrate to the United States (TECH AMERICA undated). It is recognized that selected immigration can significantly increase the human capital of a country and stimulate economic growth.

A recent news article notes that the percentage of international students in the U.S. who are majoring in a STEM field in college has risen (BIDWELL 2014). At the same time, according to Rachel Banks, director of public policy at NAFSA, Association of International Educators, the percentage of all college students who are international students has not risen over the past ten years. In her opinion, this is because some countries, such as the United Kingdom, Australia, and New Zealand have more lenient immigration policies. If true, this shift in the geographic distribution of international students foretells possible improvements in the human capital position of these countries, and a possible decline in the human capital position of the United States. As the income return to higher education in the U.S. is clearly much greater than the return to a high school education, the impacts on Barro-Lee average educational attainment would be less than the impact on a J-F human capital measure.

These two examples are only illustrative; there are many other policies which can be analyzed using a robust set of human capital wealth accounts.

#### 4. Conclusion

Human capital accounts can inform decisionmakers and could be an important complement to national income accounts. These accounts, if fully explored, could facilitate improved understanding of both current and future economic growth, and help shape specific policies, such as those targeting the optimal mix of investments in human and nonhuman capital. Also, there are ample opportunities for future research to attempt to explain why countries significantly differ in their human capital.

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# **CHAPTER 5**

# Health capital

Kenneth Arrow, Partha Dasgupta, and Kevin Mumford

# **KEY MESSAGES**

Health is an essential characteristic of human well-being.

**Health capital is an important** part of inclusive wealth.

The economic model of health capital presented in this chapter allows health to affect human well-being through three distinct channels: direct well-being, productivity, and longevity. Most health capital services influence human well-being directly rather than through the production of goods and services that are counted in GDP.

**In the absence of better** estimates of the direct and productivity effects, gains in life expectancy should be used as the primary measure of health capital.

**Annual gains in health capital** in the U.S. are worth approximately US\$10,000 per person.

#### 1. Introduction

Attempting to measure human well-being without considering health would be a great oversight. Health is central to our happiness. Health affects our enjoyment of life, our productivity in employment, and our risk of death. Our desire for good health influences our decisions regarding eating, sleeping, exercising, and our demand for medical services. As shown in Table I, total spending on medical care from both public and private source makes up an important and generally increasing share of national income in many countries.

The improvement in life expectancy at birth has been quite dramatic in most countries over the past 60 years (see Figure 1). Several studies, including Nordhaus (2005), Becker et al. (2005), Murphy and Topel (2006), and Jones and Klenow (2011), have shown that recent gains in life expectancy have been at least as important to human welfare as gains in income. The Inclusive Wealth Report 2012 treated health as a form of wealth by estimating the value of the improvement in life expectancy over a nineteen-year period. However, health capital was treated separately from other forms of capital because it was found that even modest gains in life expectancy outweighed other gains. Though understandable, this is not a theoretically-sound reason to exclude health capital from an inclusive measure of national wealth.

#### TABLE 1

Total health expenditure (percentage of GDP)

| Country       | 1995 | 2000 | 2005 | 2010 |
|---------------|------|------|------|------|
| Brazil        | 6.7  | 7.2  | 8.2  | 9.0  |
| China         | 3.5  | 4.6  | 4.7  | 5.0  |
| Germany       | 10.1 | 10.4 | 10.8 | 11.5 |
| India         | 4.0  | 4.3  | 4.2  | 3.7  |
| United States | 13.6 | 13.6 | 15.8 | 17.7 |

Source: The World Bank (2013), World Development Indicators

#### 2. Health as a capital asset

Health is a multidimensional concept. There is no single standard way to measure the health of an individual or a population group. A physician may examine a patient and measure health along several dimensions including mental health, severity of illnesses, nutrition, body mass index (BMI), risk of disease, and level of pain or discomfort. An individual may track exercise and eating behavior or rate his or her own subjective health along a scale of overall fitness. For a population group, a researcher may use life expectancy, infant mortality rate, availability of healthcare services, or prevalence of preventable diseases as indicators of the health of the group. Our term, health capital, refers to a satisfactory measure of the overall health of an individual or a population. It may be a single all-encompassing measure or perhaps a weighted combination of the health measures described above.

The question of whether it is appropriate to treat health as a capital asset it important. Doubts about treating health as a form of capital arise when one compares health to other forms of capital and notes the obvious differences. Economists generally describe capital as an input into a production function. We think of manufactured capital assets such as machines, equipment, buildings, roads, and ports that are used in the production of goods and services. Manufactured capital assets have value that is equivalent to their future marginal productivity. The productive services can be rented or the capital asset itself can be sold to another individual without destroying its value. Unlike inputs that are consumed as part of the production process, manufactured capital can be employed in the production process multiple times. Manufactured capital may depreciate over time, but it is not consumed in the production of goods and services. To summarize, economists generally think of a manufactured capital asset as (1) a durable object that could be sold to someone else, (2) an input in the production of goods and services, and (3) a store of value to achieve consumption.



## FIGURE 1 Life expectancy at birth

Source: UN Department of Economic and Social Affairs (2012) "World Population Prospects"



Like manufactured capital, health is durable. A person's health is relatively constant over time. Health depreciates, but it is not consumed as it provides current well-being. At times, health may depreciate rapidly due to some illness, similar to the risk of some catastrophe reducing the value of a manufactured capital asset. Unlike a manufactured capital asset, health capital cannot be directly purchased from a health-rich person. One cannot rent the well-being services that flow from health nor can one sell health to another individual. However, the ability to transfer a capital asset to another individual does not seem to be an essential characteristic of capital. The knowledge, skills, and abilities that make up what is commonly called "human capital" cannot be directly transferred from one person to another and this does not cause economists to question if human capital can be considered a capital asset.

Health is not commonly thought of as an input in the production of final goods and services. The evidence suggests that improvements in health do lead to productivity gains, particularly in low-income countries (BHARGAVA ET AL. 2001). The estimates suggest that large increases in health cause only small increases in GDP growth rates and there is little evidence for productivity gains from health in developed countries. However, health does provide health services - greater enjoyment of current consumption and longer life - directly to the individual. That these health services are not part of measured gross domestic product does not mean that they have no value. To the contrary, health is of great value to humans and is an essential characteristic of well-being. Our view is that health capital is similar to consumer durables (e.g., houses, consumer electronics, furniture, home appliances, and sports equipment) that provide well-being to consumers, but are



not generally direct inputs in the production of a final good or service that is counted as part of gross domestic product.

Figure 2 illustrates the point that capital assets can both directly and indirectly (through the production process) affect human well-being. Machines and business equipment primarily increase human well-being through the production process. Forests have both a direct influence on human well-being through ecological and recreational services as well as an indirect influence through the consumption of final goods for which timber is an input. Similarly, health capital has a direct influence on human well-being as well as an indirect affect through increased productivity. That health increases human wellbeing primarily through a direct channel rather than through the indirect production/consumption channel does not raise any concerns about its treatment as a capital asset.

To summarize, health is (I) durable and nontransferable, (2) both an input in the production of goods and services and the source of a flow of services which increase human well-being, and (3) a store of value to achieve the consumption of health services. Services from capital assets, which are not counted as part of GDP, including health services, have value. Therefore, the value of the capital asset itself is equal to the present discounted value of the future services. From the point of view of an economist, health is a form of capital. This chapter seeks to measure the stock of health capital, estimate its value, and measure the value of the change in health capital for several countries over a period of five years.

#### 3. The value of health capital

We begin with a stylized model to illustrate the role of health capital in providing human wellbeing. We propose a rather simple two-period model as it is sufficient for providing general intuition about how to measure and value health capital. The model is an expanded version of the Arrow et al. (2013) model. We assume that the economic agent is alive in period I with certainty, but there is uncertainty about being alive in period 2. The agent's expected lifetime utility is given by

EQUATION 1

Expected Lifetime Utility =

$$U(H,c_1) + \pi(H) U(H,c_2)$$

where *H* is health capital,  $c_{T}$  is consumption in period  $I, c_1$  is consumption in period 2, and  $\pi(H)$  is the probability of survival to the second period. We assume that the probability of survival depends on the amount of health capital where more health increases the probability of being alive in period 2, though diminishing returns imply that each additional unit of health capital has less of a positive effect on the probability of survival. The current utility (felicity) depends both on the level of health capital and the amount of consumption. There are diminishing marginal returns to both health capital and consumption. The utility function is the same in both periods and for simplicity we assume that the agent does not discount the future, though this could easily be relaxed.

The agent is endowed with financial wealth given by W(H). We assume that an increase in health causes an increase in the agent's wealth. The mechanism we have in mind is that a healthier agent is more productive and earns higher wages. Alternatively, we could assume that a healthier agent is able to work more hours and thus has a higher income. However, to keep the model focused on health, we abstract from the labor-leisure decision and simply assume that the agent is directly endowed with wealth. Making wealth a function of health capital embeds the productive impacts of health in a straight-forward way. By assumption, there are diminishing returns to additional health capital. The increase in the agent's wealth caused by an additional unit of health capital is much smaller for a healthy agent than for a malnourished one.

The agent's lifetime budget constraint is given by

EQUATION 2

$$c_1 + pc_2 + h \le W(H)$$

where wealth can be spent on either consumption in period 1, consumption in period 2, or investing in health. In our notation, an investment in health is given by *h* and we assume that health capital, *H*, is increasing in *h*. We have normalized the price of consumption in period 1 to one. Survival to period 2 is uncertain, so the agent can purchase a contract granting consumption in period 2 at price *p* which is less than one. If the probability of survival is very low, the contingent price for consumption in period 2 would also be low. We will assume that the agent can purchase period 2 consumption at the actuarially-fair price of  $p = \pi$ .

We should consider the difference between an investment in health, *h*, and consumption, *c*. Purchasing a pain reliever, like Aspirin, should be treated as consumption. A short-term pain reliever provides a health service, but it has no effect on health capital, H, because the effect is temporary. The resulting reduction in pain increases current well-being, but the effect does not carry over into other periods. In this model, the primary characteristic of an investment in health is that it increases the stock of health capital. Therefore, many healthcare services and medicines would be categorized as consumption rather than health investment. A true investment in health would increase future health services.

The agent wants to maximize expected utility given by Equation (I) subject to the budget constraint given by Equation (2). With no discounting, the agent will choose to perfectly smooth consumption by selecting  $c_1 = c_2$ . Thus we can rewrite Equation (I) as

EQUATION 3

Expected Lifetime Utility =  $U(H(h),c) + \pi(H(h))U(H(h),c)$  where consumption,  $c = c_r = c_2$ , is the same in both periods and health capital, H(h), is written as a function of health investment *h*. The first order condition with respect to health investment is given by:

EQUATION 4

$$\underbrace{\left(1+\pi\right)}_{\text{Direct Wellbeing}} \frac{\partial U\left(H,c\right)}{\partial H} \frac{\partial H}{\partial h} + \underbrace{\frac{\partial U(H,c)}{\partial c}}_{\text{Productivity}} \frac{\partial W}{\partial H} \frac{\partial H}{\partial h}}_{\text{Productivity}}$$

$$\underbrace{\frac{U(H,c)\frac{\partial \pi}{\partial H}\frac{\partial H}{\partial h}}{\text{Longevity}}}_{\text{Longevity}} = \frac{\partial U(H,c)}{\partial c}$$

Equation (4) illustrates the trade-off between using wealth for consumption or health. The additional utility from an increase in consumption is given by the right-hand side of Equation (4). The expected-utility-maximizing agent invests in additional health up to the point where, at the margin, the value of additional health is equal to the marginal value of consumption. The value of additional health has three components: well-being, productivity, and longevity as given by the three terms on the left-hand side of Equation (4). We will examine each of these three components in greater detail.

#### 3.1 Direct well-being

The first term on the left-hand side of Equation (4) is the direct utility from additional health capital. Consider a health investment that offers no increase in productivity and no increase in longevity. For example, a surgical procedure that offers long-lasting pain reduction, but does not improve the agent's ability to work nor does it offer any reduction in the risk of mortality. This hypothetical surgery's only effect is to permanently reduce the agent's chronic pain. The reduction in pain directly makes the agent better off and may also increase the agent's enjoyment of consumption. Because the pain reduction is long-lasting, the surgery is an investment which

increases health capital. Given our assumption that the surgery has no impact on the agent's productivity or longevity, the entire marginal benefit of this investment in health is captured by the direct improvement in well-being given by the first term of Equation (4):

 $(1+\pi)\frac{\partial U(H,c)}{\partial H}\frac{\partial H}{\partial h}$ 

The term  $(I + \pi)$  above represents the lasting impact of the investment in health as the increase in utility occurs in both period I and period 2. Surviving to period 2 is uncertain, so the increase in expected utility reflects that the agent will only be alive for period 2 with probability  $\pi$ . The rest of this expression is the additional utility or current well-being that the agent enjoys because the level of health capital is higher. It is of particular interest to note that the demand for health investment, h, will be larger if the probability of survival to period 2,  $\pi$ , is larger. The intuitive explanation is that longlasting medical intervention that improves well-being is more valuable to those who expect to live longer. Holding other things constant, as mortality rates decline, the demand for medical services that offers only short-term improvements in well-being will also decline as individuals substitute towards medical services that offer long-term improvements.

We are not familiar with any empirical estimates of the consumption equivalent value of this direct increase in well-being. One approach would be to estimate the willingness to pay for a medical service similar to the hypothetical surgery discussed above. The key would be to identify treatments that have no effect on either productivity or longevity. Calculating the willingness to pay for any health intervention which also affects these other two components would produce upwardly biased estimates. The second term on the left-hand side of Equation (4) is the productivity gains from additional health capital:

EQUATION 6

$$\frac{\partial U(H,c)}{\partial c}\frac{\partial W}{\partial H}\frac{\partial H}{\partial h}$$

The health investment increases the stock of health capital which increases the agent's wealth. The agent values the additional wealth because it can be spent on additional consumption, c, the marginal value of which is given by the first term in the above expression. The assumption that wealth increases as health capital increases, with diminishing returns, is a simple way to represent the increase in productivity from health.

There is a strong correlation between income and health (see FOGEL 1994). We generally assume that the causal relationship is that additional income allows an individual to make health investments which increase health. However, there is strong evidence for a causal relationship running the other direction. An increase in health causes higher labor productivity through fewer lost workdays, greater physical energy at work, and greater mental focus and ability.<sup>1</sup>

Leibenstein (1957) first proposed that workers with low levels of calorie intake would have lower productivity. There is strong evidence that improvements in nutrition lead to productivity gains in agriculture for those with low levels of calorie intake (STRAUSS 1986). Similarly, there is strong evidence that an increase in birthweight, reflecting an increase in intrauterine nutrient intake, causes an increase in future labor market income (BEHRMAN AND ROSENZWEIG 2004). The evidence shows only productivity gains for low calorie intake. Thomas and Strauss (1997) find evidence only for a positive impact of additional calories below 2,000 calories per day.

Evidence for a causal effect of health on productivity in developed countries is weaker. Several papers have shown that increases in average life expectancy in a country lead to increases in GDP growth (see BLOOM, CANNING, AND SEVILLA 2004). However, we are not aware of convincing micro evidence in developed countries that worker productivity is increasing in health. The evidence that workplace wellness programs increase productivity is mixed. While these programs generally increase worker health as measured by increased physical activity, reduction in tobacco use, and decreased body mass index, there is little evidence of productivity gains (OSILLA ET AL. 2012). The strongest evidence seems to be that an exogenous increase in worker health reduces absenteeism (BAICKER ET AL. 2010).

#### 3.3 Longevity

The defining characteristic of this simple model is that life expectancy is not fixed. We assume that an investment in health increases the probability of survival to the second period. It is clear from Equation (1) that an exogenous increase in the probability of survival,  $\pi$ , increases the expected lifetime utility. How much does the agent value an increase in the probability of survival to the second period? Note that the rate of increase in the expected lifetime utility when  $\pi$  is increased marginally is U(H,c). This means that the value of the increase in the probability of survival depends on the living standard of the agent. An agent with better health and more consumption will place a higher value on an increase in the probability of survival.

The same point is clear from the third term on the left-hand side of Equation (4):

<sup>1</sup> Jayachandran and Lleras–Muney (2013) point out that health indirectly affects productivity through education. As longevity increases, so does the return on investments in education. This encourages additional education which makes workers more productive. We do not consider this indirect relationship because improvements in education (even if motivated by increased longevity) are already included in inclusive wealth via the measurement of human capital.

#### TABLE 2

Estimated value of the average annual increase in life expectancy in the U.S.

| Time period | Value of increase   |
|-------------|---|
| 1975 – 2000 | US\$52,000  |
| 1960 – 2000 | US\$2,000   |
| 1970 – 2000 | US\$40,000  |
| 1980 – 2000 | US\$60,000  |
| 2000 – 2005 | US\$11,400  |
| 1990 – 2008 | US\$7,000   |
|             | Time period   1975 – 2000   1960 – 2000   1970 – 2000   1980 – 2000   2000 – 2005   1990 – 2008 |

#### EQUATION 7

# $U(H,c)\frac{\partial \pi}{\partial H}\frac{\partial H}{\partial h}$

The marginal value of an increase in  $\pi$  depends on the utility the agent would realize in the second period. To express U(H,c) in dollar terms, economists divide by the marginal utility of consumption and call the resulting object the value of a statistical life or VSL (ASHENFELTER 2006). It is important to note that economists do not claim that VSL is the value of life. Instead, one should think of VSL as the amount people would be willing to collectively spend in order to reduce the number of expected deaths by I.

The value of a statistical life can be inferred from individual choices. For example, workers who wash the windows of skyscrapers face a higher risk of death and are paid more than workers who wash the windows of single-story businesses. The additional compensation from assuming the additional risk of death reflects the workers' willingness to pay to reduce mortality risk. There are many similar opportunities for economists to observe a group's willingness to pay for a reduction in mortality risk. In a survey of country-level VSL estimates, Viscusi and Aldy (2003) found that VSL is approximately equal to US\$12,000 multiplied by GDP per capita raised to the 0.6 power. This implies a 2014 VSL of US\$8.3 million in the United States, US\$2.5 million in China, and US\$315,000 in Malawi.

The large VSL estimates imply that investments in health that result in even a small reduction in mortality risk have great value. Many studies have estimated the value of the increase in life expectancy in the U.S. including Nordhaus (2005), Becker et al. (2005), Murphy and Topel (2006), Jones and Klenow (2011), Arrow et al. (2012), and the Inclusive Wealth Report 2012. Table 2 reports the estimated value of the increase in life expectancy from each study. The time frame for each study is different, so we report the total estimated increase in value divided by the number of years. This results in an average increase in value per year. Differences across studies therefore are due both to differences in methods but also in the time period studied.

The estimate of the value of the average annual increase in life expectancy by Becker et al. (2005) is far lower than the other estimates. The time period they considered did not have smaller gains in life expectancy, so the difference comes from their methods. Their model implies a value of a statistical life from the other parameters of the model. Rather than going to the literature for a VSL estimate, they calibrate the other parameters of the model and the resulting VSL is fairly small. The Nordhaus (2005), Murphy and Topel (2006), and Jones and Klenow (2011) studies use similar methods and find large estimates. The Inclusive Wealth Report 2012 follows the Arrow et al. (2012) methods. Both consider a more recent time period and find similar results.

The Arrow et al. (2012) approach is to calculate the expected discounted years of life remaining for each age- and gender-specific group in the population in each year. The population-weighted average of the group-specific changes in the expected discounted year of life remaining over the period is then multiplied by the value of a statistical life year (VSLY) for that country. The VSLY is the VSL divided by the expected discounted years of life remaining and thus represents a per-year valuation of the reduction in risk of mortality.<sup>2</sup> Note that population aging mechanically decreases the average expected discounted years of life remaining, but that the increase in longevity for the old have outpaced this mechanical decrease (see Appendix I for the details of this method). The Arrow et al. (2012) approach is straight-forward requiring only life expectancy data combined with an estimate of the VSL for the country. It makes no attempt to adjust the VSL for age or cohort effects as in Aldy and Viscusi (2008). In the model presented here, an agent with higher wealth should have a higher VSL. This is consistent with the cross-country VSL estimates. However, this also suggests that where the life expectancy increases occur within the wealth distribution within a country should matter in calculating the value of the improvement in health. Our method does not account for this.

Hamilton (2012) suggests that the Arrow et al. (2012) estimates are implausibly large and

claims that this must be due to double counting. He correctly points out that VSL reflects the value of all good things that come with living, not only good health. He then argues that if we have already measured the value of natural, manufactured, and human capital, wouldn't it be double counting to then add the value of health capital as the VSL depends on the living standards which are themselves a function of the other forms of capital? This concern is understandable but incorrect. Living is complementary with consumption. Even if health offered no direct increase in well-being or an increase in productivity, health would still have value simply because it extends life and allows a person to enjoy living longer. The model presented here illustrates the important point that the value of health capital does and should depend on the level of well-being and thus by extension on the levels of the other forms of capital and on the state of technology. Note that the term U(H,c) appears in the expression for the value of increased longevity from an increase in health. That utility term represents the value of all good things that come with living.

#### 3.4 Combined value of health capital

The value of an increase in the stock of health capital is the summation of all three components: direct well-being, productivity, and longevity. Accepted empirical estimates of the first two components are lacking, so economists have relied on the third alone to estimate the value of health capital. This is fine, but it should be recognized that the resulting estimates are biased downward. The large value of health capital reported in this chapter is probably too small rather than too large.

An additional issue is how to measure health capital itself. Recall that there are various measures of health and not clear theoretical justification for selecting any particular measure. Restricting the value of health capital to the longevity component makes it easy to justify

<sup>2</sup> There are alternative methods for calculating the value of a statistical life year (VSLY). Our method implies a constant VSLY for all individuals within a country but allows for differences across countries.

using expected discounted remaining years of life expectancy as the measure of the stock of health capital. However, assume that we wish to add the productivity component to the value of health capital. Also assume that we have a convincing estimate of the effect of some measure of health, say BMI, on productivity. With this estimated effect we can value the productivity gains or losses from a change in BMI. However, it would be incorrect to refer to this as the value of the change in BMI as this would only be the productivity component. It would also be incorrect to calculate the effect of a change in BMI on longevity and then use this to supplement our longevity valuation. That would be double counting.

The Inclusive Wealth Report 2012 and Arrow et al. (2012) decision to include total factor productivity (TFP) as a measure of technological progress introduces an issue here. TFP growth is included as growth in an additional form of technological capital or time capital. Suppose that a change in BMI causes an increase in productivity. This would be picked up in the TFP growth measure as coming from technological change, when in reality it was the result of an improvement in health. This suggests that it may be appropriate to exclude the productivity component from our valuation of health capital if TFP growth is included in the measure of inclusive wealth. The health effect on productivity should already be captured by the change in TFP.

#### 4. Conclusion

Measuring health capital using only data on life expectancy seems appropriate given the measurement challenges and lack of empirical estimates for the direct welfare and productivity components of the value of health capital. That the value of the change in health capital is large is not surprising given the large willingness to pay for reductions in mortality rates. Health capital should no longer be relegated to the appendix when measuring inclusive wealth. It is likely the most important form of capital in producing human well-being. Health services primarily affect human well-being directly rather than passing through the production process to generate goods and services.

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# **APPENDIX 1:**

# Methodology for valuing the longevity component of health

This appendix provides a description of the methodology employed in Arrow et al. (2012) as well as the Inclusive Wealth Report 2012.

Let f(t) be the density of age of death, F(t) the cumulative distribution of age of death, and  $f(t|t \ge a)$  the conditional density of age of death given survival to age a. We obtain an estimate of the number of people who survive to age t out of a starting cohort of 100,000 (column l(x) of life tables) for each year and each country. From this, we calculate the unconditional number of deaths by age and divide by 100,000 to give f(t), the density of age of death. From  $f(t) = f(t|t \ge o)$  we calculate F(t):

EQUATION 1

$$F(t) = \sum_{a=0}^{t} f(a)$$

+

The conditional density of age of death is obtained from f(t) and F(t):

EQUATION 2

$$f(t | t \ge a) = [1 - F(a)]^{-1} f(t)$$

Future years are discounted at a constant rate d, assuming that the value of an additional year is independent of age. The health capital of an individual of age a is the discounted expected year of life remaining:

EQUATION 3

$$H(a) = \sum_{t=a}^{100} f(t \mid t \ge a) V(a,t)$$

where V(a,t) is given by:

$$V(a,t) = \sum_{u=0}^{t-a} \left(1-\delta\right)^{u}$$

The total health capital of all individuals of age a in a country is calculated by multiplying

H(a) by the number of people of age a in that country, P(a). Thus, the total health capital (measured in discounted life-years) of a country is:

$$H = \sum_{a=0}^{100} H(a) P(a)$$

The value of a unit of health capital is the value of a statistical life-year or VSLY. Thus, the value of the total stock of health capital is simply *H* multiplied by the VSLY.

# Part III: New insights

# Forest wealth of nations

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# **KEY MESSAGES**

**Forest ecosystems provide a huge** range of tangible and intangible benefits for human well-being. These are of immense value and represent an important component of national and global wealth.

**Demographic trends and economic growth** are exerting increasing pressure on forest capital. Accounting more fully for this wealth, and how it is changing as a result of economic and social activity, is urgently required. The estimates in this chapter provide a tentative first step in this direction.

**From a global perspective**, in 2010 for the selected countries, forest wealth amounted to more than US\$273 trillion. On the face of it

this wealth, in absolute terms, seems concentrated in relatively few countries. However, for many other countries, forest capital remains an important component of national wealth. Many of these countries (although not all) have experienced alarming losses in forest capital over the past 20 years.

**From an accounting perspective, these** losses are frequently hidden from view. It is thus essential that nations pursue better accounting to understand quantity, quality, and distribution of forest wealth. Indeed, keeping forest wealth intact – and, moreover, investing in forests to reverse past losses – is an important pre-condition for sustaining development.

#### 1. Introduction

Natural capital is widely recognized as an important component of the total wealth of a nation. Correspondingly, there have been considerable advancements in practical accounting for this natural wealth (see, for example, UNU-IHDP AND UNEP 2012, UN 2012A, WORLD BANK 2010, ONS 2014). This progress, however, has been incomplete. While a substantial amount of this work has focused on the non-renewable wealth within subsoil assets, far less remains known about renewable wealth, such as that within forest capital. There are exceptions (see, for example, BARBIER 2013, HAMILTON AND ATKINSON 2006); however, these are largely restricted to understanding forests as a store of timber value, for instance. Of course, this is only one such "ecosystem service" that forests provide. Reconciling practical efforts to account more comprehensively for total wealth is an important next step if wealth accounting is to remain relevant and useful.

In this chapter, we take a preliminary step towards incorporating some of these broader values of forest capital within a measure of inclusive wealth. Our approach is a deliberately rudimentary one: We define the value of the natural asset only in terms of the sum of the benefits provided by the stock of a unit, in this case land area of standing forest. While this gives a rough first approximation of the value of this wealth, it is important to recognize that it is an incomplete description of the value of the natural capital comprising this asset (see, for example, NCC 2014A). Yet it is a practical starting point, making use of emerging databases that have sought to synthesize the empirical record on the value of ecosystem services, and consistent with the land-based physical accounting units in existing UN efforts (UN 2012B).

While the results should be viewed as preliminary and partial, we nonetheless argue that these findings are illustrative and contribute to the process of understanding how natural capital contributes to total wealth. From a global perspective, in 2010 this forest wealth amounts to more than US\$273 trillion (2005), which is 56 times the aggregate global GDP. On the face of it, this wealth, in absolute terms, seems concentrated in a relatively few countries. However, for many other countries, forest capital remains an important component of national wealth. Many (though not all) of these countries have experienced alarming losses in forest capital over the past 20 years. While this evidence requires further scrutiny than is possible here, it does strike at the heart of critical debates about the sustainability of development, and whether other components of wealth truly compensate for such losses in their entirety.

The remainder of this chapter is organized as follows: Section 2 reviews and discusses a number of issues relating to accounting for forest capital; Section 3 outlines the methods and data used; Section 4 sets out our results. Section 5 offers conclusions considering the preliminary nature of these results, and outlines some key caveats to our interpretation of the findings in the chapter.

#### 2. Accounting for forest wealth

Forest capital produces a multiplicity of intermediate and final goods. These contributions are often overlooked, however, because they are largely "invisible". Not surprisingly, this has fed the critique of existing national income accounting systems as regards their inability to explicitly measure changes either in the productive capacity of forest resources, or the contribution of forest resources to human wellbeing. These include: (1) the contribution of forest capital as an input into production processes; (2) the broader values that society places on forest capital; (3) the benefits and services that forest capital provides beyond the production boundary and markets; (4) the impacts of current economic activity on forest capital; and, (5) changes in human well-being as a result of changes in forest capital.

Evolving work such as the UN's environmental-economy accounting framework (UN 2013A, 2013 B - HENCEFORTH "UN") show that progress is being made on addressing these issues. However, it is clear that there remains much to do. In the first Inclusive Wealth Report (UNU-IHDP AND UNEP 2012, henceforth referred to as IWR 2012) the estimate of forest capital was based on valuing timber, non-timber forest products (NTFP), and carbon. In this chapter, we extend this earlier work by including a range of other ecosystem services. In doing so, we are faced with two broad challenges. The first is the description of the physical accounting units. The second is the valuing of these physical dimensions. The link between the two is in identifying both the flows of goods and services from these physical assets and the way in which these flows translate into benefits that people value.

Forest capital facilitates this link through complex biophysical structures and processes. This complicates any judgement about what the basic physical accounting units should be. In principle, this should be focus on the capital asset itself. However, in the current context, this is far from straightforward. Natural assets, through ecological production (and perhaps a mixture of human management and inputs from other assets), provide ecosystem goods and services which ultimately give rise to benefits of value to human populations. Importantly, the natural capital giving rise to such services is likely to be an assemblage of different assets including not only, say, land under forest cover, but also soils and associated natural processes and functions.

Thus, the exact contribution of forests ultimately to human well-being depends on many factors which can be construed as assets. These include ecological factors, such as the functional and process integrity of forest ecosystems. The contribution also depends on social arrangements, such as institutions and technologies. And while underlying natural functions and processes may remain the same, how the forest is valued by people can change over time. This in turn may have implications for natural processes, depending on how values change the way in which a resource is managed.

In this respect, the units of forest land area, our basic accounting unit for this chapter, is a limitation. That is, it is unlikely to satisfactorily reflect in full this combinatorial character of the natural capital in forests. Nevertheless, it remains a useful starting point for any practical response to challenge natural capital accounting (NCC 2014B). This emphasis, for much the same reason, is also the recommendation of the UN (2013), which proposes a set of major landuse categories as the basic accounting units for constructing ecosystem accounts. From this practical perspective, concerns about the fundamentals of natural capital might be captured, to some extent, by looking at the biophysical properties of these natural areas such as soil, biodiversity and so on (HAMILTON 2014). The Natural Capital Committee (NCC) (NCC 2014B) look at not only the extent or quantity of land area such as forest land, but also at its quality and configuration.

Robust accounting, therefore, is likely to require a detailed micro-investigation of land areas. By contrast, the approach we take is by its nature broad, and highly aggregated. As such, it should be seen as the start of a process to more accurately account for the value of forest wealth, rather than the end.

Our approach combines information about land area with the growing evidence base for establishing the value of benefits that are provided by ecosystem services held within these land units. For example, the pioneering Millennium Ecosystem Assessment (MA 2005) has categorized various important benefits provided by forests, including provisioning services (food, fuel, fibre, water, etc.), regulating, supporting, and cultural services. These have also been documented by other major initiatives, including The Economics of Ecosystems and Biodiversity (TEEB) and Wealth Accounting for Valuation of Ecosystem Services (WAVES). Forest vegetation and soils capture and store atmospheric carbon dioxide (CO2) and therefore play a significant role in the global carbon cycle

and have the potential to avert climate change. In the Reduced Emissions for Deforestation and Degradation (REDD), carbon storage is viewed as a stock variable while the carbon sequestration during growth and regrowth is treated as a flow variable.

Forests play a very important role in regulating stream flows, supplying fresh water by filtering pollutants, stabilizing soils, reducing erosion and sedimentation, and moderating impacts of extreme weather events and related hazards, such as floods, storms, and landslides. Forests improve air quality, and thereby human health, and provide a habitat for various flora and fauna – indeed they are crucial for maintaining biodiversity. Exact roles depend on the forest type – for instance, canopy levels and vareity. Currently, none of these regulatory and supporting contributions are recorded in the statistical system of forest accounting.

Forest ecosystems often have very high social and cultural values in terms of recreation, aesthetic, and spiritual worth. Forests often contribute to tourism, although these contributions are recorded as activities within the tourism sector, and hence effectively lost to forestry calculations. Some communities, such as the Bishnoi communities in India, attach high spiritual values to forests, and in others forests hold religious significance. In addition to all the above, forests act as safety net in times of crisis, providing some earning opportunities and subsistence consumption possibilities. It should thus be clear that forest wealth measures must be expanded to include the important role that forests play in providing regulating, supporting, and cultural services.

This includes the diversity in structure, composition, productivity, and age of the species. In case of forests managed for the purpose of timber (i.e., production forests), the decision to harvest a particular species is based on the rotation age, which varies with the species composition. If removals are lower than, or equal to, additions, the forests are sustainable. The removals can be carried out via clear-felling trees on large tracts of land or through selective harvesting of mature trees. In case of clear-felling of trees, as the land officially remains under forests, the forest stock does not decline; however, if forests are converted to non-forest purposes after clear-felling, the wealth in the converted sector increases and the forest wealth declines. If the quality of the forest deteriorates, the area under forests remains intact but the stock of timber, carbon sink, biodiversity, and ecological functioning may vary.

In addition, forests derive value from the land value, as well as its output values, which is dependent on the purpose for which land is managed, ownership, and other institutional factors. For example, plantations are managed for timber and non-timber forest products, while national parks or sanctuaries are managed for the purpose of protecting biodiversity and maintaining ecological functions. We have assumed in this chapter that the forests are managed for provisioning, supporting, regulating, and cultural values, and the extent of area set aside by each country differs depending on country-specific priorities and socio-economic conditions. A country rich in biodiversity may allocate a higher proportion of forest land for conservation and vice versa. A global resources assessment of the Food and Agriculture Organization of the United Nations (FAO 2010) classifies the forest land by management purpose - production, protection, multiple-use, or conservation (i.e., forest in national parks, nature reserves, and other protected areas).

In addition, these ecosystem benefits and services are often dependent on each other, hence simply aggregating values is impossible. For example, carbon sequestration depends on biomass which in turn depends on the soil quality and moisture levels. Similarly, the hydrological value of forests depends on the height of the vegetation, the interception of soils, and the evapo-transpiration potential of the leaves. Ideally, they should be modeled together. In the absence of detailed information, we assume that these values are additive and the value of forests is the sum of other direct and indirect benefits provided.

The key to wealth accounting, as opposed to more mainstream accounting in the physical realm, lies in capturing shadow prices. Shadow prices are the marginal values obtained by equating both the demand and supply side. In well-functioning markets, market prices are the best approximation of shadow prices. However, as we discussed, forests provide many ecosystem services which are well outside the production boundary and for which market prices do not exist. Because of the public good nature of these goods it is difficult, if not impossible, to correct market failures associated with forests. This difficulty in assigning prices leads to policy failures. To remedy this, several non-market valuation techniques have been developed to value the ecosystem services provided by forests (see mäler and vincent 2005).

To value the asset we need to identify the descriptors of forests, which can be different depending on the forest service valued. For estimating timber values, growing stock at the beginning and end of the assessment period can be a better descriptor than the area. The shadow prices of the standing stock of timber depend on the distribution of age class in each time period, and how they change as this distribution is altered. Here, the shadow price has been estimated as the weighted net price of harvested timber (weighted price of harvested timber minus the harvesting costs), and the value of timber stock is growing stock multiplied by shadow price.

Along with timber, forests sequester carbon in forestland, as well as harvested products. The harvested timber moves out of the production boundary of forests and enters the forest product boundary. However, carbon is also embodied in timber and timber products. If the timber is used in making furniture, the carbon in it remains sequestered for a long period. However, if the timber is used in making certain short-lived products like paper, fuel wood, etc., the carbon is soon released through burning or processing into the atmosphere. In the case of production forests, the carbon moves out of the forests in the form of harvested wood but is ultimately sequestered when forests are regenerated. For protected forest areas, carbon fixation is permanent, as there are no anthropogenic disturbances and thus positive net carbon sequestration. Although old growth forests were once thought to be carbon neutral, a study published in Nature (LUYSSAERT ET AL. 2008) showed that the net ecosystem productivity of forests between ages 15 and 800 is usually positive, depending on climatic factors and the extent of nitrogen deposition. Various disturbances on forests, such as forest fires, pests, diseases, logging, conversion to other land uses, etc., can also release carbon into the atmosphere. The exact amount released depends on the extent and nature of the disturbance. The value of the forest carbon is the net carbon sequestered multiplied by the marginal social damage due to carbon emissions within its the geographical boundary.

Forests provide valuable non-timber forest products of extreme significance to specific communities, particularly in developing countries. Even in developed countries such as Sweden, forests provide significant benefits from the provision of non-timber forest products (NTFPs). In cases such as this, forest area can be a better descriptor, and as such the value of forests due to NTFPs is the present value of NTFPs (value of harvests minus cost of collection) multiplied by the area accessed by local communities.

The indirect use values provided by forests come primarily in the form of regulatory and supporting services, such as biodiversity maintenance, nutrient recycling, maintenance of hydrological cycles, watershed protection, waste assimilation, prevention of soil erosion, pollination services for agriculture, micro-climate stabilization, and so on. However, these values are very difficult to measure separately as they are often only provided as a system. At this point in time, the knowledge on these systems is limited and these values have generally been measured independently by researchers (compiled in the TEEB database, see DE GROOT ET AL. 2012). The best stock descriptors in this case would be the area underneath primary forest cover. Thus, the ecological value of the forest is the net present value of these services multiplied by the area under primary forests.

Forest ecosystems provide valuable recreational services, which are usually attributed to the tourism sector, with the exception of revenues generated directly by forest managers. There are costs associated with maintaining forests for such uses, which should be deducted from the recreational value. However, increasing forest area may not necessarily increase the recreational value by the same proportion. The recreational value of the forests is the present value of net benefit provided by recreation multiplied by the area providing recreational services. It should be noted that there are scarcity values for ecosystem services, and the forest ecosystem value can rise as the area decreases and ecosystems subject to non-linearity. One additional consideration is that some of these values of forest wealth are likely to be capitalized in other forms of wealth. For example, any pollination services or water provision ultimately attributable to the existence of forests boosts the value of agricultural land. Water regulation (perhaps via flood protection) might be reflected in the value of the produced capital protected by such services. Therefore, caution must also be applied, or at the very least caveats noted, in adding the value of forest capital to other components of total wealth.

Accounting for forest wealth clearly raises considerable issues. Interpreting the resulting accounting aggregates should also provide pause for reflection. On the one hand, if the sustainability of development is reckoned to be dependent only on changes in total wealth (per capita) over time, then what is happening to forest wealth should be viewed primarily in the context of what is happening to wealth overall. Broadly, this is the approach taken in IWR 2012. On the other hand, contributions in that same volume also emphasize the characteristics of natural capital which indicate that simply looking at overall trends in wealth is insufficient to determine whether or not development is sustainable. For example, if, as seems likely, the benefits provided by natural assets such as forests are non-substitutable, then any sustainability criterion based on wealth accounts should also be concerned about what is happening to these assets. Evidence of declining forest wealth should also prompt questions about whether these losses should be diminished or even reversed if, say, the decline in forest area has exceeded a certain threshold (see, for example, PEARCE ET AL. 1996).

#### 3. Methods and data

We take the land area officially designated as forest area as a (proximate) physical accounting unit for forest assets. We used the FAO (2010) definition of forestland "which refers to land spanning more than 0.5 ha with trees higher than 5 metres and a tree canopy cover of more than 10 percent and characterized by the absence of other predominant land uses." This particular classification includes both natural forests as well as plantations classified by management purpose - production, protection, multiple-use or conservation (i.e., forest in national parks, nature reserves, and other protected areas), as well as forest stands on agricultural lands like wind breaks, shelter belts, abandoned areas of shifting cultivation, and corridors of trees. However, this definition excludes trees established with the primary purpose of agricultural production, horticultural and agroforestry systems. This approach is broadly consistent with the United Nations framework for the System of Environmental and Economic Accounting Framework (SEEA) (UN 2013A-B) as well as, more closely, the aforementioned definition in FAO (2010).1

<sup>1</sup> The SEEA further classifies forestland into primary forest, other naturally regenerated forest/planted forests, and timber resources, by whether these are cultivated and natural timber. Proper accounting treatment needs to be accorded depending on whether timber comes from planted forest or primary forest and naturally regenerated forests. While the distinctions are important for certain countries but unimportant for

To understand the changes in the value of forest capital stock, we need measurable physical stock descriptors. This could be the area under forest, the growing stock of standing timber that forests contain, the total biomass, net carbon sequestered and stored, the biodiversity the forestland holds, and so on. In fact, these descriptors are interrelated. If the measurable descriptor valued at constant shadow price at the beginning and end of the period remains constant or increasing, we treat this as investment, otherwise it is disinvestment.

Inclusive wealth is defined as the present discounted value of all capital assets, where these stocks are each valued in terms of their respective shadow prices at time "t". The value of forest capital thus at any given point of time is the present value of the future net benefits that can be expected over the life of the resources.

Formally,

EQUATION 1

$$\mathbf{V}(\mathbf{F}(\mathbf{t})) = \int_{t=1}^{T} e^{-r(t-n)} \left( VF_{db} + VF_{ndb} \right) \mathrm{dt}$$

Where:

V(F) = per hectare value of forests.

 $VF_{db}$  = The per hectare value of direct net benefits derived from forests.

 $VF_{ndb}$  = The per hectare value of indirect net benefits derived from forests.

T =life of forests.

*r* = the discount rate.

n = the time period of assessment.

The direct benefits derived from forests might be provisioning services and recreational services directly consumed by the individual or the society. Indirect benefits include regulating and supporting services that may benefit the individual or society indirectly. As the species are ecologically interdependent, the ecological values may be interdependent on each other.

In what follows, we analyze data on forest area, growing stock and forest carbon for the period of 1990 to 2010 for a subset of countries which we have divided into low, middle, and high-income nations. The latter classification is based on income in World Development Report (WDR 2014). Country-level data on export quantity and value of round-wood from the FAO Forest Resources Assessment (2010) interactive database was used to compute the average price per cubic metre of timber for the years 1990, 2000, 2005, and 2010 for the selected countries for which data is available. The unit values are adjusted for cost of harvesting and then multiplied by the growing stock in production forest to obtain the value of forest timber wealth. The costs of harvesting takes into account the forest management type, proximity to markets, volume felled, type of species, equipment, harvesting practices employed, etc., all of which vary widely. We used average timber harvesting values based on the Global Timber and Forestry Data Project (GTAP) database, which ranged from 30 percent to 60 percent of price per tonne of timber depending on the country, the type of forest and the management regime.

Because market prices change over time, and we are seeking to understand what proportion of stock value changes due to physical changes in quantity and quality of the asset, we used the average net price per cubic metre of timber over the period of 1990 to 2010 in order to keep price constant. In reality, if market prices change due to scarcity they should be accounted for, but market prices can change due to other factors as well. Only the area under production forest is valued for timber, and as consistent fuel wood consumption and export data is not available for several countries, we could not obtain the value of fuel wood. However, the fuel wood consumed by local communities is assumed to come from twigs, branches, etc. and is included under the non-timber forest products.

The role of forests in mitigating climate change and the potential temporal and spatial damage that one metric tonne of carbon released today would cause is well established. A number of models are available to project future economic damages likely to occur due to climate change and the payments to

other countries, we treat forests as one entity at this stage.

be made now to avoid this in the future (see INTERGOVERNMENTAL PANEL ON CLIMATE CHANGE 2014) for some examples). The social cost of carbon, the global cost to all future generations caused by one tonne of carbon in a given time period has been estimated in several studies in the literature using different integrated assessment models (IAMs) (NORDHAUS 2011, HOPE 2011, TOL AND ANTHOFF 2010).

These estimates of social cost of carbon (the incremental emissions), vary considerably. A recent study by Dietz and Stern (2014) suggested that in order to avoid warming of more than two degrees above pre-industrial levels (keeping the carbon dioxide emissions to a maximum of 425 to 500 ppm), a carbon price of US\$32 to US\$103 per tonne of CO2 (2012 prices) should be implemented in 2015, raising the price from US\$82 to US\$260 in 2035 depending on how great the damage from climate change is expected to be. However, another study by Howard (2014) in a report produced for the interagency working group on social cost of carbon, updated the U.S. social cost of carbon from a central value of US\$24 to US\$37 per tonne of CO2. Mendelsohn, Dinar, and Williamson (2006) argue that climate change will have serious distributional impacts across countries and is expected to affect poorer countries disproportionally due to differences in location and adaptation ability, wealth and technology. Hence, they argue for different social cost of carbon across countries. Nordhaus (2011) using the RICE (Regional Integrated Climate-Economy) model, estimated the social cost of carbon for different countries, which vary between 1.25 for Eurasia to 29.62 for Africa (2005 U.S. dollars for the year 2015) for a tonne of carbon in case of low discount rate regime (discount rate of I percent). It is clear there is an ongoing discussion on how to the estimate of marginal social cost of carbon.

As forest carbon is now traded on voluntary carbon markets, one could use the existing market price for carbon. Carbon is currently traded at around US\$8 per tonne of CO<sub>2</sub> in the European Union's carbon market, and in U.S. markets it is traded between US\$2 and US\$5 (HOPE 2014). However, carbon markets cover only a part of the total emissions, and were affected by the global economic downturn, in addition to politically-rooted market defects.

One could also look at the level of carbon tax imposed in various countries, using avoided tax as the value of carbon sequestered. However, in several countries carbon tax is levied with the intention of raising revenues rather than lowering carbon emissions, and so it is not appropriate to use tax rates. One could use a (hypothetical) global carbon tax, i.e., if countries were to internalize the social damages they cause because of carbon emissions, the potential benefit provided by forests could be used as one measure. According to Nordhaus (2011), in an optimized climate policy the social cost of carbon equals the carbon price or carbon tax.

We have used the social cost of carbon in this study, which is the additional cost caused by an additional tonne of carbon dioxide emissions. In our case, if forests sequester carbon, we have assumed that it is the avoided social costs. From a national accounting perspective, a country would account for the damages caused as a result of its emissions, rather than account for global damage. Thus, we estimated the total annual carbon sequestered and multiplied by the social costs given in Nordhaus (2011) for calculating the avoided damages by each country.

For estimating wealth due to other provisioning services such as NTFPs, supporting and regulating services, and recreational/tourism benefits of forests we use the TEEB data base (see DE GROOT ET AL. 2012). This is a departure from the earlier estimates of IWR (2012) which used the work of Lampietti and Dixon (1995). The study by Lampietti and Dixon (1995), estimated the benefits of NTFPs as US\$190 per hectare for developed countries and US\$145 per hectare for developing countries. These values were then multiplied with the forest area and the percent of forests accessible by the population (which was assumed to be 10 percent of the total forest area).

#### TABLE 1

Values of ecosystem services from the TEEB database considered for the study

| Ecosystem service            | High income<br>(US\$ /ha/yr in<br>2005 prices) | Lower middle<br>income<br>(US\$ /ha/yr in<br>2005 prices) | Upper middle<br>income<br>(US\$ /ha/yr in<br>2005 prices) | Low income<br>(US\$ /ha/yr) in<br>2005 prices |
|------------------------------|--|---|---|---|
| Non-timber forest products   | 0.23 (1)                                       | 4-61 (5)  | 0.3–1,558 (7)   | 13–238 (2)                                    |
| Water provisioning           | 6-97 (2)                                       | -   | 10–705 (2)  | -   |
| Capturing fine dust          | 10   | -   | -   | -   |
| Gas regulation               | 9 (1)  | -   | 69 (1)  | -   |
| Water regulation             | 2  | -   | 22-25 (2)   | -   |
| Erosion prevention           | 10   | 3-2,239 (9)   | 224 – 322 (2)   | 111   |
| Biological control           | 9  | -   | -   | -   |
| Nutrient recycling           | 5  | 19  | 19–21 (2)   | -   |
| Pollination                  | 5  | 46  | 13  | -   |
| Prevention of extreme events | 8  | 92  | -   | -   |
| Biodiversity protection      | -  | 0.07–493 (7)  | 4-394 (2)   | -   |
| Genetic resources            | -  | -   | 8–148 (3)   | -   |
| Water purification           | -  | -   | 179 (1)   | -   |
| Recreation                   | 3–7,770 (10)                                   | 1-872 (6)   | 1–1,846 (9)   | 1–197 (4)                                     |

Note: the values in parentheses indicate the number of studies using either the market-based approach or cost-based methods. The values indicated display the range of values given in the database.

Our estimations are based on the approximately 320 publications, 1350 data points from 300 case studies stored in the Ecosystem Services Value Database (DE GROOT ET AL. 2012). All the values were adjusted for purchasing power parity (PPP) expressed in 2005 US\$. From this, this study considers selected forest ecosystem values (provisioning, regulating, and supporting) for different services and biomes. We dropped those values from the database considered benefit transfer; the final sample consisted of 350 values for forest ecosystem services, which were then grouped based on income, population, and biome – low-income, middle-income, and high-income countries – and by low and high density population. The values differed widely due to differences in valuation techniques. To be closer to exchange values, this study used only direct market pricing and cost-based techniques. Table I summarizes the range of values considered in this study from the TEEB database.

The values as can be seen from Table I are not uniformly available across countries. Where available we used the actual country estimates and, if not, used transferred values. Commonly, value transfer or function transfer is used by researchers for such calculations, and values can differ widely. For example, Costanza et al. (2014) used function transfer to estimate the global change in the value of ecosystem services, and earlier Costanza (1997) used value transfer. In that case, function transfer gave an estimate eight times higher than the value transfer for wetlands (COSTANZA ET AL. 2014).

This study takes a departure from the existing studies in transferring values. It makes use of a three-dimensional metric based on income group, population density, and biome. For each of these groups, we looked at the mean value per hectare of provisioning, regulating, supporting, and recreational services for a particular forest type. For example, if a low- income country has a combination of temperate and tropical forests, we used the local income country values for the respective ecosystems. The timber and carbon values were based on the actual country values. For other provisioning services, such as non-timber forest products and fuel wood, we used actual values where available and data from a similar country in a similar ecological zone when not available. The value has been expressed in terms of the value added by forestry in that particular country (with year 2005 fixed as numeraire). The coefficient so obtained is multiplied by the forest area of the respective country for which the value is being transferred, and the country's forestry value added. It is assumed this approach takes into account structural differences among the countries.

Such an approach can be justified as the standard normal variate of the countries classified by different income groups and population densities were distributed within one standard deviation from the mean. However, value added by the forestry sector alone is unavailable for many of the countries as it is aggregated under value added by the agriculture, forestry, and fishing sectors together. Hence, we used these estimates from the National Accounts Main Aggregates Database of the United Nations Statistics Division (NAS 2011) as a proxy.

For NTFPs, we assumed that only a certain percentage of forests is accessible, for which we used the GTAP data base. In addition, huge amounts of public expenditures are diverted towards maintaining forests. These expenditures were deducted from the gross benefits provided by forests. The global forest resources assessment of FAO (FAO 2010) study showed that in 2005, an average of US\$7.31 per hectare of public expenditure is allocated to forests for maintenance. There are regional differences, with costs ranging from US\$22.5/ha in Asia to US\$0.26/ha in South America. These total costs could be apportioned to different benefits or deducted from gross benefits.

#### 4. Empirical estimates of forest wealth

The world – and in particular developing countries – experienced accelerated economic growth over the study period of 1990 to 2010. The total GDP of the nations chosen in the study grew by 69 percent between 1990 and 2010, low- and middle-income countries experienced relatively higher growth, with GDP rising 66 percent to 547 percent cumulatively since 1990, while high-income countries grew between 20 and 66 percent. This growth however, would not have come without a trade-off in precious forest capital.

Figure 1 shows these apparent trade-offs between forest area and GDP change. It also illustrates that only high-income countries have managed to be in the positive quadrant, either due to tougher environmental regulations or to lower growth rate and population pressures as compared to less-developed countries. These countries also have less dependence on forests as source of livelihood, and thus managed a positive growth rate in GDP while maintaining forest capital base without compromising the asset for future generations - fitting our definition for sustainable growth. We noted a mixed trend for middle-income countries: some managed positive growth rates with an increase in forest area, while others were less successful. Low-income countries with higher concentrations of poor people clearly experienced growth at the expense

FIGURE 1 a: low income

FIGURE 1 b: lower middle income





#### **FIGURE 1**

Growth (measured in GDP) vis-a-vis forest area (in hectares) accumulation in different countries, 1990–2010

of forest resources. This also shows that as nations become richer and more developed, natural capital often stabilizes as countries rely on other forms of capital.

Forest assets in physical units can often be stable even as the quality of forest cover deteriorates. There are several indicators of forest quality, such as biodiversity, micro climate, growing stock, etc., the current information on which, at macro level, remains incomplete. Assuming that the growing stock is a reflection of the health of the forest ecosystem, Figure 2 indicates a trade–off between economic growth and forest quality. Even in this case, we see that high-income countries continue to remain in the positive quadrant, while low-income countries experience GDP growth amid deteriorating forest quality. Middle-income countries, which had been in the first quadrant, have shifted to the second quadrant. None of the countries could manage to be in the first quadrant. The demarcated area did not change much as timber was logged and not restocked in these countries – not a sustainable practice.

#### FIGURE 2 a: low income

#### MMR Percent change in GDP 400 кнм MOZ UGA BGD MLI NPL NER RWA KEN GMB MWI ZWE нті -100 --50 \_40 Percent change in growing stock

FIGURE 2 b: lower middle income





FIGURE 2 c: upper middle income





#### **FIGURE 2**

Trade-offs between economic growth and forest quality (measured by growing stock in cubic metres)

Figures 1 and 2 illustrate the changes in forest capital in physical units. The physical changes valued by respective shadow prices indicate the changes in forest wealth, which may vary significantly depending on the importance of forest ecosystem (see Figure 3).

With the exceptions of Rwanda and Gambia, most of the low-income countries depleted forest wealth while experiencing positive GDP growth,. Haiti experienced a decline in GDP while increasing forest wealth. Zimbabwe experienced declines in both GDP growth as well as forest wealth. The middle-income countries showed a mixed trend. India and China managed to be on the positive growth path while increasing their forest wealth, whereas Brazil and Indonesia are losing their rich forest capital. Most of the lower middle-income countries are growing at the expense of forest wealth. All the high-income countries in the sample have stabilized their forest wealth.

China, Brazil, India, Indonesia, USA, and Canada are the wealthiest countries in the year 2010 in terms of absolute forest wealth (see Table 2). Brazil has the second largest forest cover in the world, with some 56 percent of
FIGURE 3 a: low income



FIGURE 3 b: lower middle income

FIGURE 3 d: high income

FIGURE 3 c: upper middle income



#### **FIGURE 3**

Trade-offs between growth (measured in GDP) and forest wealth

its total geographical area under forests, while China has only 18 percent of geographic area under forests. Perhaps surprisingly, in terms of forest wealth, China emerged at the top. This is because of the difference in prices of timber from these two countries (pointing out the limitations of the export price data and other values that we used in the study). The five largest nations in the world are thus leading in terms of forest cover (aside from Russia, which was not considered in this study).

Figures 4a and 4b show changes in wealth vis-à-vis changes in GDP for the two decades

between 1990 and 2000 and 2000 and 2010. A ratio, for example, of -2 means forest wealth equivalent to two percentage points of GDP was lost in the process of increasing GDP by one unit. This ratio in general indicates the consumption of forest capital to create GDP. The decadal analysis shows that most countries had a negative ratio during both the decades 1990 to 2000, as well as 2000 to 2010. However, some countries with a negative ratio during 1990 to 2000 experienced a positive ratio from 2000 to 2010. India and China had increased their forest wealth between 1990 and 2000, but

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TABLE 2: Absolute forest wealth and per capita wealth in the countries selected for the study

| Country                  |                  |      | Absolute forest wealth<br>(in 2005 U.S. billion dollars) |        |        |        | Per capita wealth<br>(in 2005 U.S. thousand dollars) |       |      |      |
|--------------------------|------------------|------|--|--------|--------|--------|--|-------|------|------|
|                          | Income<br>group* | code | 1990   | 2000   | 2005   | 2010   | 1990   | 2000  | 2005 | 2010 |
| China                    | 2                | CHN  | 53,169   | 60,243 | 65,652 | 69,246 | 47   | 48    | 50   | 52   |
| Brazil                   | 3                | BRA  | 72,745   | 69,209 | 67,367 | 64,567 | 486  | 397   | 362  | 331  |
| India                    | 2                | IND  | 37,811   | 38,112 | 38,899 | 38,776 | 44   | 37    | 35   | 32   |
| Indonesia                | 2                | IDN  | 17,186   | 15,454 | 14,734 | 13,488 | 78   | 71    | 69   | 68   |
| United States of America | 4                | USA  | 19,560   | 20,195 | 20,497 | 20,915 | 96   | 72    | 66   | 57   |
| Canada                   | 4                | CAN  | 12,827   | 12,965 | 12,822 | 12,805 | 463  | 421   | 397  | 377  |
| Mexico                   | 3                | MEX  | 7,198  | 6,855  | 6,717  | 6,414  | 84   | 66    | 61   | 54   |
| Mozambique               | 1                | MOZ  | 4,048  | 3,843  | 3,737  | 3,645  | 298  | 210   | 178  | 152  |
| Peru                     | 3                | PER  | 3,558  | 3,487  | 3,487  | 3,433  | 163  | 134   | 126  | 117  |
| Congo, Republic          | 2                | COG  | 3,173  | 3,150  | 3,137  | 3,128  | 1,331  | 1,008 | 886  | 761  |
| Colombia                 | 3                | COL  | 3,079  | 3,029  | 3,004  | 2,950  | 92   | 76    | 70   | 64   |
| Myanmar                  | 1                | MMR  | 3,202  | 2,863  | 2,722  | 2,596  | 76   | 59    | 54   | 50   |
| Malaysia                 | 3                | MYS  | 2,354  | 2,708  | 2,563  | 2,448  | 129  | 116   | 99   | 87   |
| Turkey                   | 3                | TUR  | 1,958  | 2,066  | 2,184  | 2,212  | 36   | 33    | 32   | 31   |
| Nigeria                  | 2                | NGA  | 4,051  | 3,101  | 2,625  | 2,135  | 42   | 25    | 19   | 13   |
| Argentina                | 3                | ARG  | 2,397  | 2,218  | 2,138  | 1,999  | 73   | 60    | 55   | 50   |
| Vietnam                  | 2                | VNM  | 1,093  | 1,346  | 1,453  | 1,512  | 17   | 17    | 18   | 17   |
| Botswana                 | 3                | BWA  | 1,432  | 1,396  | 1,379  | 1,360  | 1,035  | 796   | 735  | 691  |
| Nepal                    | 1                | NPL  | 1,672  | 1,391  | 1,295  | 1,295  | 92   | 60    | 51   | 48   |
| Thailand                 | 3                | THA  | 1,270  | 1,243  | 1,228  | 1,207  | 22   | 20    | 19   | 18   |
| Philippines              | 2                | PHL  | 1,047  | 1,112  | 1,139  | 1,163  | 17   | 14    | 13   | 12   |
| Cambodia                 | 1                | KHM  | 1,461  | 1,294  | 1,196  | 1,123  | 161  | 106   | 90   | 78   |
| Zimbabwe                 | 1                | ZWE  | 1,455  | 1,240  | 1,132  | 1,025  | 139  | 99    | 89   | 78   |
| Cameroon                 | 2                | CMR  | 1,179  | 1,075  | 1,021  | 965    | 98   | 67    | 56   | 47   |
| South Africa             | 3                | ZAF  | 958  | 959    | 958    | 958    | 27   | 22    | 20   | 19   |
| Kenya                    | 1                | KEN  | 1,010  | 970    | 946    | 924    | 43   | 31    | 26   | 23   |
| Finland                  | 4                | FIN  | 817  | 897    | 889    | 891    | 164  | 173   | 169  | 166  |
| Mali                     | 1                | MLI  | 967  | 914    | 886    | 858    | 121  | 89    | 74   | 61   |
| Sweden                   | 4                | SWE  | 761  | 770    | 780    | 788    | 89   | 87    | 86   | 84   |
| Japan                    | 4                | JPN  | 638  | 715    | 742    | 746    | 5  | 6     | 6    | 6    |
| France                   | 4                | FRA  | 537  | 575    | 608    | 622    | 9  | 9     | 10   | 10   |
| Germany                  | 4                | DEU  | 441  | 496    | 512    | 532    | 6  | 6     | 6    | 7    |
| Italy                    | 4                | ITA  | 365  | 426    | 456    | 486    | 6  | 7     | 8    | 8    |

\* Note: 1 = Low income; 2 = Lower middle income; 3 = Upper middle income; 4 = High income.

| Country            |                  |      | Absolute forest wealth<br>(in 2005 U.S. billion dollars) |      |      | Per capita wealth<br>(in 2005 U.S. thousand dollars) |      |      |      |      |
|--------------------|------------------|------|--|------|------|--|------|------|------|------|
|                    | Income<br>group* | code | 1990   | 2000 | 2005 | 2010   | 1990 | 2000 | 2005 | 2010 |
| Papua New Guinea   | 2                | PNG  | 529  | 506  | 493  | 480  | 127  | 94   | 81   | 70   |
| Romania            | 3                | ROM  | 449  | 449  | 451  | 453  | 19   | 20   | 21   | 22   |
| Morocco            | 2                | MAR  | 350  | 364  | 393  | 389  | 14   | 13   | 13   | 12   |
| Chile              | 4                | CHL  | 367  | 378  | 382  | 385  | 28   | 24   | 23   | 22   |
| Guatemala          | 2                | GTM  | 398  | 387  | 382  | 376  | 45   | 35   | 30   | 26   |
| Uganda             | 1                | UGA  | 572  | 469  | 417  | 367  | 33   | 19   | 15   | 11   |
| Guyana             | 2                | GUY  | 435  | 371  | 348  | 325  | 600  | 499  | 457  | 414  |
| Malawi             | 1                | MWI  | 372  | 342  | 325  | 309  | 39   | 30   | 25   | 21   |
| Namibia            | 3                | NAM  | 351  | 322  | 307  | 291  | 248  | 170  | 151  | 134  |
| Ghana              | 2                | GHA  | 397  | 340  | 314  | 286  | 27   | 18   | 15   | 12   |
| Pakistan           | 2                | PAK  | 418  | 348  | 312  | 273  | 4    | 2    | 2    | 2    |
| Bangladesh         | 1                | BGD  | 258  | 254  | 251  | 249  | 2    | 2    | 2    | 2    |
| Poland             | 4                | POL  | 185  | 213  | 231  | 248  | 5    | 6    | 6    | 7    |
| Spain              | 4                | ESP  | 174  | 226  | 227  | 240  | 4    | 6    | 5    | 5    |
| Republic of Korea  | 4                | KOR  | 83   | 110  | 123  | 138  | 2    | 2    | 3    | 3    |
| Bulgaria           | 3                | BGR  | 99   | 108  | 118  | 127  | 11   | 13   | 15   | 17   |
| Austria            | 4                | AUS  | 94   | 108  | 114  | 113  | 12   | 13   | 14   | 13   |
| Czech Republic     | 4                | CZE  | 89   | 97   | 101  | 104  | 9    | 9    | 10   | 10   |
| Niger              | 1                | NER  | 162  | 111  | 104  | 100  | 21   | 10   | 8    | 6    |
| United Kingdom     | 4                | GBR  | 85   | 90   | 93   | 96   | 1    | 2    | 2    | 2    |
| Rwanda             | 1                | RWA  | 74   | 45   | 76   | 87   | 10   | 5    | 8    | 8    |
| Costa Rica         | 3                | CRI  | 72   | 67   | 70   | 73   | 23   | 17   | 16   | 16   |
| Algeria            | 3                | DZA  | 77   | 73   | 71   | 67   | 3    | 2    | 2    | 2    |
| Switzerland        | 4                | CHE  | 51   | 54   | 55   | 57   | 8    | 8    | 7    | 7    |
| Hungary            | 3                | HUN  | 47   | 51   | 54   | 54   | 5    | 5    | 5    | 5    |
| Sri Lanka          | 2                | LKA  | 75   | 64   | 57   | 53   | 4    | 3    | 3    | 3    |
| Gambia             | 1                | GMB  | 45   | 47   | 48   | 49   | 50   | 39   | 34   | 29   |
| Dominican Republic | 3                | DOM  | 41   | 42   | 41   | 41   | 6    | 5    | 4    | 4    |
| Ireland            | 4                | IRL  | 29   | 30   | 30   | 30   | 8    | 8    | 7    | 7    |
| Greece             | 4                | GRC  | 19   | 21   | 21   | 22   | 2    | 2    | 2    | 2    |
| Netherlands        | 4                | NLD  | 13   | 14   | 14   | 15   | 0.9  | 0.9  | 0.9  | 0.9  |
| Denmark            | 4                | DNK  | 7  | 8    | 10   | 11   | 1.3  | 1.4  | 1.9  | 1.9  |
| Luxembourg         | 4                | LUX  | 4  | 5    | 5    | 5  | 0.1  | 10.3 | 0.1  | 0.1  |
| Egypt              | 2                | EGY  | 4  | 5    | 6    | 6  | 10.0 | 0.1  | 9.7  | 8.9  |
| Haiti              | 1                | HTI  | 3  | 4    | 4    | 3  | 0.5  | 0.4  | 2.5  | 0.3  |

\* Note: 1 = Low income; 2 = Lower middle income; 3 = Upper middle income; 4 = High income.



FIGURE 4 b



#### FIGURE 4

Decadal average change in wealth to change in GDP





the growth rate declined thereafter. The opposite happened in Vietnam and Cost Rica due to afforestation programs In Costa Rica especially, forest cover increased due to various paymentfor-ecosystem service schemes. Brazil, which experienced decline in forest wealth during the period of 1990 to 2000, gained wealth between 2000 and 2010 due to active conservation policies, strict enforcement of forest laws and discouraging agricultural expansion in forest lands. Indonesia continues to lose forest wealth while growing GDP due to illegal logging and trade of timber, as well as deforestation for conversion to palm oil plantations. Some countries, including Romania, Bulgaria, and Haiti, experienced negative decadal growth during 1990 to 2000 while increasing the forest capital, while Zimbabwe experienced negative decadal growth during 2000 to 2010 while increasing the capital. The change in wealth to change in GDP ratio in rest of the countries is in the direction of the sign indicated. It is a matter of concern that of the 50 percent of the countries that experienced decline in wealth change to GDP growth ratios during 2000 to 2010, a majority are from low-income and lower middle-income countries.

The largest absolute declines in wealth occurred in Brazil, Nigeria, Indonesia, Nigeria,

Myanmar, and Zimbabwe. China, the U.S., India, Vietnam, and Turkey gained forest wealth between 1990 and 2010. In China and Vietnam significant efforts have been made to reforest, leading to increased forest cover since 2000. There has also been a country-wide ban on logging in both. In China, the low productive



FIGURE 5 Composition of forest wealth in selected countries, 2010

agricultural lands were converted to forests under the Grain for Green Program (the largest national payment-for-ecosystem services scheme). India is pursuing a national goal of covering one-third of its geographic area with forest and tree cover through greening, and has implemented a ban on green felling since 1987. The U.S. and Turkey, like other countries, have embarked on massive afforestation programs, resulting in an increase in forest area and hence total forest wealth.

Figure 5 shows the contribution of different broad categories of services – provisioning, recreational, regulating, and supporting – to the total forest wealth in selected countries. We see that the percentage contribution differs but this is largely governed by the availability of



Annual change in population

#### **FIGURE 6**

Relation between change in population and change in forest wealth (in percentage), 1990–2010 values and data. Moreover, we have not included all possible typology of values here due to non-availability.

It is important to analyze the reasons for decline in forest

wealth in addition to changes in GDP. Population is likely one determining factor. Hence, per capita measures may be more important than absolute ones. Figure 6 shows this mixed trend: countries at the top in population growth rate, such as Niger, Uganda, Gambia, Mali, Kenya, Congo, Cameroon, Nigeria, and Ghana, have clearly depleted forest capital. These countries are dependent on forests for fuel, wood, and charcoal. Some of these countries - Nigeria, Ghana, and Gambia - are very rich in mineral resources. To sustain growth and avoid the vicious circle of poverty and environmental degradation, more investment in forest capital is required in countries with higher population growth rates and rich natural resources. During the period of 1990 to 2010, Botswana, Mexico, Thailand, Niger, and Costa Rica emerged as the top gainers in growth in per capita forest wealth. However, the countries which led in absolute

wealth could not maintain per capita forest capital amid growing populations, with the exception of China.

In fact, only 22 percent of the countries have increased their per capita wealth since 1990, as seen in Figure 6. Clearly, the demographic pressure was high not only in lower-income and lower middle-income, but also in some high-income countries. What this means is that population growth has outpaced forest capital generation. From a standpoint of intergenerational sustainability, keeping per capita wealth constant is necessary.

Table 3 summarizes the trade-offs that the countries face between increasing growth vis-àvis the forest wealth for the four classified groups, to facilitate the understanding of these per-capita growth rates: I) Group I - Positive GDP growth and appreciation of forest capital; 2) Group 2 -Negative GDP growth and decline in forest capital; 3) Group 3 - Decline in GDP and increase in forest capital; and 4) Group 4 - Decline in both GDP and forest capital.

Table 2 illustrates lost forest wealth in selected countries. For the selected group of countries

#### TABLE 3 Changes in forest wealth and per capita capital vis-à-vis GDP growth

|  | Changes in absolute forest wealth,<br>1990–2010  | Per capita changes in forest wealth   |  |  |
|--|--|---|--|--|
| Growth in GDP while investing in forest capital    | USA, China, India, Sweden, Japan, Finland,<br>Egypt, Spain, Chile, France, Vietnam,<br>Switzerland, Ireland, Denmark, Gambia,<br>Rwanda, Netherlands, Germany, Czech<br>Republic, Poland, Italy, Philippines,<br>Romania, Republic of Korea, Morocco,<br>Bulgaria, Greece, Austria, UK, Costa Rica,<br>Hungary, Canada, Malaysia | China, Sweden, Japan, Finland, Egypt,<br>Spain, France, Vietnam, Denmark,<br>Netherlands, Germany, Czech Republic,<br>Poland, Italy, Romania, Republic of Korea,<br>Bulgaria, Greece, Austria, UK, Hungary  |  |  |
| Growth in GDP while depleting forest capital       | Brazil, Indonesia, Peru, Mexico, Colombia,<br>Mozambique, Myanmar, Argentina,<br>PNG, Republic of Congo, Cameroon,<br>Thailand, Guyana, Nepal, Guatemala,<br>Ghana, Namibia, Nigeria, Mali, South<br>Africa, Botswana, Cambodia, Pakistan,<br>Bangladesh, Algeria  | Brazil, Canada, USA, Indonesia, Peru,<br>Mexico, Colombia, Mozambique, Myanmar,<br>Argentina, PNG, Algeria, Republic of<br>Congo, Malaysia, Cameroon, Thailand,<br>Turkey, Guyana, Nepal, Guatemala,<br>Ghana, Namibia, Nigeria, Mali, South<br>Africa, Botswana, Cambodia, Pakistan,<br>Bangladesh, Philippines, Morocco, Costa<br>Rica, Kenya |  |  |
| Decline in GDP as well as depleting forest capital | Zimbabwe   | Zimbabwe, Haiti   |  |  |
| Decline in GDP while increasing forest capital     | Haiti  |   |  |  |

considered in the study, 2.87 percent of forest area has been lost since 1990. However, in monetary terms, we see a net positive change in overall forest wealth due to accumulation of timber (at the rate of 0.02 percent), accumulation of carbon in old growth forests, and improvement in the quality of forests in biodiversity-rich regions. While developed countries are maintaining their forest capital, most low-income and lower middle-income countries, in which natural resources are often a principal source of income for the poor, are losing forest capital,. At the same time, developing low- and middle-income countries are the prime drivers of GDP growth from 1990 to 2010. Thus, we see apparent difficult choices to be made between economic growth and forest capital. Because of the existence of lower limits to environmental degradation, there are thresholds beyond which the collapse of certain natural resources can be irreversible. Concerted effort is needed to recognize the significant amount of forest wealth that is lost as a result of these tradeoffs, and efforts must be made to sustain natural capital for intergenerational equity.

#### 5. Discussion and conclusions

This chapter has focused on estimating forest capital wealth in 68 countries using estimates on forest area, growing stock, and carbon, based on the FAO database and the ecosystem values published in the TEEB database. It was not possible to include all ecosystem services due to the non-availability of information. Therefore, these results provide only rudimentary, yet useful, estimates of forest wealth, and whether they are being managed sustainably. In addition, the estimates helps understand the exact contribution of forests, as well as the real beneficiaries of forest capital, and aid in decision-making by looking at whether investment in forest capital is adequate vis-a-vis other assets.

The results indicate at least two key messages. The first is that collating these ecosystem values within an accounting framework confirms that forest capital is often a significant component of total wealth. The second is that many countries appear to be losing significant amounts of this wealth. In effect, these countries are mining this renewable resource, likely due to a mixture of socioeconomic and demographic pressures. Whether or not these losses can be compensated by possible gains in other forms of wealth is a critical question. Because of the particular characteristics of forest assets and the often distinctive nature of benefits these stocks provide, this question is at least open to considerable debate.

Towards the beginning of this chapter, we were at pains to emphasize the preliminary nature of the empirical approach that we take in obtaining results for this number of countries. Having now outlined our findings, it is worth reiterating this message once more. In particular, our estimates are subject to an array of caveats which include (but are not restricted to) the following:

- Valuing ecosystem services: our coverage here is necessarily limited by the extent of the empirical record. While practical valuation of ecosystems has advanced tremendously, there remain – relatively speaking
   – substantial gaps. This raises a number of issues for using these data, not least reliance on transferring values between different countries.
- Quality of forest capital: while the evidence base for values makes clear that forests provide multiple ecosystem services, there is less routine modeling of the way in which these services are dependent on each other. For example, carbon sequestration depends on forest biomass, which in turn depends on properties such as the quality of soil and water moisture that the forest holds. This has a number of subsequent ramifications.
- Spatial diversity: while it is possible to talk generically about the benefits forest capital provides, more robust accounting needs to distinguish the spatial productivity of different forest assets. Consequently, a more "bottom-up" evaluation of land areas is needed. As such, the largely "topdown" approach taken here is only a placeholder for those more detailed efforts.
- Thresholds: it is clear therefore that forest capital accounting should ideally be based on a bedrock of metrics indicating the

physical properties of the asset. This could also include knowledge of how far these ecosystems are from thresholds: i.e., some non-linearity after which (marginal) benefits provided are dramatically less. This is a matter of particular importance, for example, when evaluating changes in forest area (and, perhaps, changes in quality).

• Asset values: The value of a forest asset is essentially a forward-looking accounting magnitude, albeit based on what can be measured in the present. Yet, the stream of future services provided by this asset that will be valued in the future might depend on a number of factors, such as development outcomes (e.g., changing incomes per capita) and, indeed, the scarcity of the asset itself. Abbott and Fenichel (2014), for example, demonstrate a number of important complications in using (marginal) ecosystem benefits to value natural assets.

These are substantial conceptual and practical challenges. Not surprisingly then, natural capital accounting for forest assets remains a work-in-progress. In the interim, we argue that even rudimentary – but more comprehensive – values offer useful guidance, especially if the alternative is to assign (implicitly) a zero value to these elements. This holds only to the extent that these initial steps to assessing forest wealth across countries are used as the basis for further methodological progress rather than viewed as the end of the accounting process itself.

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# Challenges to ecosystem service valuation for wealth accounting

**Edward Barbier** 

#### **KEY MESSAGES**

**In recent years, substantial progress** has been made by economists working with ecologists and other natural scientists in valuing some ecosystem goods and services.

However, difficulties in measurement, data availability and other limitations still preclude the valuation of certain ecosystem services.

There is often uncertainty associated with estimated ecosystem service values, and even more so with scaling up of local values to regional or national levels or updating these values annually, which poses problems for their use in wealth accounts. In the absence of reliable estimates, the temptation is to use "second-best" estimates, or to transfer values from other locations; however, such methods should be used with caution and only under specific circumstances, at the risk of generating unrealistic values.

**Progress in incorporating ecological capital** in wealth accounts therefore requires developing more accurate methods of valuing ecosystem goods and services and applying them to a wider range of ecosystems.

#### 1. Introduction<sup>1</sup>

The Inclusive Wealth Report 2012 recognized that ecosystems should be viewed as economic assets, as through their natural functioning and habitats they provide important goods and services to the economy (BARBIER 2012A, UNU-IHDP AND UNEP 2012). If ecosystems are to be considered capital assets – or ecological capital for short – then efforts to modify net domestic product (NDP) to include natural and human capital should account for the contributions of ecosystems, as well and depreciation of these assets (see Figure I). Specifically, NDP should be adjusted to include two contributions due to ecological capital:

- the value of the direct benefits provided by the current stock of ecosystems; and
- any capital revaluation as a result of conversion of ecosystems to other land uses, with the "price" of changes in ecological capital reflecting the present value of the future direct and indirect benefits of ecosystems (BARBIER 2012A).

However, estimating these benefits poses many challenges. The most significant problem is that very few are marketed. Some of the products provided by ecosystems, such as raw materials, food, and fish harvests, are bought and sold in markets. But the majority of ecosystem goods and services are not, and thus do not have any market value. These include many services arising from ecosystem processes and functions that benefit human beings largely without any additional input from them, such as coastal protection, nutrient cycling, erosion control, water purification, and carbon sequestration. In recent years, substantial progress has been made by economists working with ecologists and other natural scientists in applying environmental valuation methodologies to assess the welfare contribution of these services. The various valuation methods employed are essentially standard economic techniques.<sup>2</sup>

Nevertheless, some of the non-marketed benefits of ecosystems are proving more elusive to value compared to others. Difficulties in measurement, data availability, and other limitations often preclude the application of standard economic valuation methods to certain ecosystem services. Consequently, there are gaps in in both geographic coverage, as well as scale of application. There is often uncertainty associated with estimated ecosystem service values, and even more so with scaling up of local values to regional or national levels. The latter issue especially poses a problem for wealth accounting, which requires aggregate values at the regional or national level. These must be replicated and updated annually, yet most valuations are for case studies in a specific location and time period.

Because of the difficulty in valuating these non-market benefits, only a handful ecosystem services have been calculated, and these estimates are often unreliable. If one or more valuation studies for a given ecosystem service exist, the correct estimate for wealth accounting purposes must still be chosen carefully. In the absence of reliable estimates, the temptation is to use "second-best" estimates, or to transfer values from other locations. However, such methods should be used with caution and only under specific circumstances, at the risk of generating unrealistic values. Finally, because for wealth accounting purposes ecosystem service values need to be aggregated to the regional or national level and replicated annually, it is also important to understand the specific challenges of "scaling up" values estimated only for a specific case study location and time period.

<sup>1</sup> Research assistance provided by Sheikh Muhammad Eskander.

<sup>2</sup> For example, Barbier (2007, 2011a, and 2011b), Bateman et al. (2011), EPA (2009), Freeman (2003), Hanley and Barbier (2009), Mendelsohn and Olmstead (2009), NRC (2005), and Pagiola et al. (2004) discuss how these standard valuation methods are best applied to ecosystem services, emphasizing in particular both the advantages and the shortcomings of the different methods and their application.



#### **FIGURE 1**

Adjusting GDP for reproducible, human, natural, and ecological capital

These challenges for applying ecosystem service valuation for wealth accounting are the focus of this chapter.<sup>3</sup> To facilitate the discussion, examples are drawn from one type of ecosystem – estuarine and coastal systems (ECEs). There is a growing literature of valuation studies

focusing on the benefits of these systems, and policy-makers have expressed an interest in incorporating estuarine and coastal ecosystem services in national wealth accounts and environmental policy decisions (BÖRGER ET AL. 2014, RICHARDSON ET AL. 2014, UNU-IHDP AND UNEP 2012). Thus, ECEs provide many examples of the challenges faced in valuing ecosystem services for wealth accounting.

<sup>3</sup> A more general discussion of the rationale, methodologies, and challenges in valuing ecosystem services in a wealth accounting framework can be found in "Ecosystem Services and Wealth Accounting", Chapter 8 of the Inclusive Wealth Report 2012 (Barbier 2012a).

#### FIGURE 2

Valuing ecosystem goods and services and ecological capital depreciation



#### 2. Valuing ecosystem goods and services

In identifying the ecosystem services provided by natural environments, a common practice is to adopt the broad definition of the Millennium Ecosystem Assessment that "ecosystem services are the benefits people obtain from ecosystems" (MEA 2005). Although this definition has been interpreted in different ways, a consensus is now emerging as to precisely what ecosystem services are and how they relate to ecological processes and functions.

First, a wide range of valuable goods and services to humans arise in myriad ways via the structure and functions of an ecosystem. For example, some of the living organisms found in an ecosystem might be harvested or hunted for food, collected for raw materials, or simply valued because they are aesthetically pleasing. Some ecosystem functions, such as nutrient and water cycling, can also benefit humans through purifying water, controlling floods, recharging aquifers, reducing pollution, or simply by providing more pleasing environments for recreation. These various benefits provided by an ecosystem via its structure and functions are what is meant by ecosystem services.

Second, although they are the source of ecosystem services, the structure and functions of an ecosystem are not synonymous with such services. Ecosystem structure and functions describe the components of an ecosystem and its biophysical relationship regardless of whether or not humans benefit from them. In contrast, as stated by EPA (2009), "ecosystem services are the direct or indirect contributions that ecosystems make to the well-being of human populations". Quantifying these contributions, or "benefits", in terms of human welfare is often referred to as valuing ecosystem services.

Figure 2 summarizes why quantifying and valuing ecosystem services is important for wealth accounting purposes. Human drivers of ecosystem change, such as pollution, resource exploitation, land conversion, species introductions, and habitat fragmentation, affect the structure and functioning of ecosystems. Assessing and quantifying this impact is important, as it alters the ecological production of ecosystem goods and services that benefit humans. These changes and alteration to ecosystems, especially through habitat degradation, fragmentation, and conversion, constitute the key ways in which ecological capital is "depreciated".<sup>4</sup> The role of economic valuation is to measure explicitly gains and losses in human welfare over time due to these changes. These values can then be used to estimate the direct benefits provided by the current stock of ecosystems, and any losses

<sup>4</sup> In addition, the stock of ecological capital can be enhanced through restoration of degraded ecosystems or establishing new ecosystems through converting other land uses, in which case the objective of wealth accounting is to measure any resulting "appreciation" of ecological capitaldeon through such restoration activities. See Barbier (2011a) for further discussion, and Barbier (2014) for the specific wealth accounting methodology required, which is applied to the example of coastal wetland restoration in Louisiana, USA.

of the future direct and indirect benefits that occur when this ecological capital is degraded, fragmented, or converted to other land uses.

The biggest challenge to quantifying and valuing ecosystem services is inadequate knowledge to link changes in ecosystem structure and function to the production of valuable goods and services (BARBIER 2011A AND 2011B, NRC 2005, POLASKY AND SEGERSON 2009). This is certainly true for the various ecosystems found in coastal areas and estuaries, which are bodies of water and their surrounding coastal habitats typically found where rivers meet the sea.

The key habitats comprising estuarine and coastal ecosystems (ECEs) include marshes, mangroves, forested swamps, near-shore coral reefs, seagrass beds, oyster reefs, mud flats, barrier islands, and sand beaches and dunes. For many of these habitats, we often do not know how variation in ecosystem structure, functions, and processes give rise to the change in an ecosystem good or service, although we are starting to learn about some of these impacts. For example, in the case of coastal wetlands, the change could be in the spatial area or quality of a particular type of wetland, such as a mangrove, marsh vegetation, or swamp forest. The subsequent loss of habitat and vegetation may not only influence shellfish and other species that depend on the habitat but also reduce protection of shorelines and control of erosion. In addition, the loss of important wetland breeding and nursery habitat may influence a variety of valuable goods and services provided in neighboring marine systems, such as commercial or recreational fisheries. Alternatively, changes in ECE habitat could be due to variation in the flow of water, energy, or nutrients through the system, such as the variability in tidal surges due to coastal storm events, the influx of organic waste from onshore pollution, or the impacts of oil spills and other human-induced hazards.

Another problem encountered in quantifying and valuing ecosystem services is that very few are marketed. Some of the products provided by ECEs, such as raw materials, food, and fish harvests, are bought and sold in markets. Given that the price and quantities of these marketed products are easy to observe, there are numerous ways to estimate the contribution of the environmental input to this production (barbier 2007, freeman 2003, mcconnell AND BOCKSTAEL 2005). However, many other key ECE services arising from ecosystem processes and functions do not lead to observable marketed outputs. These include many services that benefit human beings largely without any additional input from them, such as coastal protection, nutrient cycling, erosion control, water purification, and carbon sequestration. In recent years, substantial progress has been made by economists working with ecologists and other natural scientists in applying environmental valuation methodologies to assess the welfare contribution of some of these services, such as coastal protection, carbon sequestration, and habitat-fishery linkages (EPA 2009, NRC 2005, BARBIER 2011A, BARBIER ET AL. 2011, POLASKY AND SEGERSON 2009).

Table I provides some examples of how specific ECE goods and services are linked to the ecological structure and functions underlying each service. It also cites, where possible, economic studies that have estimated the values arising from the good or service. The list of 94 studies included in Table I is only representative of the literature on economic valuation of ECEs; nevertheless, the table gives an indication of the range of valuation estimates available for specific goods and services, and is thus instructive.

As Table I indicates, most valuation studies for ECE goods and services are for tourism, recreation, education, and research; maintenance of fishing, hunting, and foraging activities; and,deon increasingly, storm protection. There are fewer valuation estimates of ecosystem services associated with regulatory functions, such as erosion control, flood protection, water pollution and sediment control, and carbon sequestration. There are no ECE valuations of maintaining temperature and precipitation. Such patchy coverage across different ecosystem goods and service is typical for other ecosystems as well. Again, this is in large part attributed to

#### TABLE 1

Examples of estuarine and coastal ecosystem services and valuation studies

| Ecosystem structure and function  | Ecosystem services  | Valuation examples (94 studies)  |  |  |  |  |
|---|---|--|--|--|--|--|
| Attenuates and/or dissipates<br>waves, buffers wind   |   | Badola and Hussein (2005); Barbier (2007) and (2012a); Barbier et al. (2008); Barbier et al. (2013); Barbier and Enchelmeyer (2014); Costanza et al. (2008); Das and Crépin (2013); Das and Vincent (2009); Kim and Petrolia (2013); King and Lester (1995); Landry et al. (2011); Laso Bayas e al. (2011); Mangi et al. (2011); Petrolia and Kim (2009) and (2011); Petrolia et al. (2014); Pompe and Rinehart (1994); Wilkinson et al. (1999). (19 studies)  |  |  |  |  |
| Provides sediment stabiliza-<br>tion and soil retention   | Erosion control   | Huang et al. (2007); Landry et al. (2003); Parsons and Powell (2001);<br>Rulleau et al. (2014); Sathirathai and Barbier (2001). (5 studies)  |  |  |  |  |
| Water flow regulation and control   | Flood protection  | Morgan and Hamilton (2010); Turner et al. (2004). (2 studies)  |  |  |  |  |
| Provides nutrient and<br>pollution uptake, as well as<br>retention, particle deposition,<br>and clean water | Water pollution and sediment control  | Breaux et al.(1995); Byström (2000); Leggett and Bockstael (2000);<br>Massey et al. (2006); Smith (2007); Smith and Crowder (2011); Turner et<br>al. (2004);van der Meulen et al. (2004). (8 studies)  |  |  |  |  |
| Generates biogeochemi-<br>cal activity, sedimentation,<br>biological productivity                           | Carbon sequestration  | Barbier et al. (2011); Luisetti et al. (2011); Mangi et al. (2011); Pendleton et<br>al. (2012); Sikamäki et al. (2012); Thompson et al. (2014). (6 studies)  |  |  |  |  |
| Climate regulation and stabilization  | Maintenance of tem-<br>perature, precipitation                                  | (O studies)  |  |  |  |  |
| Generates biological produc-<br>tivity and diversity  | Raw materials and food  | Janssen and Padilla (1999); King and Lester (1995); Naylor and Drew<br>(1998); Nfotabong Atheull et al. (2009); Pinto et al. (2010); Ruitenbeek<br>(1994); Sathirathai and Barbier (2001). (7 studies)   |  |  |  |  |
| Provides suitable reproduc-<br>tive habitat and nursery<br>grounds, sheltered living<br>space               | Maintains fishing,<br>hunting, and foraging<br>activities                       | Aburto-Oropeza et al. (2008); Barbier (2003), (2007) and (2012); Barbier<br>and Strand (1998); Bell (1997); Freeman (1991); Janssen and Padilla<br>(1999); Johnston et al. (2002); Lange and Jiddawi (2009); Luisetti et<br>al. (2011); McArthur and Boland (2006); Milon and Scrogin (2006);<br>O'Higgins et al. (2010); Plummer et al. (2012); Samonte-Tan et al. (2007);<br>Sanchirico and Mumby (2009); Stål et al. (2008); Swallow (1994); White<br>et al. (2000). (20 studies)   |  |  |  |  |
| Provides unique and aes-<br>thetic landscape, suitable<br>habitat for diverse fauna and<br>flora            | Tourism, recreation, edu-<br>cation, and research                               | Bateman and Langford (1997); Birol and Cox (2007); Brander et al.<br>(2007); Brouwer and Bateman 2005; Coombes et al. (2009); Johnston<br>et al. (2002); Kaoru (1993); Kaoru et al. (1995); King and Lester (1995);<br>Kreitler et al. (2013); Landry and Liu (2009); Lange and Jiddawi (2009);<br>Luisetti et al. (2011); Mathieu et al. (2003); Milon and Scrogin (2006);<br>Othman et al. (2004); Pendleton (1995); Pendleton et al. (2012); Smith<br>and Palmquist (1994); Smith et al. (1991); Tapsuwan and Asafu–Adjaye<br>(2006); Turner et al. (2004); Whitehead et al. (2008). (23 studies) |  |  |  |  |
| Provides unique and aes-<br>thetic landscape of cultural,<br>historic, or spiritual meaning                 | Culture, spiritual and reli-<br>gious benefits, existence<br>and bequest values | Bateman and Langford (1997); Milon and Scrogin (2006); Naylor and<br>Drew (1998); Subade and Francisco (2014). (4 studies)   |  |  |  |  |

the difficulty of determining how changes in ecosystem structure and function influence the ecological production of services, and that many key services are not marketed or directly related to some form of marketed economic activity and thus difficult to value.

#### 3. Selecting valuation studies

Even if good valuation studies exist for one or more ecosystem services, one has to be careful to select the correct estimate for wealth accounting purposes. This often requires knowledge of the valuation technique used as well as the target population whose willingness to pay (WTP) for the service has been assessed. This issue can be illustrated with the example of the storm protection service of restored marsh in Louisiana, USA.

The number of economic studies that estimate the protective value of estuarine and coastal ecosystems has increased in recent years, especially for marsh and mangroves (see Table I). Since Hurricane Katrina in 2005, the U.S. Gulf Coast, and especially the state of Louisiana, has been the focus of many estimates of ECE protective values. Most of these studies have focused on estimating the WTP for the post-Katrina coastal restoration efforts in Louisiana. The reported values for a few key studies are summarized in Table 2, which for comparison purposes, are presented in constant 2005 US\$.

#### TABLE 2

Valuation estimates for the storm protection benefits of coastal restoration, Louisiana

|  | Valuation method and purpose  | Year of<br>esti–<br>mate | Coastal<br>area<br>(ha) | Mean WTP<br>(current US\$)              | Mean WTP<br>(2005 US\$)                 |
|--|---|--------------------------|-------------------------|---|---|
| Petrolia et al.<br>(2014)ª                                       | Choice experiment survey of U.S.<br>households for coastal wetland<br>and barrier island restoration,<br>southeast Louisiana            | 2011                     | 94,547                  | \$149 per household<br>\$181,220 per ha | \$129 per household<br>\$157,341 per ha |
| Petrolia and Kim<br>(2011) <sup>6</sup>                          | Contingent valuation survey of<br>Louisiana households for prevent-<br>ing future coastal land losses,<br>Louisiana                     | 2009                     | 181,013                 | \$53 per household<br>\$628 per ha      | \$48 per household<br>\$571 per ha      |
| Barbier et al.<br>(2013), Barbier<br>and Enchelmeyer<br>(2014) ° | Ex ante expected damage function<br>estimate of reduced storm surge<br>flooding damages to residential<br>property, southeast Louisiana | 2006                     | 2.98                    | \$41 per ha                             | \$40 per ha                             |
| Landry et al. (2011)   | Choice experiment survey of U.S.<br>and Louisiana households for<br>flood risk reduction from coastal<br>restoration, Louisiana         | 2007                     | NA                      | \$103 per household                     | \$97 per household                      |

Note: WTP = willingness to pay; ha = hectares; NA = not available.

a) One-time value, intermediate-scale restoration, consequential respondents only. The WTP per ha estimate is based on the value of the storm protection benefits of Louisiana coastal restoration to the estimated number of households in the U.S. of 114,991,721 in 2011 (Petrolia et al. 2014) population.
b) Annual value, based on a lower bound annual WTP for preventing future coastal wetland losses of \$111 per household survey respondent, and households citing storm protection benefits as a top priority were 47.6 percent more likely to pay for preventing coastal wetland loss compared to all others.
c) Based on \$99 per 0.1 increase in wetland-water ratio per m along a 5,961 meter (m) hurricane transect, and a \$24 per 0.001 increase in bottom friction through in wetland vegetation along the same transect. The estimate here assumes that both benefits result from the creation of wetland of 2.98 ha (596.1 m by 50 m) along this transect.

Although all of the studies estimate the value of the same ecosystem service (storm protection), ECE habitat (restored salt marsh), and location (Louisiana, USA), there are important differences among them. With the exception of Petrolia et al. (2014), who estimate the marginal WTP of the U.S. population for coastal restoration in Louisiana, all values are for Louisiana residents. In addition, the studies differ in terms of which components of the storm protection benefits from restoring coastal wetlands are estimated.

As Kousky (2013) has pointed out, estimating the economic costs associated with natural disasters, such as storm surges or high-velocity winds, is difficult both conceptually and practically. For example, the range of costs include damages to commercial and residential property, infrastructure losses, emergency and cleanup costs, negative capital shocks to affected regions, mortality or illness, and altered risk perceptions (KOUSKY 2013). Over time, there may also be shoreline retreat as coastal residents and economic activity move inland, and even impacts on economic growth (HALLEGATTE AND GHIL 2008, KOUSKY 2014, LOAYZA ET AL. 2012, SMITH ET AL. 2006, TOYA AND SKIDMORE 2007).

However, a further problem for estimating the protective value benefits of ECE restoration is that these benefits must be assessed from an ex ante perspective, i.e., in anticipation of protection against future storm damages and economic costs. That is, as ECEs are restored, they are expected to reduce the likelihood, or risk, that an undesirable event, such as coastal flooding from storm surges, might occur. The WTP for increased wetland restoration is therefore directly related to the reduction in perceptions of the risk of bad outcomes occurring, such as mortality or illness, damages to property and other assets, and the disutility associated with risk exposure (i.e., risk aversion). As shown by Barbier (2014), these different factors mean that the overall marginal WTP by a representative coastal household for the storm protection benefits from restored wetlands comprises three components:

- the effect on the expected utility of the household from the change in the subjective probability that the storm occurs;
- the marginal benefit of the reduction in the exposure to the risks associated with the storm occurring,<sup>5</sup> and;
- the marginal benefit from the reduction in expected losses, such as to property, inflicted by any storm.

Different valuation methods can be employed to measure all three components simultaneously or, alternatively, one or more individually. This is reflected in the various studies listed in Table 2. For example, Petrolia et al. (2014) and Petrolia and Kim (2011) estimate the ex ante WTP for the full range of storm protection benefits of coastal wetland restoration, i.e., all three components simultaneously. In contrast, Barbier et al. (2013) and Barbier and Enchelmeyer (2014) calculate only the marginal benefit from the reduction in expected property losses from coastal wetland protection, whereas Landry et al. (2011) value the marginal benefit of the reduction in the risk of stormrelated flooding.

From a wealth accounting perspective, some of these values may be relevant for accounting for the value of the "additional ecological capital" that is created when coastal wetlands are restored. In particular, the value of the storm protection service of restored ECEs, in terms of the likely reduction in property and other asset damages will already be reflected by changes in net domestic product (NDP), which takes into account changes in the value of these assets.

For example, in the U.S. national accounts, residential housing services are a component of personal consumption expenditure (PCE), and consequently part of gross domestic product (GDP). The rental value of tenantoccupied housing and the imputed rental value

<sup>5</sup> For example, a reasonable assumption is that the household dislikes exposure to the risks associated with a storm, and that this disutility includes not only risk aversion but also the risk of possible injury, illness, or death if storm-related surge flooding or other damage occurs.

of owner-occupied housing are both part of PCE housing services, reflecting the amount of money tenants spend for the service of shelter and the amount of money owner occupants would have spent had they been renting. Deducted from the income generated from housing services are expenses such as maintenance and repairs, insurance, property taxes, mortgage interest, and any subsidies such as disaster relief payments, leaving a profit-like remainder of business income that is recorded as actual (for tenant-occupied housing) or imputed (for owner-occupied housing) rental income. As rental income measures income from current production, capital gains or losses resulting from the change in prices of existing assets are excluded. The services of non-residential dwellings and land owned by a business are included in its income from sales, and thus included in the accounting of income for the appropriate sector for each business.

In sum, if the expected damages avoided arise through protecting another economic asset, such as property, then there is no addition to wealth arising from this ecosystem service. To adjust NDP further for ECE protection values based on their indirect contributions in terms of protecting dwellings, buildings, and other property would therefore amount to double counting (BARBIER 2012B). On the other hand, if restoring estuarine and coastal systems involves a change in land use and values, then NDP should be further modified to reflect any resulting change in capitalized land values (BARBIER 2014).

The reduction in expected storm surge damages is only one component of the marginal WTP for any restoration. This ex ante WTP will also depend on avoiding or lowering the risks associated with the storm, which may be valued especially highly by more risk-averse households.<sup>6</sup> Valuations of flood risk exposure and reduction reveal that this WTP is in excess of expected flood damages, most likely reflecting disutility of risk exposure as well as mortality and morbidity risks (BROUWER ET AL. 2009, LANDRY ET AL. 2011, NAVRUD ET AL. 2012). Hedonic housing price studies before and after major U.S. hurricanes reveal that subjective risk perceptions, including risk aversion, influence housing prices as well as the purchase of flood insurance (BIN AND POLASKY 2004, HALLSTROM AND SMITH 2005, PETROLIA ET AL. 2013, SMITH ET AL. 2006).

To date, most studies of the protective benefit of ECEs do not estimate any resulting impacts on either the disutility from risk aversion or the risk of possible injury, illness, or death if the storm surge happens. One exception is Landry et al. (2011), who estimate that the average U.S. household is WTP \$103 for coastal restoration to reduce flood risk in New Orleans (see Table 2). Some ex-post valuations have documented the role of ECEs, notably mangroves, in reducing storm-related deaths. For example, Das and Vincent (2009) estimate that, during the 1999 cyclone that struck Orissa, India, there would have been 1.72 additional deaths per village within 10 km of the coast if mangroves had been absent. Similarly, Laso Bayas et al. (2011) calculate that, during the 2004 Indian Ocean tsunami, mangroves, forests, and plantations situated between villages and the coastline in Acheh, Indonesia may have decreased loss of life by 3 to 8 percent.

Finally, as mentioned previously, it is also possible through survey methods to estimate the entire marginal WTP for storm protection benefits. For example, Petrolia et al. (2014) employ a choice experiment survey for different coastal wetland restoration programs in southeast Louisiana, which include different levels of protection afforded by restored wetlands (see Table 2). They find that the average U.S. household in their national survey is willing to pay US\$149 for an intermediate increase in storm surge protection through coastal wetland restoration, but will pay only US\$2 more for a further increase to high levels of protection. Similarly,

<sup>6</sup> This may especially be a problem when ex post damages from past storms are used as the basis for measuring the ex ante WTP for expected damages (Barbier 2007, Freeman 1989 and 2003, pp. 243–247).

Petrolia and Kim (2011) estimate that a lower bound annual WTP for preventing future coastal wetland losses is US\$111 per household survey respondent in Louisiana. As households citing storm protection benefits as a top priority were 48 percent more likely to pay for preventing coastal wetland loss compared to all others, one can infer that the overall storm protection benefits were \$53 per household (see Table 2).

#### 4. Second-best valuation methods

A common problem is a lack of reliable estimates of the value of ecosystem services that can be used for including the contribution of ecological capital into wealth accounts. In such circumstances, it is tempting to employ "back of the envelope" estimates as a substitute for the lack of reliable valuation studies. For example, in the case of estimating the value of the storm

#### FIGURE 3

Replacement cost vs. expected damage function estimation of protective value



prevention and flood mitigation services of the "natural" storm barrier function of mangrove and other ECEs, the replacement cost method is often used as a shortcut valuation approach, as this method involves simply estimating the costs of replacing coastal habitat by constructing physical barriers to perform the same services (CHONG 2005, KING AND LESTER 1995, MANGI ET AL. 2011, SATHIRATHAI AND BARBIER 2001).

However, the replacement cost approach should be used with caution in estimating the value of an ecosystem service, such as storm protection. First, one is essentially estimating a benefit (e.g., storm protection) by a cost (e.g., the costs of constructing seawalls, groins, and other structures), and second, the human-built alternative is rarely the most cost-effective means of providing the service (BARBIER 2007, FREEMAN 2003, SHABMAN AND BATIE 1978).

Figure 3 illustrates the limitation of using the replacement cost method to estimate the protective value of an ECE. Assume that the ecosystem comprises a coastal wetland, such as a marsh or mangrove, of initial landscape area So. The cost of the storm protection service provided by the ecosystem is "free" and thus corresponds to the horizontal axis, oSo. However, suppose part of the wetland is lost or converted, and so the ecological landscape decreases to SI. The replacement cost method would suggest that the value of this loss in wetland area could be estimated by the cost of "replacing" the lost wetlands with seawalls, breakwaters, levies, and other human-built structures to reduce storm surge and waves.

In Figure 3, the marginal cost of an alternative, human-built coastal storm barrier is  $MC_h$ . Thus, the "replacement cost" of using the human-built barrier to provide the same storm protection service as the SoS1 amount of wetlands lost is the difference between the two supply curves, or area SoABS1. However, this overestimates the benefit of having the wetlands provide the storm protection service. The true benefit of this ecosystem service is the demand curve, or total willingness to pay, for the service provided by SoS1 amount of wetlands less the

#### TABLE 3

Valuation of storm protection service of mangroves, Thailand, 1996–2004

| Annual deforestation rate             | FAO <sup>ª</sup><br>18.0 km <sup>2</sup> | Thailand <sup>b</sup><br>3.44 km <sup>2</sup> |
|---------------------------------------|--|---|
| Valuation approach (US\$)             |  |   |
| Replacement cost method: <sup>c</sup> |  |   |
| Annual welfare loss                   | 25,504,821                               | 4,869,720                                     |
| Net present value (10% discount rate) | 146,882,870                              | 28,044,836                                    |
| Net present value (12% discount rate) | 135,896,056                              | 25,947,087                                    |
| Net present value (15% discount rate) | 121,698,392                              | 23,236,280                                    |
| Expected damage function approach:    |  |   |
| Annual welfare loss                   | 3,382,169                                | 645,769                                       |
| Net present value (10% discount rate) | 19,477,994                               | 3,718,998                                     |
| Net present value (12% discount rate) | 18,021,043                               | 3,440,818                                     |
| Net present value (15% discount rate) | 16,138,305                               | 3,081,340                                     |

a) FAO estimates from FAO (2003). 2000 and 2004 data are estimated from 1990–2000 annual average mangrove loss of 18.0 km2.
b) Thailand estimates from various Royal Thailand Forestry Department sources reported in Aksornkoae and Tokrisna (2004). 2000 and 2004 data are estimated from 1993–1996 annual average mangrove loss of 3.44 km2.

c) Based on replacement cost method assumptions of Sathirathai and Barbier (2001).

Source: Adapted from Barbier (2007).

costs of providing protection. In Figure 3, this net benefit corresponds to area SoCDSI. Thus, the replacement cost method overestimates the net benefits of the storm protection service by area ABCD.

A more reliable method would be to estimate directly the net benefit of the storm protection service, or area SoCDS1 in Figure 3. As discussed in the previous section, one such method is the expected damage function (EDF) approach. If there is an increase in wetland area, such as through ecological restoration, then the value of this change is the total amount of expected flood damage losses avoided (BARBIER AND ENCHELMEYER 2014, BARBIER ET AL. 2013). If there is a reduction in wetland area, as shown in Figure 3, then the welfare loss is the total increase in expected flood damages resulting from a storm event (BARBIER 2007). In both instances, the valuation would be a compensation surplus measure of a change in the area of wetlands and the storm protection service that they provide.

A comparison of using the expected damage function (EDF) approach described in the previous section and replacement cost method of estimating the welfare impacts of a loss of the storm protection service due to mangrove deforestation in Thailand confirms that the latter method tends to produce extremely high estimates compared to the EDF approach (BARBIER 2007). The comparison of annual and net present values produced by the two methods is depicted in Table 3. But the expected damage function has its own limitations; for example, when households are risk averse, using ex post damage estimates can be a poor proxy for the ex ante willingness to pay to reduce or avoid the risk from storm damages (BARBIER 2007, FREEMAN 2003, PP. 243-247). Nevertheless, because the EDF approach is a direct compensation surplus measure of a change in the area of ECEs and the storm protection service that they provide, it is a more reliable method of estimating the protective value of these ecosystems.

Shabman and Batie (1978) suggest that the replacement cost method can provide a reliable valuation estimation for an ecological service, but only if the following conditions are met: (1) the alternative considered provides the same services; (2) the alternative compared for cost comparison should be the least-cost alternative; and (3) there should be substantial evidence that the service would be demanded by society if it were provided by that least-cost alternative. Unfortunately, very few replacement cost studies meet all three conditions. One study that does estimates the value of using wetlands for abatement of agricultural nitrogen load on the Baltic Sea coast of Sweden (BYSTRÖM 2000). The replacement value of the wetlands is defined and estimated as the difference between two cost-effective reductions of agricultural nitrogen pollution: one that uses wetlands for nitrogen abatement, and one that does not. The results show that the use of wetlands as nitrogen sinks can reduce by 30 percent the total costs of abating nitrogen pollution from agriculture in Sweden.

#### 5. Transferring values

A second approach that is becoming more frequently employed to overcome the lack of reliable valuation estimates for ecosystems is benefit or value transfer. This method involves taking estimates of economic value from one site and "transferring" them to a similar location elsewhere (NRC 2005, JOHNSTON AND ROSENBERGER 2010, PLUMMER 2009, RICHARDSON ET AL. 2014, ROSENBERGER AND STANLEY 2006, TROY AND WILSON 2006). In the benefit transfer literature, the location from which the valuation estimates are taken is called the study site, because it is the site that has already been "studied" in some way to obtain the original valuation estimate. The location to which the estimates are applied is called the policy site.

Plummer (2009) reviews the extensive environmental economics literature on the limits to implementing benefit transfer, especially in the context of coastal and marine ecosystem goods and services. He concludes that the errors in applying this technique can be minimized provided that there is sufficient ecological and economic correspondence between the study and policy sites. Plummer (2009) suggests that "lack of correspondence" can be reduced when:

- the ecosystem at the study site is a good match for the ecosystem under consideration at the policy site (i.e., ecological correspondence), or;
- the respective populations of the study and policy sites do not differ considerably in terms of income levels, benefits derived from the ecosystem, preferences, employment and economic opportunities, household characteristics (e.g., occupation, education, number of adults and children, etc.), and other attributes that would cause wide variances in willingness to pay estimates between populations at the study site and populations at the policy site (economic correspondence).

In addition, Troy and Wilson (2006, p. 436), who have applied the benefit transfer method in valuing freshwater and coastal wetland benefits, also conclude:

"One of the biggest potential pitfalls in value transfer occurs when values are drawn from study sites that are situated in very different contexts than targeted policy sites. For example, to simply assume that the economic value of a freshwater wetland in one ecological region is going to be the same for a freshwater wetland in a wholly different region would be inappropriate."

The advancement in benefit transfer methods and modeling techniques, including the application of geographical information systems (GIS) and meta-regression analysis, means that there are more opportunities to use these methods as a way of extrapolating and transferring estimated ecosystem service values from one location, population, and time to other locations, populations, and periods. This is exemplified by a study that compares the importance of income and cultural differences in a coastal benefits transfer exercise in Ireland (HYNES ET AL. 2013). However, technique is not a substitute for reliability. If there is a lack of economic and ecological correspondence between study and policy sites, transferring values between the two sites through GIS and other methods will simply lead to inaccurate valuation estimates (TROY AND WILSON 2006).

Similarly, there are potential drawbacks of applying benefit transfer through meta-analysis regression. This requires knowledge of the values of the independent variables for the policy site of interest, and assumes that the statistical relationship between the dependent and independent variables is the same between the study and policy sites (RICHARDSON ET AL. 2014, ROSENBERGER AND STANLEY 2006). If one can statistically control for these differences in ecological and economic correspondence, this reduces the benefit transfer errors. In addition, there needs to be a sufficient number and variety of reliable policy site valuation studies to make the meta-analysis regression applicable in the first place. For example, as Table 2 shows, only a handful of ECE goods and services may serve this purpose. Unfortunately, benefit transfers may be less helpful in overcoming the lack of reliable estimates for the benefits associated with some important regulatory functions, such as erosion control, flood protection, water pollution and sediment control, carbon sequestration, and maintenance of temperature and precipitation.

#### 6. Scaling values

Often for wealth accounting purposes, it is ideal to have aggregate values of ecosystem services at the regional or national level that can be replicated and updated annually. In contrast, most ecosystem service values, such as those listed in Table I for ECEs, apply to case studies in a specific location and over a specific time period. The problems associated with "scaling up" these values to the regional or national level, or applying past estimates to different and multiple time periods, demand serious consideration. Another problem involved with scaling is defining the population of beneficiaries for each service, e.g., how many people benefit from the storm protection service of a given mangrove forest. Simple approaches to scale temporally and spatially ecosystem values for wealth accounting should be used with caution.

A common approach to creating an aggregate benefit measure at a regional or national level is to take some average land unit value, such as US\$ per hectare (ha) or acre, estimated for an ecosystem service at a specific location, and multiply this value by the total area of all similar ecosystems found within a region or country. However, using a simple physical index of area, such as ha or acres, to expand a value estimate to another scale violates basic economic principles, such as diminishing marginal utility, changing relative scarcity and substitutability, and thus will lead to large estimation errors (BOCKSTAEL ET AL. 2000, BULTE AND VAN KOOTEN 2000, RICHARDSON ET AL. 2014). The problem of this approach is summarized by Richardson et al. (2014, p. 3):

"For example, the per-hectare fishery production value of a coral reef cannot simply be multiplied by the total acreage of coral reefs in a country, region or the world to calculate the total fishery production value of the country's, region's or world's coral reefs. The fish harvest lost with the loss of the one local reef may be compensated through increased fish imports from other areas or a switch to other foods. Such substitution possibilities generally decline with increasing spatial scale: the loss of a nation's, region's or the entire world's seafood harvest will have an increasingly nonmarginal effect on the relative scarcity and thus will increase the marginal value of a unit of seafood (and thus of the reef) because it will be much harder to substitute for than the loss of one reef."

Extending valuation estimates using a simple physical index of area may not even be appropriate for valuing entire ecological landscapes. Evidence suggests that the ecological production functions underlying some ECE services vary spatially across habitat area (AGUILAR-PERERA AND APPELDORN 2008, BARBIER ET AL. 2011, GEDAN ET AL. 2011, KOCH ET AL. 2009, PETERSON AND TURNER 1994, ROUNTREE AND ABLE 2007). In particular, storm protection and support for marine fisheries provided by coastal wetland habitats, such as mangroves and salt marshes, tend to decline with the distance inshore from the seaward edge.

Increasingly, valuation studies of ECE services are taking into account this spatial variability. In the Gulf of California, Mexico the mangrove fringe with a width of 5 to 10 meters has the most influence on the productivity of near-shore fisheries, with a median value of US\$37,500 per hectare. Fishery landings also increased positively with the length of the mangrove fringe in a given location (ABURTO-OROPEZA ET AL. 2008). In Thailand, the values for storm protection service and support for offshore fisheries also vary spatially, with the benefits higher on the seaward fringe rather than further inland (BARBIER ET AL. 2008 AND BARBIER 2012B). However, when the values for these services are assessed in combination with the benefits from controlling nutrient pollution across an entire coral reef-seagrass-mangrove seascape, both the seaward and furthest inland boundaries of the mangrove habitat have higher values than the interior of the system (BARBIER AND LEE 2013).

There are also limitations to transferring values estimated for one specific time period to other periods. One problem is that a valuation estimate may simply be outdated, especially if preferences or demand for a specific ecosystem service have changed considerably over time (RICHARDSON ET AL. 2014). In addition, advances in valuation methodologies may mean that past benefit estimates based on less reliable methods are now less valid or reliable (JOHNSTON AND ROSENBERGER 2010). For example, as discussed above, through the 1990s, estimates of storm protection and erosion control services of ECEs mostly employed the replacement cost method, which tends to overestimate values (BARBIER 2007, CHONG 2005). In contrast, more recent

estimates use improved valuation methods that more accurately measure willingness to pay for these services (see, for example, the studies listed in Table 2).

Finally, temporal variability in the ecological production of key ecosystem services may also affect the value of that service. For example, the coastal protection of the wave attenuation by vegetation contained in certain ECEs, such as mangroves, marshes, and seagrass beds, can vary if damaging storm events occur when plant biomass and/or density are low (BARBIER ET AL. 2011, KOCH ET AL. 2009). This is particularly important in temperate regions, where seasonal fluctuations of biomass may differ from the seasonal occurrence of storms. For example, along the U.S. Atlantic coast the biomass of seagrass peaks in the summer (April to June), and decreases in the fall (July to September), when storm events usually strike. In tropical areas, mangrove and seagrass vegetation has relatively constant biomass throughout the year, so the coastal protection service is relatively unaffected by seasonal or temporal variability. Nonetheless, the value of coastal protection of tropical ECEs is affected significantly when storm surge occurs during high or medium, as opposed to low, tides (KOCH ET AL. 2009).

#### 7. Conclusion

Significant progress has been made in recent years in estimating the values of key ecosystem services. This progress has sparked interest in using such valuation estimates to measure and account for the changes of ecological capital in wealth accounts. As shown in Figure I, the methodology for extending adjustments to net domestic product (NDP) to allow for the contributions of ecosystems is straightforward. For example, employing this methodology, the Inclusive Wealth Report 2012 demonstrates how NDP in Thailand has been affected by annual mangrove loss that has occurred since the 1970s (BARBIER 2012A). The three principle mangrove benefits are coastal protection, habitat-fishery linkages, and the collection of wood and non-wood products. The overall wealth benefits of these services was US\$0.57 per person in Thailand in the 1970s, when large-scale mangrove deforestation first started, but only \$0.28 per person by the 2000s, when deforestation had slowed considerably.

However, adapting valuation estimates of ecosystem goods and services for wealth accounting purposes poses some unique challenges. To facilitate and guide such approaches, this chapter has discussed many of these important challenges. Difficulties in measurement, data availability and other limitations can prevent the application of standard economic valuation methods to certain ecosystem services, such as coastal protection, nutrient cycling, erosion control, water purification, and carbon sequestration. Yet it is often these benefits that make the contribution of ecological capital so valuable. Consequently, efforts to incorporate ecosystems into wealth accounts frequently encounter significant gaps in the geographic coverage, scale of application, and reliability of the estimated ecosystem service values available from existing studies.

To overcome such gaps and produce aggregate regional or national ecosystem service values that can be updated annually, there is increasing interest in transferring and scaling up existing values, or to produce new values quickly through more expedient methods. As discussed in this chapter, such approaches need to be used with caution, as they are prone to problems of measurement error and bias.

In conclusion, there is no "short cut" substitute for the lack of reliable valuation studies for ecosystem good and services. Progress in incorporating ecological capital in wealth accounts is therefore ultimately tied to further progress in developing more accurate methods of valuing ecosystem goods and services and applying them to a wider range of ecosystems. Conducting additional and sound ecosystem valuation studies is the only dependable approach to addressing the gaps in geographic coverage and the scale of application that prevent incorporation of ecological capital in wealth accounting.

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## Using inclusive wealth for policy evaluation: the case of infrastructure capital

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#### KEY MESSAGES

Wealth accounting to date has focused primarily on the *assessment* of *past* performance in economies, by measuring changes to produced, natural, and human capital.

**In order to use inclusive** wealth for policy *evaluation*, we must estimate the impacts of a given policy on the trajectories of the capital stocks that comprise wealth.

**Infrastructure is an important policy** domain because proposed changes to current systems affect many, if not all, capital stocks, which results in capital stock interactions and trade-offs.

A *systems* view of policy evaluation is necessary in order to map and quantify these impacts and trade-offs; this can be managed using conceptual and mathematical models that capture integrated physical and economic processes.

**To illustrate how one might** conduct wealthbased policy evaluation, we use two infrastructure case studies – coal-fired power generation in China and the High Aswan Dam in Egypt. The case studies rely on integrated physical and economic models to quantify capital stock impacts of *past* infrastructure decisions.

**Such models can be used** to evaluate *prospective* infrastructure systems as well, although doing so requires careful consideration of future uncertainty. Scenario analysis is a useful and flexible method for incorporating uncertainty into wealth-based policy evaluation.

#### 1. Introduction and motivation<sup>1</sup>

Inclusive wealth theory suggests that to sustain intergenerational human well-being, a society must sustain its productive base - the collection of its capital assets (DASGUPTA AND MÄLER 2000, HAMILTON AND CLEMENS 1999). The sum value of these capital assets, valued at social prices (also called shadow prices), is defined as the inclusive wealth of the society. Empirical work in wealth accounting has principally dealt with the assessment of past performance in various economies, by measuring changes to produced. natural, and human capital (ARROW ET AL. 2012, MUÑOZ ET AL. 2012). To conduct policy evaluation however, whether in infrastructure or other sectors, we must estimate the impacts of policy interventions on the future trajectories of the capital stocks comprising wealth. Wealth-based policy evaluation is equivalent to carrying out a social cost-benefit analysis (DASGUPTA 2009). In other words, if a policy were to increase the inclusive wealth of a society (valued at social prices), then the present discounted value of its social profits will be positive, making it acceptable (ARROW ET AL. 2003a).

Policies that can impact wealth at national scale often involve proposed investments in infrastructure. Infrastructure capital, a subset of produced capital, is commonly defined as the built civic assets essential for the functioning of an economy and society. Essential infrastructure sectors include electric power, water supply, agriculture production, and transportation, among others<sup>2</sup>. Infrastructure capital is typically planned, built, and managed in dedicated systems of assets (e.g., power grids, water networks, mass transit systems) and can be publicly or privately owned. These systems benefit

society by supplying or transforming inputs in the form of electricity, clean water, food, and roads. Reliable access to these services facilitates economic production and consumption. However, these systems can also impact society negatively – through air pollution, greenhouse gas emissions, and the degradation of ecosystems – as a consequence of their productive activity. Furthermore, the benefits and costs of prospective infrastructure to human and natural capital differ in magnitude and timing, and can vary with technology. Thus, assessing the effects of proposed changes in infrastructure capital on the inclusive wealth – and ultimate well-being – of a society is not trivial.

The economics literature has largely focused on the effect of infrastructure on economic growth. For example, Calderon et al. (2014) use a "synthetic" measure of infrastructure, along with human and produced capital, to estimate a production function. Their synthetic measure includes physical quantities of roads, power, and telecommunications, in contrast to pecuniary measures such as spending or capital formation. Other studies have linked growth in infrastructure to total factor productivity through direct and indirect mechanisms (GRAMLICH 1994, ROMP AND DE HAAN 2007, STRAUB ET AL. 2008). Related work has assessed network effects and the non-linear impact of infrastructure services on productivity gains (HURLIN 2006, RÖLLER AND WAVERMAN 2001), as well as the mediating role of institutions (ESFAHANI AND RAMÍREZ 2003).

Infrastructure policy evaluation necessitates a different set of perspectives and methods than *ex post* assessment, which tries to establish a relationship between infrastructure investments in the aggregate and overall economic growth, typically at national levels and within rigid sectoral boundaries. Infrastructure planning traditionally occurs at different geographical scales (local, regional, national), varying temporal scales (near- to mid-term, long horizon), using sectorspecific approaches (power networks are different from transportation networks), and firmly embedded in an institutional context (centralized, coordinated, or decentralized market-based

<sup>1</sup> The authors would like to acknowledge the national science foundation and the center for complex engineering systems of MIT for partially funding this work.

<sup>2</sup> For a list of what the U.S. Department of Homeland Security defines as critical infrastructure sectors, please see: http://www.dhs.gov/ critical-infrastructure-sectors

projects). However, infrastructure planning increasingly emphasizes a systems view that reconciles time scales, cross-sectoral interdependencies, and spatial relationships (HALL ET AL. 2014). Integrating evaluation at the project level (a single, large multi-purpose dam), sector level (electric power generation), and systems level (interdependent water-energy system) with noninfrastructure capital stock linkages (human and natural capital impacts) is therefore at the frontier of infrastructure policy.

Inclusive wealth provides a framework for evaluating infrastructure through measured changes to capital stocks, the underlying drivers of development. Quantifying these impacts - for instance, the impact of coal-fired electric power on air quality and thus health and human capital - requires systemic understanding of the biophysical and economic processes that link the capital stocks in an economy and mediate their movement through time. Ekins (2012) developed a systems model of wealth creation by focusing on the role of each capital stock in the production process of an economy, and the capital feedback effects that occur as a result of production (in particular see Figure 1 in his paper). However, he draws less attention to the linkages that exist between each of the capital stocks, and the processes that influence their interactions. Finally, while previous work has laid the theoretical foundation for doing wealth-based policy/project evaluation (ARROW ET AL. 2003A, DASGUPTA 2009), to our knowledge it has yet to be operationalized for infrastructure.

To this end, we suggest a framework for evaluating prospective infrastructure interventions according to their impact on capital stocks at national scale. The framework builds on the welfare-theoretic foundations of inclusive wealth, thus while the chapter focuses on infrastructure, the framework is general to other policy domains. Application of the framework to specific infrastructure decisions can be carried out with integrated models that quantify the impacts and interactions between capital stocks. We demonstrate this with two case studies of infrastructure evaluation – power generation in China and the high Aswan dam in Egypt. We also discuss scenario analysis as a useful approach for incorporating uncertainty in evaluations.

## 2. A framework for evaluating infrastructure capital

This section develops a conceptual foundation for organizing wealth-based policy evaluation, followed by a mathematical formulation specific to infrastructure.

## 2.1 Mapping linkages between capital stocks: conceptual model

Conducting wealth-based policy evaluation requires analyzing factors not traditionally included in policy or project evaluation. For example, an evaluator contemplating construction of a hydroelectric dam generally confines the evaluation to the technical parameters of alternative design concepts and the narrow application of cost-benefit analysis (WORLD COMMISSION ON DAMS 2000). The evaluator is less likely to consider all of the extended and propagating impacts the dam will have on natural and human capital in the region. This relatively limited scope of evaluation neglects impacts, positive or negative, outside the small boundary of analysis. In contrast, wealth-based policy evaluation suggests a systems view of prospective policies to ascertain their full impact on the trajectories of all capital stocks. Figure I conceptualizes an economy as a dynamic system of linked capital stocks; because they are integrated a policy or perturbation affecting one stock will subsequently impact another through various channels.

Figure 1 consists of four interacting feedback loops that affect well-being via production and consumption. We assume, as is common in the literature, that well-being is a function of consumption, recognizing at the same time that well-being is derived from non-market goods and services and that consumption in

#### **FIGURE 1**

Inclusive wealth of an economy represented as a dynamic, interconnected system (stocks denoted in boxes)

Note: the "+" symbol at the head of an arrow means that a change in one direction (increasing or decreasing) at the tail will lead to a change in the same direction at the head. The "-" symbol at the head of an arrow means that change in one direction at the tail



an economy may not be equitably enjoyed. The first loop, RI, is a reinforcing loop, which means that an increase (decrease) in one variable will ultimately feed back to produce further increase (decrease) in that variable in the future, all else being equal. The reinforcing effect denoted by this loop is analogous to the neoclassical growth models, where economic output Y is a function of the stock of produced capital K, and some fraction of Y is reinvested to produce more K, thus increasing output further. Similarly, as production and consumption in an economy increase and fuel economic growth, the stock of produced capital will also grow. As this stock grows, it reinforces further production and consumption in the economy.

The second reinforcing loop, R2, operates via human capital. Human capital generally consists of the education, skills, core competencies, and health of both the working and total population in a society. As the stock of produced capital accrues, the stock of human capital also increases. Benefits of produced capital such as access to reliable electricity, better schools and hospitals, and modern technology and education enhance labor productivity. Furthermore, as recent macroeconomic literature suggests, enhanced human capital leads to more production and consumption (e.g., ROMER 1989). The reinforcing loops RI and R2 denote the mechanisms of economic growth studied most in the econometric literature. Scholars have attempted to derive empirical estimates of the elasticities of (produced) capital and labor (now called human capital) in various countries of the world. However, there is more to sustainable development than these reinforcing loops alone reveal.

Modern society depends heavily on natural capital - non-renewable and renewable resources and local and global ecosystems - to provide the goods and services we enjoy. Bringing natural capital into the picture introduces the notion of *balancing* loops, where an increase (decrease) in one variable will feed back through the balancing mechanism to decrease (increase) that same variable in the future. The balancing loop, BI, is the frequently-discussed resource depletion loop. Growth in the stock of produced capital leads to a decline in natural capital, eventually diminishing the resourcedependent produced capital. Technological progress (not included explicitly in the figure) dampens the balancing effect, and the timing and force with which this loop will impinge on a society is greatly debated. Furthermore, while the diagram aggregates stocks of produced and natural capital into single variables, scarcity

and other market or regulatory pressures may drive substitution between assets. The extent to which produced capital can replace natural capital in production processes, or natural capital stocks can substitute for each other, again depends heavily on technological progress, and the social, economic, and political context of the nation.

The second balancing loop, B2, captures the impact of depleting natural capital stocks on human capital via pollution and ecosystem degradation. In other words, a country pollutes and degrades ecosystems by mining natural capital to develop its produced capital base. These deleterious effects, in whatever form they are manifest, decrease the stock of human capital predominantly through reductions in health, such as premature deaths, spread of infectious diseases, and increased morbidity. As the stock of human capital decreases, production and consumption in the economy also decrease. In some cases, reductions in natural capital may have an immediate impact on well-being, where, for instance, the physical presence of old-growth forests or sacred groves are a source of well-being in a society.

The relative strengths of the two types of loops affect the trend of development. If the reinforcing loops dominate the balancing loops, then development tends to be sustainable because produced and human capital grow faster than natural capital declines. In contrast, if the balancing loops dominate the system, development tends to be unsustainable, because as natural capital declines, it exerts downward pressure on human and produced capital<sup>3</sup>. Production and consumption could still be increasing in the latter case, particularly if quantified using measures like GDP (gross domestic product), which fail to capture many aspects of human well-being. Thus, while not explicitly indicated in Figure I, it is important to recall the central tenet of inclusive wealth theory, namely that well-being moves in the same direction as wealth, the value aggregation of all capital stocks in an economy's productive base.

The figure also delivers another key *systems* intuition: produced capital, including its infrastructure assets, is a central driver of change in the other capital stocks. It facilitates gains in production and consumption, both directly and indirectly via human capital. Produced capital choices also drive resource depletion and the externalities affecting human health and wellbeing. Evaluating policies for the produced capital sector – whether in extraction, manufacturing, or infrastructure – therefore requires evaluation of impacts to *all* other capital stocks.

Thus while our framework focuses on infrastructure, it could be extended to include other forms of resource- and pollution-intensive produced capital, such as factories and heavy industry. Infrastructure, however, is unique in that every country needs some basic level of infrastructure, while goods derived from heavy industry can be imported. Thus, infrastructure planning and its effect on national sustainable development is a germane problem for every national government.

### 2.2 Measuring capital stock impacts of infrastructure: mathematical model

What are the impacts of a coal-fired vs. solarpowered electricity system on the stocks of human and natural capital, as well as production and consumption and well-being, over time? To answer questions like this one, we develop a framework that extends the formulation of Dasgupta (2009), Dasgupta and Mäler (2000), and Arrow et al. (2003b). We define Y(t)as gross national product at time t, a measure of the output achievable given the produced capital stock K(t), the human capital stock H(t), the extraction rate R(t) of a natural capital stock N(t), and total factor productivity A(t). Total

<sup>3</sup> There is a third mode of behavior, whereby the negative effect of the balancing loops on the capital stocks exactly cancels out the positive effect of the reinforcing loops. In this case the rate of the change in the capital stocks will be zero and the system will be in steady-state. By inclusive wealth theory, this situation represents sustainable development.

factor productivity measures the residual proportion of output not explained by the inputs of production, capturing the effect of technological progress, the efficiency with which inputs are used, and institutional conditions. If we ignore that total factor productivity can be a function of investments in produced and human capital and the rate at which natural capital is extracted, then gross national product can be described as:

EQUATION 1

#### Y(t) = A(t)f(K(t), H(t), R(t))

Furthermore, we define the depreciation rate corresponding to produced and human capital as  $\delta_{\kappa}$  and  $\delta_{H}$  respectively and investment in each as  $I_{\kappa}$  and  $I_{H}$ . We assume that N(t) is a renewable resource such that it grows according to a regeneration function g(N(t)) and declines according to the extraction rate R(t). On timescales relevant to policy-making for sustainable development, we would assume no regeneration function for non-renewable resources. Aggregate consumption at time t is denoted C(t), and we assume that [intergenerational] well-being is derived from the utility of that consumption. The rate of change for each capital stock is therefore:

EQUATION 2

$$\frac{dK(t)}{dt} = Y(t) - C(t) - I_H(t) - \delta_K K(t) = I_K(t)$$
$$\frac{dH(t)}{dt} = I_H(t) - \delta_H H(t)$$
$$\frac{dN(t)}{dt} = g(N(t)) - R(t)$$

We then modify the structure to account for the effects of infrastructure capital – an explicit subset of produced capital – on human capital (the same could be done for natural capital). Consider the example of a country's electric power system. A fossil-based power plant consumes some amount of non-renewable natural capital (e.g., oil, natural gas, coal) and converts it to usable electricity via a set of technologies and processes. The process generates some amount of pollution, which may negatively affect the surrounding communities. Well-being may decline directly as a result of simply seeing the pollution, but to illustrate the links between capital stocks we focus only on the impacts that pollution has on human capital. Thus we modify the equations according to the following:

EQUATION 3

$$\frac{dK(t)}{dt} = I_K(t) = \frac{dK^{IN}(t)}{dt} + \frac{dK^O(t)}{dt}$$
$$\frac{dH(t)}{dt} = I_H(t) - \delta_H H(t) \pm \delta_H^{IN} H(t)$$
$$\delta_H^{IN} H(t) = \frac{dK^{IN}(t)}{dt} \frac{\partial H(t)}{\partial K^{IN}(t)}$$
$$\frac{\partial H(t)}{\partial K^{IN}(t)} = f(K^{IN}(t))$$
$$K^{IN}(t) \equiv \langle a_1(t), a_2(t), \cdots, a_n(t) \rangle$$

Investment in produced capital is split between investments in infrastructure capital  $K^{IN}$ , which we'll assume for now is just in electric power, and other non-infrastructure produced assets *K*<sup>0</sup>. For our purposes, we care only about changes to K<sup>IN</sup>. Increases to human capital come from investment, education being the primary driver in the literature (BARRO AND LEE 2013, KLENOW AND RODRIGUEZ-CLARE 1997). Reductions come from the "normal" (or baseline) depreciation  $\delta_{\mu}$  in human capital, which results from people naturally aging and dying. The net infrastructure-induced effect  $\delta^{IN}_{\mu}$  on human capital could be positive or negative. For instance, while pollution has an obvious negative impact on human capital, infrastructure can also provide reliable access to lighting so that students can study at night.

Finally, the marginal effect of another unit of power generation on human capital depends on the attributes a(t) of the electric power system. This vector of attributes defines the system's technology, efficiency, fuel source, environmental controls, age, etc. at time *t*.
An important concept in infrastructure capital is *inertia*. By inertia we mean that the underlying attributes a(t) of the infrastructure capital stock cannot instantaneously change; there is a lag between an investment decision and its change to the system. Inertia is important because evaluating the merits of a proposed change to the infrastructure stock requires modeling the inertia of the system for projecting its evolution and subsequent impacts into the future. Consider again the example of a fossil fuel-based electricity system. The system is a collection of infrastructure capital assets, comprising different vintages. These vintages may be associated with different power production efficiencies, factor input requirements, and pollution emission levels. As a result, an electric power system with old, inefficient, and polluting assets will likely have a more damaging net effect on capital stocks than a newer, more efficient system. Furthermore, even if all new builds were from renewable sources, the inertia in the energy system means that in the absence of forced retirements many of the fossil-based assets will remain for decades - some coal power plants in OECD countries have been in operation for almost 100 years. Thus, in the absence of retrofitting, these plants will continue to pollute the air and degrade ecosystems.

The impact on human capital of pollution from infrastructure could manifest in several ways. Recall the equation for human capital used in the 2012 Inclusive Wealth Report (UNU-IHDP AND UNEP 2012, Annex 2): EQUATION 4

 $H = e^{\rho A} \times Pop_{15+} \times \int_{t=0}^{T} \bar{r} e^{-\delta t} dt$ 

Aggregate human capital H is the product of human capital per adult (determined by the market return  $\rho$  on educational attainment A), the adult population, and the average discounted labor compensation over an entire life's working period, T. Pollution from coal-fired electric power could either directly reduce the adult population via premature deaths, or it could decrease a life's working period through extended absenteeism at work. Importantly, the youth population is not represented in the human capital equation. Yet, premature mortality and respiratory disease for the younger population will have the largest impact on human capital decline if aggregated over an entire lifetime since the rate of return on that capital in the earlier years is higher (HECKMAN 2000).

Double counting is another important issue. The "normal" depreciation of human capital must be clearly separated from the infrastructure-induced depreciation in the accounting exercise. If done appropriately, one could devise two trajectories of human capital, where the effect of infrastructure-induced decline is either included or not (Figure 2).

The two mechanisms through which infrastructure-induced impacts can decrease human capital have different implications for inclusive wealth. The direct effect of pollution on population through premature deaths is an impact on the physical *stock* of human capital. The indirect effect that shortens a life's working period is an impact on the shadow price of human capital; workers work less over time because of pollution-related illnesses, and wages are foregone due to premature deaths. While we focus on the human capital impacts of infrastructure in order to develop the framework, marginal impacts to natural capital could be measured in the same way - for instance, the ecosystem degradation effects of coal- vs. solar-powered generation.

#### 3. Case studies of infrastructure policy

In this section we apply the framework outlined above to two case studies of *past* infrastructure decisions: (1) power generation in China and (2) the high Aswan Dam in Egypt. We summarize two studies that used counterfactual analysis techniques and integrated physical and economic models, and reanalyze their conclusions in the context of inclusive wealth. Integrated

#### FIGURE 2

Infrastructure-adjusted human capital may be lower than the baseline projection due to healthrelated reductions in the adult population and/or life's working period.



models are useful for wealth-based policy analysis because they provide a method for implementing the systems view in evaluation by capturing capital stock linkages and interactions (models focused exclusively on either the physical world or the economy cannot readily do this). We use the modeling results from these studies to evaluate whether or not inclusive wealth (and therefore well-being) increased or decreased as a result of the corresponding infrastructure interventions.

## 3.1 Air pollution and health impacts of power generation in China

China experienced rapid growth in produced capital over the period of 1990 to 2008 (Figure 3). Growth in produced capital is one of the primary reasons China's development was considered sustainable over this 19-year period; in fact China had the largest per capita growth rate in wealth across all 20 countries analyzed in the 2012 Inclusive Wealth Report, despite massive population growth (MUÑOZ ET AL. 2012). At the same time, this rapid development has led to increased pollutant emissions, mostly from the country's coal-fired power generation. Degraded air quality is one of the most prominent and visible environmental impacts of the electricity system, and China's development more broadly (WORLD BANK AND SEPA 2007).

We can quantify the impact of electric power infrastructure on human capital by retrospectively analyzing the impact that air pollution has had on Chinese health and economy. Matus et al. (2012) quantified such impacts stemming from ozone and particulate matter pollution between 1975 and 2005. They estimated mortality (deaths) and morbidity (disease) using concentration information coupled with health impact functions drawn from epidemiological literature. Using a computable general equilibrium (CGE) model, they incorporated medical costs as well as mortalities in the form of a labor shock. The model calculates welfare losses that accrue due to: (I) wages foregone from premature mortality and morbidity and (2) broader economic losses from reduced consumption and investment and labor input. While the authors call (I) welfare loss, this quantity is equivalent to the second two terms of Equation (4), namely



#### FIGURE 3

Percent change in capital stocks per capita in China compared to the base year (1990 to 2008); reproduced using data from the UNU– IHDP and UNEP (2012) Data annex



|  | Monetary<br>value (billions<br>of 1997 US\$) | Share of total<br>welfare loss<br>(%) |
|--|--|---------------------------------------|
| Total welfare loss                                   | 69.0   | 100.0                                 |
| Direct loss due to mortalities from chronic exposure | 42.6   | 61.7                                  |
| Mortalities that occurred in 2005                    | 4.4  | 6.4                                   |
| Loss in 2005 from prior year cases                   | 38.2   | 55.4                                  |
| Direct loss due to other health outcomes             | 6.7  | 9.7                                   |
| Non-fatal health outcomes                            | 6.5  | 9.4                                   |
| Mortalities from acute exposure                      | 0.2  | 0.3                                   |
| Broader economic losses                              | 19.7   | 28.6                                  |

#### TABLE 1

Decomposition of welfare loss in 2005; these estimates represent the difference between what historically occurred and what would have occurred under a cleaner infrastructure scenario. Reproduced with permission from Matus et al. (2012).

the reduction in working adults times foregone earnings, calculated dynamically over the 30-year time period. As Table I shows, the wages lost due to mortalities represented the largest share of welfare loss (61.7 percent), indicating that the human capital effects of air pollution in China far exceed other effects<sup>4</sup>. To then estimate the marginal effect of coal-fired power infrastructure on human capital –  $\delta_H^{IN}H(t)$  from Equation (3) – one would need to relate ozone

occurred as a result of lower gross income reduced the produced capital stock. While not as large as the human capital impacts, this illustrates another important linkage, namely that marginal reductions in human capital can lead to less produced capital via less gross income.

<sup>4</sup> In the broader economic losses (28.6 percent of total welfare loss), the reduced investment that

and  $PM_{_{IO}}$  concentration changes to each marginal unit of new generation.

While health impacts from air pollution increased monotonically in China between 1975 and 2005, overall economic growth (production and consumption) increased at a faster rate. Thus, the welfare losses from air pollution actually decreased as a percentage share of total welfare over time. In the context of Figure 3 this result makes sense; produced capital (and importantly population!) grew faster between 1975 and 2005 than the decline in human capital caused by the air pollution of those produced assets (coal-fired power infrastructure being the primary air pollution culprit). Despite the large and growing impact on human capital, the analysis suggests that coal-fired power has facilitated a net increase in wealth in China over the past several decades.

To complete the full social cost-benefit analysis of electric power infrastructure, one would need to calculate the [social] costs of the alternatives in addition to the healthrelated benefits, something that the Matus et al. (2012) study leaves to future work. The analysis would also have to include natural capital stock impacts. This includes changes to the country's coal stock, but also impacts on groundwater quality, soil toxicity, and global climate change (a much trickier task since climate is a publicly owned natural asset). Finally, to evaluate proposed changes to China's electricity system we need to take account of the infrastructure systems that will be in place, the pace at which their marginal effects can change (via environmental retrofits or large-scale investments in renewable energy), and, therefore, their net impact on capital stocks over time. If the wealth gains to human and natural capital of a cleaner electricity system outweigh the wealth reductions stemming from higher costs, then a transition away from coal-fired power should be implemented in China.

# 3.2 The capital stock impacts of the High Aswan Dam in Egypt

The global debate about large dams is ... complex because the issues are not confined to the design, construction and operation of dams themselves but embrace the range of social, environmental and political choices on which the human aspiration to development and improved well-being depend. ... at the heart of the dams debate are issues of equity, governance, justice and power - issues that underlie the many intractable problems faced by humanity... The direction we must take is clear. It is to break through the traditional boundaries of thinking and look at these issues from a different perspective, ... [one that] reflects a comprehensive approach to integrating social, environmental and economic dimensions of development. (WORLD COMMIS-SION ON DAMS, 2000)

This is the case when analyzing the High Aswan Dam (HAD) in Egypt, completed in 1971, its purpose to convert the variable and uncertain flows of Nile River water (Figure 4) into a predictable and controllable supply stored in Lake Nasser. It also added 2,100 MW of power generation capacity to the electricity grid. Strzepek et al. (2008) conducted a retrospective study of the economic impacts of the dam, and compared these impacts to the counterfactual scenario of the dam never having been constructed.

The study connected a water resources systems model with a CGE model, which takes water availability as an input into various sectoral production functions (energy: hydropower; transportation: navigation; agriculture: irrigation). The sectoral impacts were estimated as a comparative static simulation using a CGE model of the 1997 Egyptian economy as the comparative year. The stable water supply scenario (with HAD) provided the CGE with the mean of the effectively constant historical releases of HAD from 1971 to 1999, while the variable flow scenario (without HAD) provided the CGE with



#### FIGURE 4

Mean monthly Nile flow at Aswan and growing seasons; the dam resulted in larger and more stable flows in the summer growing season. Reproduced with permission from Strzepek et al. (2008)

an ensemble of 71 years of natural Nile flows that would have entered Egypt. Each year is run independently as a comparative static simulation, and a distribution of model outputs and statistics are calculated.

Interestingly, the study found that stable water supply led to farmers growing lower value crops in the summertime (e.g., rice), such that agricultural production in the summer growing season would have actually been higher without the dam. However, the HAD also facilitated double cropping of winter crops (e.g., wheat), which have lower irrigation requirements and higher market value. Furthermore, without the HAD years of low water flows would have generated negative shocks to the agriculture, energy, transportation, and tourism sectors, though during years of high flows the economy would not have been able to capture the additional benefits. Thus, after accounting for the risk premium of a stable water supply, the total annual net benefit of the dam was estimated to be EGP 7.1 billion (~ US\$1 billion) to EGP 10.3 billion (~ US\$1.5 billion), or 2.7 percent to 4.0 percent of 1997 GDP.

While Strzepek et al. (2008) provide a useful starting point for evaluating the dam, wealthbased evaluation needs to consider impacts of the dam on capital *stocks* in the Egyptian economy, not *flows* of production, and GDP. The dam is unique in that its construction represented a step change to infrastructure capital, as opposed to a marginal change. For the purpose of illustration, however, this doesn't really matter, as our central concern is approximating the capital stock impacts of the dam. Some of these impacts can be extracted from the integrated model of Strzepek et al. (2008), but further literature was consulted in order to get an inclusive assessment of positive and negative effects (ABU ZEID 1990, ROBINSON ET AL. 2008):

- Natural capital:
  - I.2 million acres of new agricultural land owing to increased water availability in the Aswan area
  - Transfer of I million acres from seasonal to perennial irrigation
  - Collapse of Sardine fishery in the Eastern Mediterranean basin
  - Stability of water levels<sup>5</sup> in upstream canals (increase in boat tourism and protection from floods); drop in water levels downstream
  - Rising salinity, coastal erosion, reduction in fertility of Nile Valley

5 Valuing water stock changes is a challenge for a variety of reasons; for a review see Perry (2012).

- Produced capital:
  - 2,100 MW of hydroelectric power capacity
  - Establishment of fertilizer, iron and steel, sugar, plywood, milk, pulp, and paper industries
- Human capital:
  - Population in Aswan area increased from 280,000 in 1960 to 1 million in 1990
  - Increase in schistosomiasis near Lake Nasser resulting from permanent high water levels in canals.

Attributing a perfect causal link from the HAD to any of these impacts is a challenge: for instance, some of the population growth that occurred in the area may have happened without the dam. These impacts - some quantitative, some qualitative - also focus only on physical changes to capital assets. In other words, we need values attached to the changes in order to evaluate whether the dam led to a net gain in wealth, and these values should in principle capture all social benefits and costs to affected stakeholders. Still, while the list certainly indicates important trade-offs and distributive effects that occurred as a result of the dam -Mediterranean fishermen worse off. Aswan rice farmers better off, for example - evidence from the literature seems to suggest that, overall, the dam has led to an increase in the productive base of the Egyptian economy. After all, what if the 2,100 MW of clean hydropower generation were instead fueled by coal?

#### 3.3 Synthesis of the case studies

In both of these cases we reassessed the findings of existing infrastructure studies in terms of their wealth implications. The use of integrated physical and economic models allowed us to capture some of the linkages between capital stocks and estimate subsequent capital impacts. In the case of air pollution in China, we analyzed the marginal impacts of coal-fired electric power infrastructure by linking degraded air quality with health impacts, and therefore human capital losses via premature deaths and foregone wages. For the high Aswan Dam in Egypt, we analyzed a larger set of capital stock impacts, rather than one specific pathway of linkages. While both evaluations could be expanded, the results suggest that integrated modeling can both quantify marginal impacts of infrastructure on individual capital stocks *as well as* approximate aggregate capital stock impacts. The scale and detail of the impacts that are measured will depend on the structure of the models and the nature of the integration.

We chose these cases to illustrate the use of integrated models for measuring capital stock impacts resulting from infrastructure interventions. Our approach represents a step toward the use of inclusive wealth as a framework for infrastructure policy evaluation. However, the cases do not represent the diverse practical circumstances, nor the managerial complexity, inherent in planning and delivering infrastructure projects (LESSARD ET AL. 2014). Yet while the development considerations of a transportation corridor will certainly be different than a hydroelectric dam, we argue that integrated systems modeling applies generally and could be used across sectors and scales to estimate wealth effects. Other infrastructure case studies leveraging integrated models that could inform wealth-based policy evaluation include transportation in Mozambique (ARNDT ET AL. 2012) and the U.S. (SCHAEFER AND JACOBY 2005), energy in South Africa (ARNDT ET AL. 2011), and multiple infrastructure sectors in the UK (HALL ET AL. 2014). Future work in these areas, among others, will help make inclusive wealth more credible as a framework for generating policy insight in infrastructure planning challenges around the world.

In our cases we also had the benefit of hindsight: the "historical" scenario in each was based on data and events that already occurred. The real challenge for policy analysis stems from the evaluation of *prospective* interventions. Central to this challenge is incorporating uncertainty while measuring the impact of interventions on capital stocks. Uncertainty is important because our assumptions about how the future will unfold affect which policies we choose, whether in infrastructure or other sectors. The following section discusses methods for incorporating future uncertainty in models to garner policy-relevant insight during the evaluation of prospective interventions.

## 4. Modeling and managing uncertainty in prospective policy evaluation

Inclusive wealth theory requires an assumption of how the future will unfold. In particular, assumptions about the evolving political economy (DASGUPTA 2009), both under a business as usual (BAU) policy setting, as well as under alternative policies that perturb the BAU, will affect the trajectory of wealth. Yet the future is almost certain to unfold in unanticipated ways, contrary even to the forecasts of experts (STERMAN 1991). Planning for the future must reconcile the reality of uncertainty and the need for thoughtful decisions today. In this section we discuss how various forms of scenario analysis can be used to inform wealth analyses and policy evaluation.

# 4.1 The sensitivity of oil wealth to price uncertainty

In general, when doing policy evaluation we should be concerned with the sources of uncertainty that will have the largest impact on the magnitude and variability of future wealth. Specifically, we need to identify and assess which uncertain parameters can lead to vastly different wealth outcomes through time. Consider the calculation of oil wealth, a form of natural capital, from the 2012 Inclusive Wealth Report (UNU-IHDP AND UNEP 2012, Annex 2); we reproduce and build on this example to illustrate the sensitivity of oil wealth to price uncertainty using scenarios. Oil wealth, Ot, in a particular year is the product of the stock, St, of oil reserves not yet extracted (measured in barrels) and the unit rental price (US\$ per barrel). The rental price is the 19-year average of the inflation-adjusted market price of oil, P, over the horizon 1990 to 2008, net of the unit extraction cost c. The equations are below; dt is the GDP deflator; Xt is the amount of oil extracted.

EQUATION 5

$$O_t = (\overline{P} - c)S_t$$
$$\overline{P} = \frac{1}{19} \sum_{t=1990}^{2008} \frac{P_t d_{base}}{d_t}$$

 $S_t = S_{t-1} - X_t$ 

The movement in oil wealth directly tracks the movement of the oil stock since the rental price is assumed constant at the 19-year market average. We know the oil wealth of countries over this time horizon with relative certainty, since we are using actual or realized price data and known stock values. But what if oil prices had evolved differently over the same horizon? In answering this question we can determine how sensitive oil wealth is to different counterfactual price paths that *could* have occurred in the same time period.

First, we simulate many possible paths using geometric Brownian motion (GBM), a stochastic process frequently used to model stock and commodity prices. The equations for the stochastic process and its effect on the oil price are shown below:  $\mu$  is the percentage drift parameter;  $\sigma$  is the percentage volatility parameter;  $W_t$  is a Wiener process which introduces random variation in the price movement;  $R_t$  is the annual growth in oil price in a particular year. We use a discretized form of the GBM model to generate an evolution of oil prices,  $P_t$ , all from the same initial condition  $P_o$  (the price in 1990).

#### $dR_t = \mu R_t dt + \sigma R_t dW_t$

The chart on the left in Figure 5 shows simulation results for 1000 randomly generated oil

#### **FIGURE 5**

1000 oil price trajectories (1990 to 2008) modeled using GBM.



price scenarios, using values of three percent and ten percent, respectively, for the drift and volatility parameters. The chart on the right in Figure 5 collapses this information by plotting the 5<sup>th</sup>, 50<sup>th</sup> (median) and 95<sup>th</sup> percentile intervals of simulated prices. We compare this percentile forecast with the actual price path (thick black line) over the period from 1990 to 2008.

The region in the chart between the 5<sup>th</sup> percentile (lower dotted line) and the 95<sup>th</sup> percentile (upper dotted line) is the region that captures 90 percent of the price possibilities. In other words, we would expect to see 90 percent of the 1000 simulated price paths in this region at any given point in time. Note that not only was the actual price series quite far from the forecasted median result for most of the horizon, actual prices also fell below the 5<sup>th</sup> percentile level of our simulated results for about half the time. This underscores the point that future outcomes can be drastically different from reasonable forecasts.

We then assess the impact of prices on oil wealth,  $O_{t}$ , following the method above from the 2012 Inclusive Wealth Report. The simulation

calculates  $\overline{P}$  for all 1000 price trajectories and generates a wealth estimate for the oil stock for every year in the 1990 to 2008 horizon. The year-to-year changes in physical oil stocks are the same for all 1000 value estimates; only the price  $\overline{P}$  differs.

Figure 6 represents the distribution of oil wealth in Saudi Arabia in the year 2008, the end of the time horizon. On the left, the histogram represents the range of wealth outcomes, weighted by their frequency. The Value at Risk or Gain (VARG) curve on the right represents the cumulative frequency of the oil wealth estimate. In other words, it conveys information about the total possibility of oil wealth falling above or below a certain wealth threshold. Using the particular GBM formulation discussed above, there is a 30 percent chance that oil wealth could have been below the actual value of US\$2.23 trillion (where the red dashed line intersects the VARG curve). Correspondingly, oil wealth could have exceeded the actual realization by 70 percent for the year 2008.

We can use these simulation results when considering the impact of uncertain oil prices on future oil wealth. Across all scenarios, oil



#### FIGURE 6

The distribution of 2008 oil wealth in Saudi Arabia under oil price uncertainty.

wealth could have ranged from negative US\$2 trillion (cases where oil price is less than extraction cost) to positive US\$15 trillion, even though these extreme cases are much less likely than the central estimate. The range of the simulated wealth possibilities is thus US\$17 trillion with a standard deviation of US\$2.7 trillion. Given that total wealth in Saudi Arabia in 2008 was approximately US\$5 trillion under the baseline calculations in the 2012 report (MUÑOZ ET AL. 2012), our results show that future wealth in countries like Saudi Arabia will be very sensitive to the oil price trajectory. The same goes for other economies whose wealth estimates are tightly linked to fossil fuel prices. However, to conduct a similar analysis on future price trajectories (e.g., 2008 to 2026), the model would need to account for additional complexities not captured in the simple GBM formulation, for example, future extraction rates being a function of the evolving market price.

## 4.2 Using scenario analysis to inform policy choice under uncertainty

The previous analysis described a method for modeling oil price uncertainty. Each of the 1000 price paths represented a particular state of the world that could have unfolded. For planning and evaluating investments in infrastructure, or evaluating policies more generally, we need to understand how policy choices may fare across these alternative futures. Such scenario-based planning techniques for managing uncertainty have frequently been employed in multiple domains<sup>6</sup>.

Generally, the aim is to provide an analytical platform for identifying, assessing, and

<sup>6</sup> While the discussion in this chapter is general, scenario analysis for managing uncertainty and doing long-term planning is a rich field, both topically and methodologically. Some key sources include Wack (1985), van der Heijden (1996), Nakicenovic et al. (2000), and Raskin et al. (2002). For a more recent survey of studies and techniques, see Chapter 4 of KPMG (2012).

choosing among near-term policies based on their long-term performance under conditions of deep uncertainty (LEMPERT ET AL. 2003). For wealth-based policy evaluation, long-term performance can be construed as the extent to which development is sustainable; that is, how fast or slow and in what direction wealth is moving. The conditions of uncertainty could include manifestations of the oil price, as discussed above, the onset of ecological tipping points<sup>7</sup>, or numerous other social, economic, or environmental circumstances that are difficult to predict. Finally, near-term policies could represent investments in renewable energy or public transit systems in the case of infrastructure, but also social or environmental policy like universal healthcare or a carbon tax.

A conceptual framework for using scenarios to inform near-term policy choice is shown in Figure 7. Whereas in the case studies of Section 3 we evaluated just two scenarios (historical and counterfactual), many more could potentially be evaluated using this framework. These could include additional policies (smaller or larger Aswan dams, different locations along the Nile; stricter or more lenient air quality thresholds in China) as well as different futures (upstream Nile water resources development; different population or economic growth rates in China). These additional scenarios would in turn affect the capital stock interactions and thus the marginal impacts of the associated infrastructure. For example, the marginal (and total) effect of coal-fired power on human capital in China will depend on where the population lives, how fast it grows, and the pace of electricity and health technology progress.

A key element of this scenario approach lies in evaluating policy *robustness*. A robust

policy will perform *reasonably well* over the space of plausible futures. This is in contrast to an *optimal* policy, which performs best under a particular set of future conditions. Lempert et al. (2003) summarize the distinction well: "... the best response to deep uncertainty is often a strategy that, rather than being optimized for a particular predicted future, is well-hedged against a variety of different futures". Since sustainable development demands evaluation of policies over multiple decades, the probability of any particular set of circumstances manifesting is exceedingly small. Thus, the search for policy robustness is warranted, and one could argue, more desirable than optimality.

Scenario analysis and similar methods often seek to incorporate more than one value function, or objective (e.g., KEENEY AND RAIFFA 1976), during evaluation as well. In contrast, most of the work in inclusive wealth uses the discounted utility of future consumption as the sole value function of intergenerational wellbeing<sup>8</sup>. This may be appropriate so long as the shadow prices for each capital stock are correct, but this is a challenge in practice (SMULDERS 2012). Furthermore, when evaluating policies particularly those in infrastructure - it is likely that the various stakeholders involved will be differentially concerned with the impacts to capital stocks. Some may be dedicated to minimizing the loss of natural capital (independent of the benefits that may accrue elsewhere) and others with maximizing the gains to human capital. In the case of the High Aswan Dam, Mediterranean Sardine fishermen sought to minimize the loss to downstream fish stocks while Aswan rice farmers sought to increase the development of upstream agriculture stocks, competing objectives as it were. Different capital stock impacts are therefore reflective of different - and often competing - value functions

<sup>7</sup> Natural systems can exhibit "tipping points", which are thresholds across which ecosystems – and possibly dependent socioeconomic systems – exhibit qualitatively different behavior (Lenton et al. 2008). Examples include coral-reef bleaching, eutrophication of lakes, desertification of once arable land, and loss of the tundra permafrost (Lenton et al. 2008, Scheffer et al. 2001), among others.

<sup>8</sup> Ekins (2012) discusses why focusing only on consumption is problematic and reviews other contributors to well-being not widely recognized in the literature.

amongst those affected by the proposed policy change.

In principle, robust policies will perform reasonably well across all of these value functions, in addition to all futures. Searching for policy robustness, however, requires that the resultant outcomes from alternative policies be made explicit, such that trade-offs in impacts across capital stocks and affected stakeholders can be assessed during evaluation. When presented with this trade-off information, policy-makers would be forced to reveal their inherent valuations of the capital stocks upon selecting a policy.

In summary, scenario analysis provides a flexible approach for incorporating future uncertainty and different value functions into policy evaluation. It is particularly appropriate given the long time scales, complex interactions and multi-stakeholder nature of sustainable development planning. In practice, scenario approaches are implementable either in a simulation or optimization framework, the difference being that an optimization framework requires weighting each of the value functions (and the metrics that comprise them) *a priori* to derive optimal policies. Given the diversity of metrics potentially relevant to stakeholders and policy-makers (including the capital assets but also other socioeconomic indicators), as well as the numerous sources of long-term uncertainty in the calculations, a simulation approach is likely to be more illustrative of the trade-offs inherent in policy evaluation.

#### 5. Conclusion and future work

All of the recent work in wealth accounting has focused on assessing the past performance of economies. This chapter contributes to the literature by focusing instead on policy evaluation, in particular how policies impact the trajectories of capital stocks over time. Our focus is on infrastructure capital - a subset of produced capital - because infrastructure choices affect many, if not all, of the capital stocks in appreciable ways, and every national government must plan and manage infrastructure systems. We argue that a systems view of the linked capital stocks in an economy facilitates fuller accounting of wealth impacts. And with the aid of conceptual and mathematical models, we can better capture these systemic linkages to quantify potential impacts.

#### **FIGURE 7**

Wealth-based scenario analysis should search for near-term policies that will perform reasonably well across the landscape of plausible futures and values.



Our framework for evaluating policy impacts, while specific to infrastructure, is general to the broader exercise of policy evaluation. The key element of the framework lies in quantifying the marginal effects of increasing (decreasing) one capital stock on the others. In the case study of air pollution in China, we analyzed the marginal impacts of coal-fired electric power assets by linking degraded air quality with health impacts, and therefore human capital losses via premature deaths and foregone wages. In the case of the high Aswan Dam in Egypt, we analyzed a larger set of capital stock impacts instead of one specific pathway of linkages. Both cases employed counterfactual analysis techniques and integrated physical and economic models. While more comprehensive wealth accounting is needed in each, results indicate that coal-fired power in China and the dam in Egypt have led to (or at the very least facilitated) a net increase in wealth in the respective countries. The analyses also provided preliminary validation for the use of integrated models in doing wealth-based policy evaluation.

Evaluation of prospective policies, whether in infrastructure or other sectors, requires explicit consideration of future uncertainty. This is because assumptions about how the future will unfold - e.g., the evolution of market (or shadow) prices, ecological tipping points, and other socioeconomic factors - can influence projections of capital stock trajectories and therefore our conclusions about whether development will be sustainable. Scenario analysis techniques are useful for modeling and managing uncertainty to inform near-term policy decisions. Furthermore, a thorough scenario approach to wealth-based policy evaluation will consider not just multiple futures, but also evaluate policies based on multiple value functions, whereas inclusive wealth theory currently just focuses on one (the discounted utility of future consumption). Moving forward, evaluation of prospective infrastructure policies will need to leverage integrated models in a way that allows exploration of many futures, many impacts, and thus many scenarios.

There are numerous areas for future work in policy evaluation using inclusive wealth. On a methodological level this includes the following: making total factor productivity a function of infrastructure investment; improved accounting for the positive as well as negative marginal impacts of infrastructure on the other capital stocks; and modifying current models to better estimate the wealth effects of policy interventions (most still focus on the impacts to GDP, not wealth). While the case studies employed particular types of integrated models, we use the term broadly to include any model that captures linkages across the produced, natural, and human world. Such models, regardless of the level of detail, are useful for quantifying the capital stock impacts of alternative policies.

Another area for future work is in conducting additional case studies from various countries of the world in different policy domains. As we demonstrated in the China and Egypt case studies, previous work using integrated models can be usefully reanalyzed in the context of inclusive wealth. Case study analysis can then be made more compelling by examining additional scenarios and their associated wealth impacts.

Methodological modifications and the analysis of additional case studies will go a long way toward furthering the policy-analytic potential of the inclusive wealth framework. This will be beneficial for the practical efforts of the inclusive wealth community, namely, informing the development and implementation of policies commensurate with long-term sustainable development.

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# Annexes

# The inclusive wealth conceptual framework – inclusive wealth and well-being

#### EQUATION 2

We begin by drawing on a revised definition of sustainable development (DASGUPTA AND DURAIAPPAH 2012).

#### **DEFINITION 1**

By sustainable development we mean a pattern of societal development along which (intergenerational) well-being does not decline.

We therefore state that intergenerational wellbeing V(t) is

EQUATION 1

$$V(t) = \int_t^\infty [U(\boldsymbol{\mathcal{C}}(t))e^{-\delta(\tau-t)}]d au$$
 ,  $\delta \ge 0$ 

where C(t) denotes a vector of consumption flows at time *t* and  $\delta$  the discount rate.

Hence U(C(t)) denotes utility (the satisfaction that one enjoys from consuming goods and services) flow at time *t*. This flow of goods and services can range from material goods to services nature provides such as aesthetic gratification or spiritual values, among others.

At any point in time, one can measure how stocks of assets evolve or vary. In doing this one is able to determine the productive base of an economy. Formally, we create an economic forecast by assuming a resource allocation mechanism.

Let K(t) denote a set of vector stocks of capital assets at time *t*. Then for a given K(t), C(t), U(C(t))and together with (I), we can write

1 Equations in this section are adapted from Arrow et al. (2012) and Dasgupta and Duraiappah (2012)

#### V(t) = V(K(t), M, t)

where V(t) denotes intergenerational well-being at t, K(t) denotes a set of vector stocks of capital assets at time t and M denotes an evolving political economy. The proof whereby inter-temporal welfare V(t) will be non-decreasing at time t if and only if the rate of change in inclusive wealth or inclusive investment is non-negative at time t (PEARCE AND ATKINSON 1995, DASGUPTA AND MÄLER 2000).

Differentiating V(t) with respect to t in Equation (2), we obtain:

EQUATION 3

$$d\mathbf{V}(t) = \frac{\Delta \mathbf{V}}{\Delta t} + \sum_{i} [(\Delta \mathbf{V}(t) / \Delta \mathbf{K}_{i}(t)) (d\mathbf{K}_{i}(t) / dt] \ge 0$$

Equation (3) illustrates the criterion for sustainability. In other words it shows the sustainability of an economy's development.

Let  $\Delta V(t)$  represent a small change in V(t), and  $\Delta K_i(t)$  represent a small change in capital asset *i* at time *t*. Using Ddoneefinition I together with Equation (3), we obtain the shadow price of asset *i* at time *t*. The shadow price reflects the marginal value contribution to intergenerational well-being for a unit change in the respective capital asset.

EQUATION 4

$$P_i(t) \equiv \frac{\Delta \mathbf{V}(t)}{\Delta \mathbf{K}_i(t)}$$
 for all i

Let  $\Delta t$  represent a small passage of time following *t*. To measure inclusive wealth we represent Q(t) as the shadow price of time that is

#### $Q(t) = \Delta V(t) / \Delta t$

Q(t) is the shadow price of time as a result of changes in social well-being brought about solely by the passage of time. These are caused by exogenous changes taking place in the economy (e.g., changes in international trade prices over which a country has no control or technological changes that a country enjoys without having been responsible for them). Inclusive wealth therefore treats time as a surrogate for these exogenous events.

Using Equation (6) below we can construct an aggregate index of a country's stock of capital assets by using the shadow prices as weights. This index of a country's stock of capital assets is known as the Inclusive Wealth Index.

EQUATION 6

$$\mathbf{W} = \mathbf{Q}(t)t + \sum_{i} P_i(t)\mathbf{K}_i(t)$$

where  $P_i(t)$  refers to shadow prices of capital assets  $K_i(t)$ . Institutions are reflected in the  $P_i(t)$ 's via M (DASGUPTA AND DURALAPPAH 2012).

An important relationship or linkage exists between changes in inclusive wealth at constant prices and intergenerational well-being (ARROW ET AL. 2012). To show this formally, let  $\Delta$  denote these changes.

EQUATION 7

$$\Delta \mathbf{V}(t) = \left[\frac{\Delta \mathbf{V}(t)}{\Delta t}\right] \Delta t + \sum_{i} \left[\frac{\Delta \mathbf{V}(t)}{\mathbf{K}_{i}(t)}\right] \Delta \mathbf{K}_{i}(t)$$

By using Equations (4) and (5) we can express equation 7 as

EQUATION 8

$$\mathbf{V}(t) = \mathbf{Q}(t)\Delta t + \sum_{i} P_{i}(t)\Delta \mathbf{K}_{i}(t)$$

From our equations derived, (6) and (8) are equivalent stating that the change in well-being is equal to the change in wealth and this is equal to the change in the capital asset base or productive base of a nation.

#### The productive base of a country

The inclusive wealth of a country as illustrated by Equation (6) is typically the total value of the different capital assets a country owns. The typical capital assets commonly known are produced capital, human capital, natural capital, and social capital. As Figure I shows, the way the different capitals are used is to a large extent determined by the evolving cultural and social norms. Similarly, the constituents of human well-being are also determined by evolving cultural norms. The weights society places on these different capital assets are determined by the relative shadow prices of each capital asset category and its respective elements.

Figure 1 also shows that waste produced by the economic system flows back into the ecological system and can cause degradation to these systems, which result in a decline in some of the ecosystem services they provide, which in turn form part of the natural capital asset base. These negative externalities are captured through damage functions on the overall productive base of a country. Damages caused by climate change are an example of such an impact on the inclusive wealth of a country.

The framework developed for the inclusive wealth accounts makes a fundamental difference from many previous studies on natural capital and life supporting systems found in the natural sciences literature. For example, a recent article in Nature on planetary boundaries (ROCKSTROM 2011) points toward natural systems as the key to life-supporting systems, and goes on to identify critical thresholds and tipping points beyond which no changes should occur. We, on the other hand, take a different approach, whereby we suggest that in addition to natural systems, there is also a need for a critical level of human and produced capital for human well-being. We argue that critical levels of each capital asset must be identified as a system working in a holistic framework toward the end goal of improving human well-being. Therefore, in addition to critical levels of natural capital, we argue for critical levels of human



#### FIGURE 1

The productive base and human well-being

(Source: UNU-IHDP and UNEP 2012)

and produced capital necessary for achieving a minimum level of well-being and maintaining those for the present and future generations.

#### Types of capital assets

The list of types of capital assets can be substantially long and in many cases difficult to account for and value. In this paper, we focus on the three main categories of capital known to be relatively large (compared to other forms of capital) and for which relatively good data is available from which to compute changes over time

Produced or manufactured capital (roads, buildings, ports, machinery, equipment) In common parlance, including national accounts, this category essentially represents the entirety of the list of capital assets. When national income accountants and international organizations speak of investment, they usually mean the accumulation of reproducible capital. Reproducible capital is frequently called "manufactured capital."

Human capital (education, skills, tacit knowledge, health) This category is embodied

not transferable without cost from one person to another. Education, skills, and health are ends as well as means. They have intrinsic worth, but are also of indirect value (investment in human capital raises a person's productivity).

in people. Human capital is

Natural capital (local ecosystems, biomes, subsoil resources) Natural capital refers to stocks of nature which produce a range of ecosystem services. Ecosystem services are the benefits ecosystems provide for human well-being. Today it has become broadly accepted that nature should enter explicitly in economic calculations.

There has been much research over the past few decades on social capital. In the inclusive wealth framework, the role of social capital comprised of institutions, culture, and religion are treated as enabling assets: they enable the allocation of goods and services. The present discounted value of social well-being as a function of today's stocks of capital assets reflects the role of this social capital in the allocation of resources (DASGUPTA 2014).

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## ANNEX 2 Methodology

This annex elaborates upon the methodological and data issues regarding the wealth accounts presented in Chapters I and 2. We focus here on the three capital forms (human, produced, and natural capital), as well as the three adjustments to these asset categories: carbon damages, oil capital gains, and total factor productivity.

#### 1. Human capital

For human capital calculations, we followed Arrow et al. (2012) and Klenow and Rodríguez-Clare (1997). According to the method employed in these works, human capital per individual (*h*) can be defined as a function of educational attainment (*Edu*) and the additional compensation over time for this training, which is assumed to be equivalent to the interest rate, ( $\rho$ ), which was fixed at 8.5 percent in this case. Additionally, the amount of human capital per person is assumed to increase exponentially with the interest rate and the average educational attainment per person – consistent with an economy in steady state. Thus it is obtained that:

EQUATION 1

 $h = e^{(Edu.\rho)}$ 

*Edu* is represented by the average years of total schooling per person and is obtained from Barro and Lee (2010), where data are presented every five years for our time period of study (i.e., 1990, 1995, 2000, 2005, and 2010). This implied that we were forced to use linear interpolations for estimating the years of total schooling on an annual basis. There were, however, no estimates for Nigeria in the Barro and Lee (2010) dataset. In this case we estimated the indicator by combining educational attainment parameters from Nigeria, which were available in the Human Development Index database, and the progress made in this indicator by other African countries, particularly Ghana, South Africa, and Kenya, using Barro and Lee (2010). Thus, readers should be particularly cautious when interpreting the human capital trends for Nigeria.

Human capital per capita is further extended by considering the population of the country who achieve the average years of total education. Assuming that individuals engage in formal education at the age of five, then only the portion of the population of the age of 5 plus the average years of the educational attainment,  $P_{5+Edu}$ , or older are considered in the measurement of human capital. Therefore, the total amount of human capital in a country is:

EQUATION 2

 $H=h P_{5+Edu}$ 

As the interest rate ( $\rho$ ) is constant over time, changes in human capital are driven either by a change in the number of people educated, or by an increase or decrease in the average years of education.

The shadow price per unit of human capital,  $SP_{hc}$ , is obtained by computing the present value of the average labor compensation per unit of human capital,  $\bar{r}$ , received by workers over the expected life's working period, *T*, i.e.: EQUATION 3

$$SP_{HC(t)} = \int_{0}^{T(t)} \bar{r} e^{-\delta t} dt$$

The parameter *T* is obtained by using various demographic and socio-economic indicators, such as population, labor force, and mortality rates, which were all considered by age cohort and gender. In our case, these shadow prices were computed for each nation for every year within the time period of 1990 to 2010. Concerning the discount rate,  $\rho$ , it was fixed at

Data sources used in the measurement of human capital

| Variables                              | Data sources  |
|--|---|
| Educational attainment                 | Barro and Lee (2010)  |
| Population by age, gender, time        | United Nations Population Division (2011)   |
| Mortality rates by age, gender, time   | World Health Organization (2012)  |
| Labor force rates by age, gender, time | International Labour Organization (2013)  |
| Market rate of interest                | Klenow and Rodríguez-Clare (1997)   |
| Discount rate                          | It is assumed a rate of 8.5 percent   |
| Employment                             | International Labour Organization (2013)<br>Conference Board (2013)   |
| Compensation of Employees              | United Nations Statistics Division (2012)<br>OECD (2013)<br>Feenstra et al. (2013)<br>Lenzen et al. (2013)<br>Conference Board (2013) |

8.5 percent. Thus, the Human Capital Wealth (*HCW*) of a country is calculated as follows:

$$HCW_{(t)} = e^{(Edu_{(t)} \cdot \rho)} \cdot P_{5+Edu(t)} \cdot \int_{0}^{T_{(t)}} \bar{r} \cdot e^{-\delta \cdot t} dt$$

In order to obtain per capita estimates of human capital, we divided by the total population of the country, *P*, so as to be consistent with the other types of per capita wealth, namely, produced and natural capital. Therefore:

$$\begin{aligned} & HCWper\ capita_{(t)} = \frac{HCW_{(t)}}{P_{(t)}} = \\ & e^{(Edu_{(t)} \cdot \rho)} \cdot \frac{P_{5+Edu(t)}}{P_{(t)}} \underbrace{\cdot \int_{0}^{T_{(t)}} \bar{r} \cdot e^{-\delta \cdot t}.dt}_{\text{Term II}} \\ & \underbrace{\quad \text{Term II}}_{\text{Term III}} \quad \text{Term IIII} \end{aligned}$$

Under this income-based methodological framework, human capital is driven by the above three terms, where: changes in Term 1 depend on changes in the average educational attainment; in the case of Term II, movements are subject to the structure of the country population; and Term III evolves based on years remaining, *T*, for which the average person with human capital will receive monetary compensation, as earnings per unit of human capital,  $\bar{r}$ , are kept constant over time. We decided to capture those changes in the shadow price driven by the parameter *T*, since they relate to aspects of the population, such as mortality rates and participation in the labor market.

The data sources used to compute the calculations in this human capital framework are reported in Table I.

#### 2. Produced capital

With regard to produced capital, we followed the method developed originally by Harberger (1978) and applied, for example, in King and Levine (1994) as well as the recent work done in the Penn World Tables v.8 (FEENSTRA ET AL. 2013), who based their calculations on the perpetual inventory method (PIM) by setting an

Key variables and data sources used in the measurement of produced capital

| Variables         | Data sources                                  |
|-------------------|---|
| Investment        | United Nations Statistics Division (2013a)    |
| Output            | United Nations Statistics Division (2013a)    |
| Depreciation rate | Feenstra et al. (2013)                        |
| Assets lifetime   | It is assumed indefinite depreciation periods |

initial capital estimate. Regarding the initial estimate, *Ko*, the economy is assumed to be in a steady state, implying that the capital-output ratio is constant in the long term, and can be derived as follows:

EQUATION 6

$$k = \frac{1/y}{(\delta + \gamma)}$$

- -

where *k* is the capital-output ratio; 1 is investment; *y* is the output of the economy; *r* is the steady-state growth rate of the economy, estimated as a weighted average growth rate of the economy under study and the global economy (see KING AND LEVINE 1994);  $\delta$  is the depreciation rate of the capital, which is assumed to be 4 percent across countries and time<sup>1</sup>.

This ratio is subsequently multiplied by the output of the economy under study in order to obtain a first estimate of the produced capital stock in the initial period, *Ko*. Our first empirical estimates were carried out for the year 1970 in order to minimize errors in the time period under study (1990 to 2010). As capital depreciates over time, the initial capital estimate retained in 1990 would be about 46 percent, and less than 20 percent in 2010. This way, potential errors in the departure point (year 1970) would be attenuated in the relevant period.

Subsequently, the PIM allows capturing the dynamics in the produced capital accumulation by looking at annual changes in investment.

The corresponding formula of the PIM combined with the initial estimate is:

EQUATION 7

$$K_t = (1 - \delta)^t K_0 + \sum_{j=1}^t I_j (1 - \delta)^{t-j}$$

Finally, regarding the lifetimes of assets, we have assumed indefinite depreciation periods. For further details on this method see, for instance, King and Levine (1994).

#### 3. Natural capital

#### 3.1 Agricultural land

#### 3.1.1 Cropland

The values for this natural capital asset were obtained by analyzing the physical amount of cropland area available every year, and corresponding shadow prices. We used cropland data from the Food and Agriculture Organization of the United Nations (FAO) for calculating the physical changes over the time span in study, 1990 to 2010. For valuing of this asset, we conceptually appealed to the net present value (NPV) of future rental flows. We applied this evaluation on an annual basis so as to obtain the average wealth value per hectare for the entire period of analysis. Once this wealth value per hectare was obtained, we multiplied by the total number of hectares available for cropland in the country. This method is also used in other studies, such as the World Bank (2011). In our method, however, we introduce minor

<sup>1</sup> We arrive to this depreciation rate by taking the country average from Feenstra et al. (2013) for our reference period of 1990 to 2010.

Key variables and data sources used in the measurement of agricultural land

| Variables                   | Data sources                      |
|-----------------------------|-----------------------------------|
| Quantity of crops produced  | FAO (2013)                        |
| Price of crops produced     | FAO (2013)                        |
| Rental Rate                 | Narayanan et. al. (2012)          |
| Harvested area in crops     | FAO (2013)                        |
| Discount rate               | It is assumed a rate of 5 percent |
| Permanent crops land area   | FAO (2013)                        |
| Permanent pasture land area | FAO (2013)                        |

modifications by analyzing a vast number of crops (159) in order to arrive at a representative rental price per hectare for a specific year.

Concretely, we estimated the average rental price per hectare (RPA) for country *i* in year *j* as follows:

EQUATION 8

$$RPA_{ij} = \frac{1}{A} \sum_{k=1}^{159} R_{ik} P_{ijk} Q_{ijk}$$

where *Q*, *P*, and *R* are the quantity of production of crop *k*, (with k=I,2,..,I59), the price per amount of crop *k* and the rental rate of crop *k*, respectively; *j* is the year of analysis, running from 1990 to 2010; and *A* is the total area harvested. To estimate the rental rate by crop group, we mapped FAO crop classification (HS) with those sectoral rental rates provided by Narayanan et al. (2012).

To calculate the value of total wealth per hectare (*Wha*) we estimated the present value of future rental flows as follows:

EQUATION 9

$$Wha_{ij} = \sum_{t=1}^{\infty} \frac{RPA_{ij}}{(1+r)^t}$$

where *r* is the discount rate, assumed to be equal to 5 percent, and *t* is the planning horizon, here assumed to reach infinity. Subsequently, we used the average wealth values per hectare  $(\overline{Wha_i})$  over the study period as a proxy of the shadow price:

**EQUATION 10** 

$$\overline{Wha}_i = \frac{1}{21} \sum_{j=1}^{21} Wha_{ij}$$

Finally, the total wealth in cropland (*WCL*) for country i in year j is derived as follows:

$$WCL_{ii} = \overline{Wha}_i.CLA_{ii}$$

where *CLA* is the physical amount of total cropland area of country *i* in year *j*, while *WCL* is the total wealth of cropland in the corresponding year and nation.

#### 3.1.2 Pastureland

For pastureland, we applied a similar methodology as that used to compute cropland value, as we aimed at obtaining the total wealth per hectare of pastureland and the corresponding physical quantity available during the period of analysis. However, while it is possible to find data for production, prices, and rental rates of the products stemming from this kind of land, it is difficult to link such rents to a particular amount of land involved in the production process (unlike cropland). Given this limitation, we assumed rents per hectare in pastureland to be equal to those of cropland. This also implies that the total wealth per hectare in pastureland is identical to the estimates in the previous section for cropland. Therefore the total wealth in pastureland (*WPL*) was estimated as follows:

$$WPL_{ij} = Wha_i \cdot PLA_{ij}$$

where *PLA* is the physical amount of pastureland area available in the period *j* in country *i*.

#### 3.2 Forest resources

In this version of the IWR accounts we excluded cultivated forest from the forest measures. The main reason for this is that the activity of cultivating a forest, where labor is required to carry out the process, is considered a production activity in the System of National Accounts (UNITED NATIONS 2009). This also implies that the corresponding changes in cultivated forest stock over time are part of gross fixed capital formation and inventories, which can both be treated as inputs in the produced capital model. Cultivated forest has therefore been accounted for under produced capital in the 2014 version of the IWR accounts, while our forest accounts reflect naturally regenerated forest and what is considered within the natural capital accounts.

#### 3.2.1 Timber

As a starting point, we estimated the volume of timber commercially available. This is done by multiplying the total forest area (excluding cultivated forest) by timber density per area and percentage of total volume that is commercially available<sup>2</sup> – all these parameters are countryspecific and were obtained from the Forest Resources Assessment (FAO 2010, FAO 2006, FAO 2001, AND FAO 1995). Unfortunately, parameters regarding volume, area, and density of forest are only available for the following years: 1990, 2000, 2005, and 2010. We therefore carried out linear interpolations to derive estimates for those years without data availability.

With regard to stumpage price, we followed the World Bank's (2006) method by adopting a weighted average price of two different commodities: industrial roundwood and fuelwood, which are also country-specific parameters. The weight attached to the different prices is based on the quantity of the commodity manufactured, while industrial roundwood and fuelwood prices are obtained from the value and quantity exported and produced, respectively. Three further steps were applied regarding the rental price estimates: (I) we converted the annual estimated values from current to constant prices by using each country-specific gross domestic product (GDP) deflator; (2) subsequently, we used information on the regional rental rates for timber estimated by Bolt et al. (2002). Such rates are assumed to be constant over time. (3) we then estimated the average price over the entire study period (1990 to 2010), thereby obtaining our proxy value for the shadow price of timber.

Concerning the estimates of total timber wealth, we multiplied the average rental price over time obtained in the last step by the total volume of timber commercially available every year:

EQUATION 13

Wealth of  $Timber_t = Stock$  commercially available  $t \cdot \overline{Price} \cdot Rental Rate$ 

#### 3.2.2 Non-Timber Forest Benefits

One way to assess ecosystem services (ES) from an asset perspective is by valuing the expected flows of ecological services over time at their marginal contribution to economic welfare (UNITED NATIONS STATISTICS DIVISION 2013B). This can be represented by the following formulation:

$$ESW_t = \int_t^T P_t \cdot (Q_t \cdot r_t) \cdot e^{-\delta \cdot t} \cdot dt$$

<sup>2</sup> This parameter is only available for the naturally generated forest and cultivated forest together. Therefore, we assumed that this coefficient also applies to only naturally regenerated forest.

Average annual value per hectare in US\$ of 2005 on the basis of 262 values.

|    | Select service        | Temperate and boreal<br>forests<br>Unit = USD/yr/ha | Tropical forest<br>Unit = USD/yr/ha |  |  |
|----|-----------------------|---|-------------------------------------|--|--|
|    | Provisioning services |   |                                     |  |  |
| 1  | food                  | 23  | 107                                 |  |  |
| 2  | water                 | 146   | 137                                 |  |  |
| 3  | genetic               | 2 451   |                                     |  |  |
| 4  | medical               |   | 475                                 |  |  |
| 5  | raw materials         |   |                                     |  |  |
| 6  | ornamental            |   |                                     |  |  |
|    | Regulating services   |   |                                     |  |  |
| 7  | air quality           | 868   | 223                                 |  |  |
| 8  | climate               |   |                                     |  |  |
| 9  | extreme events        | 0   | 33                                  |  |  |
| 10 | water flows           | 2   | 14                                  |  |  |
| 11 | waste                 | 40  | 343                                 |  |  |
| 12 | erosion               | 1 342   |                                     |  |  |
| 13 | soil fertility        | 37 129  |                                     |  |  |
| 14 | pollination           | 418 54  |                                     |  |  |
| 15 | bio control           | 20 13   |                                     |  |  |
|    | Habitat services      |   |                                     |  |  |
| 16 | nursery               |   | 17                                  |  |  |
| 17 | genepool              | 506   | 396                                 |  |  |
|    | Cultural services     |   |                                     |  |  |
| 18 | aesthetic             |   |                                     |  |  |
| 19 | recreation            | 27  | 257                                 |  |  |
| 20 | inspiration           | 0   |                                     |  |  |
| 21 | spiritual             |   |                                     |  |  |
| 22 | cognitive             | 0   |                                     |  |  |
|    | Total                 | 2,091   | 2,990                               |  |  |

Source: Van der Ploeg and de Groot (2010).

Key variables and data sources used in the measurement of forest wealth

| Variables   | Data sources                                   |
|---|--|
| Forest stocks   | FAO (2010), FAO (2006), FAO (2001), FAO (1995) |
| Forest stock commercially available                       | FAO (2006)                                     |
| Wood production   | FAO (2013)                                     |
| Value of wood production                                  | FAO (2013)                                     |
| Rental rate   | Bolt et al. (2002)                             |
| Forest area   | FAO (2013)                                     |
| Value of non-timber forest benefits (NTFB)                | Van der Ploeg and de Groot (2010)              |
| Percentage of forest area used for the extraction of NTFB | World Bank (2006)                              |
| Discount rate   | It is assumed a rate of 5 percent              |

where:

*ESW:* ecosystem service wealth

 $\delta$ : discount rate, which is here assumed to be fixed at 5 percent.

*T:* planning horizon. We assume here infinite periods.

*t:* year under analysis. We carry out these

estimates for the time period between 1990 and 2010.

 $P_{t}$ : marginal contribution of the ES flows to inter-temporal economic welfare. In empirical studies, the monetary value of a hectare per year (USD/ha/yr) is often used. For example, the World Bank (2011) and Arrow et al. (2012), based on the work of Lampietti and Dixon (1995), estimated this parameter at around 150 US\$/ha/yr. In this version of the IWR accounts we updated this parameter using the Ecosystem Service Valuation Database, ESVD, (VAN DER PLOEG AND DE GROOT, 2010). The average value per hectare varies considerably for temperate and boreal forest, and tropical forest as shown in Table 4. Since the ESVD presents information for these two types of forests, we weighted the corresponding values by the share of each forest type in the total forest of the country in order to arrive to the final value of the benefits per hectare and year.

*Q<sub>t</sub>*: total forest area in the country under analysis excluding cultivated forest.

 $r_i$ : the fraction of the total forest area which is accessed by individuals to obtain benefits. Term  $Q_t . r_t$  reflects therefore the area of the forest in interaction with the population and contributing to economic welfare at time t. The ES literature has made emphasis on the idea of considering only the portion of the ecosystem services contributing to economic welfare, in contrast to taking into account the whole asset area (see, for example, TALLIS ET AL. 2012). We here assume that this value is 10 percent as in the work of the World Bank (2006).

#### 3.3 Fossil fuels

In our analysis, we followed the approach used by Arrow et al. (2012). The methodology of valuing the wealth of these stated components largely follows the same procedure.

We referred to BP Statistical Review of World Energy (BP 2013) for prices of coal, natural gas, and oil. For coal, we averaged prices from four sources: the United States, northwestern Europe, Japan coking, and Japan steam. For natural gas, we averaged prices from five sources: the European Union (EU), United Kingdom, the United States, Japan, and Canada. Lastly, we averaged the prices of four types of oil grades:

Key variables and data sources used in the measurement of fossil fuels

| Variables   | Data sources                                  |
|-------------|---|
| Reserves    | U.S. Energy Information Administration (2013) |
| Production  | U.S. Energy Information Administration (2013) |
| Prices      | BP (2013)                                     |
| Rental rate | Narayanan et al. (2012)                       |

Dubai, Brent, Nigerian Forcados, and West Texas Intermediate. We adjusted for inflation before averaging over time using the U.S. GDP deflator. We obtained the rental prices by multiplying the above estimated prices and the corresponding sectoral rental rates from Narayanan et al. (2012).

We set the end of year reserves of natural gas and oil to 2010 and obtained this dataset from the U.S. Energy Information Administration (2013). In the case of coal, the reference year used was 2008; the data was also obtained from the U.S. Energy Information Administration.

The stocks of natural gas, oil, and coal for a year previous to 2010 were estimated as follows: EQUATION 15

#### $Stock_{t-1} = Stock_t + Production_t$

where the corresponding stock under study in year *t*-*i* is derived from the production and the stock in year *t*. Finally, we computed the wealth of resource *i* (coal, natural gas, and oil) by multiplying the stocks and the unit rental price for each of our 140 countries for the period under study:

#### EQUATION 16

Wealth of  $Resource_t 'i' = Stock_{ti} \cdot \overline{Price_{ti}} \cdot Rental Rate_i$ 

#### 3.4 Metals and minerals

In order to value metals and minerals, we followed the method used by Arrow et al. (2012). We set the reserves base to 2010 and obtained reserves data from the U.S. Geological Survey published in their Mineral Commodity Summaries and/or Minerals Yearbooks (U.S. GEOLOGICAL SURVEY 2013A). We focused on ten mineral types: bauxite, copper, gold, iron, lead, nickel, phosphate, silver, tin, and zinc.

Production data are based on United States Geological Survey numbers, published in the Mineral Commodity Summaries and/ or Minerals Yearbook (U.S. GEOLOGICAL SURVEY 2013A-B), and from the World Mineral Statistics Archive, contributed by permission of the British Geological Survey (BRITISH GEOLOGICAL SURVEY 2013). We filled in the missing years by extrapolating linearly.

Previous years' stocks were calculated by using the following equation:

## $Stock_{t-1} = Stock_t + Production_t$

where the production and the stock in year *t* are used to compute the amount of the mineral available in year *t*-*t*.

For prices, we used world annual market prices for the ten mineral commodities from the World Bank (2013) and U.S. Geological Survey (2013A-B) for the period of 1990 to 2010. We converted to year 2005 constant prices and computed average prices for each mineral. As with fossil fuels, we retrieved sectoral rental rates of different mineral industries from Narayanan et al. (2013) and multiplied them by the corresponding prices.

Key variables and data sources used in the measurement of minerals

| Variables   | Data sources  |
|-------------|---|
| Reserves    | U.S. Geological Survey (2013a)                                      |
| Production  | U.S. Geological Survey (2013a) and British Geological Survey (2013) |
| Prices      | U.S. Geological Survey (2013–b) and World Bank (2013)               |
| Rental rate | Narayanan et al. (2012)   |

We finally valued minerals by multiplying mineral stocks by rental prices to obtain the total mineral wealth for the period under study:

Wealth of Resource<sub>t</sub> 'i' = Stock<sub>ti</sub> ·  $\overline{Price_{ti}}$  · Rental Rate<sub>i</sub>

#### 4. Adjustments

#### 4.1 Carbon damages

Carbon damage estimates are based on the method developed in Arrow et al. (2012). The key methodological steps can be described as follows: (I) obtain the total global carbon emissions for the period under analysis, 1990 to 2010; (2) derive the total global damages as a function of the emissions; and (3) allocate the global damages to the countries according to the potential effect of global warming in their economies.

**Global carbon emissions:** Two sources of carbon emissions were taken into account: (i) carbon emissions stemming from fuel consumption and cement, which were obtained from the Carbon Dioxide Information Analysis Center (BODEN ET AL. 2011); and (ii) emissions resulting from global deforestation. In this case, we used FAO (2013) data on the changes in annual global forest land. It is further estimated that the average carbon release per hectare is equal to 100 tonnes of carbon (LAMPIETTI AND DIXON 1995).

**Global carbon damages:** The damages per tonne of carbon released to the atmosphere are estimated at US\$50 (see TOL 2009). By multiplying the total amount of global tons of carbon released to the atmosphere by the price per ton, we obtain the total global carbon damages. Note that this parameter is constant over time.

Assigning carbon damages to countries: To calculate the distribution of the damages that each region suffers, we referred to the study of Nordhaus and Boyer (2000). This study presents the distribution of damages which different regions and the global economy as a whole will suffer as a percentage of the corresponding regional and global GDP. By using country and global GDP information, we were able to reestimate regional percentage damages in terms of the total global GDP – and not related to the country GDP – as initially presented in Nordhaus and Boyer (2000). Finally, we apportioned the global damages estimated in previous steps two according to this latter percentage.

#### TABLE 8

Key variables and data sources used in the measurement of carbon damages

| Variables                               | Data sources                                  |  |  |
|---|---|--|--|
| Carbon emission                         | Boden et al. (2011)                           |  |  |
| Forest area                             | FAO (2013)                                    |  |  |
| Carbon release per<br>hectare of forest | Lampietti and Dixon<br>(1995)                 |  |  |
| Carbon cost                             | Tol (2009)                                    |  |  |
| Climate change impacts                  | Nordhaus and Boyer<br>(2000)                  |  |  |
| GDP                                     | United Nations Statistics<br>Division (2013a) |  |  |

#### 4.2 Oil capital gains

As noted in Chapter I, gains in oil prices are separately accounted for in the wealth accounts. In order to include this adjustment, we assumed an annual increase of 5 percent in the rental price of oil, which corresponds to the annual average oil price increase during the years 1990 and 2010 (BP 2013). These increments in the rental price of oil are multiplied by the stock of oil available in each period. Data on oil stock rely on the method presented above in Section 3.3.

Conversely, countries that depend on oil imports may be negatively affected as their capacity to build other capital forms is impacted by higher prices. We therefore allocate those gains in oil prices to those nations that consume the commodity. To do so, we used data on oil consumption from our country sample as well as total world oil consumption. We were thus able to estimate the way in which the oil capital gains have to be distributed among the countries in this study. Finally, we subtracted the oil capital gains from the losses due to oil consumption, thereby obtaining the net oil capital gains/losses.

## 4.3 Total factor productivity measurement considering natural capital

We used a deterministic nonparametric analysis called Malmquist Productivity Index, based on the data envelopment analysis (see review for FÄRE ET AL. 1994, KERSTENS AND MANAGI 2012). The index based on distance function is suitable for assessing the relation between multivariate inputs and outputs. In addition, the measurement takes into account the efficiency of resource use and productivity changes. Using the distance function specification for the index, we can formulate our problem as follows:

#### EQUATION 19

$$T(t) = \left\{ \left( x_t, y_t \right) : x_t \text{ can produce } y_t \right\}.$$

Let  $x = (x^1, ..., x^M) \in R^M_+$  and  $y = (y^1, ..., y^N) \in R^N_+$ be the input and output vectors, respectively. The technology set, defined by Equation (19), consists of all feasible input vectors,  $x_t$ , and output vectors,  $y_t$ , at time t, and satisfies certain axioms that are sufficient to define meaningful distance functions. Inputs used in this case are produced, human, and natural capital, and outputs are GDP. Compared to previous studies on inclusive wealth TFP computation, we explicitly added natural capital as an additional input, along with the two other conventional factors commonly taken into account, produced and human capital. The distance function is defined as

#### EQUATION 20

$$d(x_t, y_t) = \max\{\delta; (x_t, y_t / \delta) \in T(t)\}$$

where  $\delta$  is the maximal proportional amount to which  $y_t$  can be expanded, given technology T(t). In this analysis, we characterize production technology as having constant returns to scale. This formulation produces an output-oriented distance function. Equation (21 ) is a formulation of the Malmquist Productivity Index (*M* as *TFP* to inclusive investment), as follows:

$$M\left(GDP_{i,t}, H_{i,t}, P_{i,t}, N_{i,t}, GDP_{i,t+1}, H_{i,t+1}, P_{i,t+1}, N_{i,t+1}\right)$$

$$= \left[\frac{d^{t}\left(GDP_{i,t+1}, H_{i,t+1}, P_{i,t+1}, N_{i,t+1}\right)}{d^{t}\left(GDP_{i,t}, H_{i,t}, P_{i,t}, N_{i,t}\right)} \times \frac{d^{t+1}\left(GDP_{i,t+1}, H_{i,t+1}, P_{i,t+1}, N_{i,t+1}\right)}{d^{t+1}\left(GDP_{i,t}, H_{i,t}, P_{i,t}, N_{i,t}\right)}\right]^{1/2}$$

where *d* is the geometric distance to the production frontier caused by production inefficiency, while the frontier denote the best available technology from the given inputs and outputs; *i* refers to the country under analysis, running *i* from I up to 140 nations in our sample; *GDP* is the corresponding value of gross domestic product; *H* stands for human capital; *P* represents produced capital; *N* represents natural capital. Thus we capture the productivity change from the variations in inefficiencies between two years. This methodology is widely used in the measurement of productivity (see, for example, TANAKA AND MANAGI 2013).

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### **ANNEX 3**

## Data

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#### Wealth

|    | Country/Year                     | 1990       | 1995       | 2000       | 2005       | 2010       |
|----|----------------------------------|------------|------------|------------|------------|------------|
| 1  | Afghanistan                      | 73,403     | 94,802     | 103,928    | 127,882    | 151,524    |
| 2  | Albania                          | 114,564    | 109,064    | 107,980    | 117,338    | 129,651    |
| 3  | Algeria                          | 1,167,108  | 1,221,564  | 1,267,935  | 1,319,205  | 1,409,174  |
| 4  | Argentina                        | 2,371,915  | 2,488,599  | 2,667,146  | 2,881,785  | 3,129,905  |
| 5  | Armenia                          | 68,530     | 64,309     | 63,353     | 66,322     | 73,326     |
| 6  | Australia                        | 8,264,944  | 8,722,411  | 9,290,457  | 10,174,447 | 11,484,564 |
| 7  | Austria                          | 2,720,433  | 3,048,956  | 3,215,545  | 3,508,807  | 3,729,039  |
| 8  | Bahrain                          | 59,209     | 72,000     | 83,824     | 99,768     | 178,028    |
| 9  | Bangladesh                       | 446,055    | 509,427    | 613,828    | 719,702    | 832,121    |
| 10 | Barbados                         | 55,845     | 57,386     | 59,594     | 64,533     | 66,562     |
| 11 | Belgium                          | 3,427,027  | 3,698,778  | 3,949,349  | 4,297,522  | 4,532,762  |
| 12 | Belize                           | 19,447     | 20,430     | 21,568     | 23,024     | 24,224     |
| 13 | Benin                            | 74,266     | 78,537     | 85,076     | 93,842     | 107,053    |
| 14 | Bolivia (Plurinational State of) | 1,014,643  | 1,007,867  | 1,007,083  | 1,005,307  | 999,736    |
| 15 | Botswana                         | 105,004    | 114,238    | 122,809    | 135,049    | 146,694    |
| 16 | Brazil                           | 12,345,895 | 13,480,646 | 14,609,964 | 15,579,794 | 16,439,769 |
| 17 | Bulgaria                         | 376,510    | 363,389    | 353,531    | 359,804    | 386,815    |
| 18 | Burundi                          | 21,567     | 23,423     | 23,845     | 27,366     | 31,548     |
| 19 | Cambodia                         | 122,363    | 123,940    | 126,859    | 131,797    | 139,011    |
| 20 | Cameroon                         | 380,359    | 386,601    | 399,149    | 410,420    | 430,305    |
| 21 | Canada                           | 13,181,342 | 13,630,760 | 14,449,766 | 16,006,849 | 17,109,382 |
| 22 | Central African Republic         | 188,370    | 188,007    | 187,459    | 186,939    | 186,578    |
| 23 | Chile                            | 1,021,418  | 1,120,579  | 1,255,217  | 1,417,214  | 1,652,052  |
| 24 | China                            | 18,571,020 | 20,600,640 | 23,025,303 | 26,122,801 | 31,969,803 |
| 25 | Colombia                         | 2,423,524  | 2,638,037  | 2,848,572  | 3,022,557  | 3,315,723  |
| 26 | Congo                            | 269,412    | 270,155    | 266,719    | 263,506    | 267,092    |
| 27 | Costa Rica                       | 194,581    | 218,719    | 259,570    | 298,885    | 344,774    |
| 28 | Côte d'Ivoire                    | 222,704    | 252,194    | 274,679    | 297,501    | 315,782    |
| 29 | Croatia*                         | 632,797    | 655,601    | 656,831    | 690,218    | 729,929    |
| 30 | Cuba                             | 381,691    | 383,320    | 387,377    | 399,260    | 415,882    |
| 31 | Cyprus                           | 156,160    | 179,633    | 209,279    | 233,910    | 272,687    |
| 32 | Czech Republic**                 | 1,282,725  | 1,357,598  | 1,417,312  | 1,548,996  | 1,635,440  |
| 33 | Democratic Republic of the Congo | 1,119,619  | 1,133,175  | 1,140,858  | 1,159,528  | 1,184,607  |
| 34 | Denmark                          | 2,353,061  | 2,363,554  | 2,454,732  | 2,608,616  | 2,799,237  |
| 35 | Dominican Republic               | 263,419    | 300,769    | 339,221    | 387,839    | 443,651    |
| 36 | Ecuador                          | 429,445    | 446,332    | 454,320    | 471,343    | 500,857    |
| 37 | Egypt                            | 727,607    | 794,660    | 896,349    | 1,004,425  | 1,178,805  |

\* Data for this country are available from 1991, thus all data refer to 1991 instead of 1990. \*\* Data for this country are available from 1992, thus all data refer to 1992 instead of 1990.
Inclusive wealth change percentage change with respect to the base year, 1990

Inclusive wealth per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 5,632   | 4,790   | 4,547   | 4,631   | 4,824   | -15  | -19  | -18  | -14  |
| 34,827  | 34,722  | 35,152  | 37,347  | 40,462  | 0    | 1    | 7    | 16   |
| 46,132  | 43,178  | 41,526  | 40,111  | 39,731  | -6   | -10  | -13  | -14  |
| 72,664  | 71,398  | 72,220  | 74,501  | 77,449  | -2   | -1   | 3    | 7    |
| 19,333  | 19,952  | 20,595  | 21,632  | 23,714  | 3    | 7    | 12   | 23   |
| 483,439 | 481,411 | 484,778 | 498,661 | 515,734 | 0    | 0    | 3    | 7    |
| 354,661 | 384,211 | 401,707 | 426,230 | 444,269 | 8    | 13   | 20   | 25   |
| 120,126 | 128,785 | 131,346 | 137,648 | 141,087 | 7    | 9    | 15   | 17   |
| 4,238   | 4,336   | 4,737   | 5,119   | 5,596   | 2    | 12   | 21   | 32   |
| 215,202 | 217,855 | 222,772 | 238,566 | 243,520 | 1    | 4    | 11   | 13   |
| 344,459 | 366,947 | 388,116 | 412,659 | 423,145 | 7    | 13   | 20   | 23   |
| 102,222 | 92,730  | 86,096  | 81,950  | 77,734  | -9   | -16  | -20  | -24  |
| 15,559  | 13,897  | 13,053  | 12,293  | 12,097  | -11  | -16  | -21  | -22  |
| 152,384 | 134,905 | 121,229 | 109,910 | 100,680 | -11  | -20  | -28  | -34  |
| 75,959  | 72,045  | 69,860  | 72,001  | 73,093  | -5   | -8   | -5   | -4   |
| 82,498  | 83,292  | 83,761  | 83,768  | 84,330  | 1    | 2    | 2    | 2    |
| 42,693  | 43,485  | 44,157  | 46,492  | 51,614  | 2    | 3    | 9    | 21   |
| 3,850   | 3,848   | 3,741   | 3,774   | 3,763   | 0    | -3   | -2   | -2   |
| 12,837  | 11,097  | 10,192  | 9,867   | 9,832   | -14  | -21  | -23  | -23  |
| 31,226  | 27,733  | 25,459  | 23,381  | 21,956  | -11  | -18  | -25  | -30  |
| 475,846 | 465,180 | 471,177 | 495,823 | 502,972 | -2   | -1   | 4    | 6    |
| 64,185  | 56,497  | 50,643  | 46,527  | 42,394  | -12  | -21  | -28  | -34  |
| 77,452  | 77,767  | 81,403  | 86,936  | 96,534  | 0    | 5    | 12   | 25   |
| 16,216  | 16,969  | 18,143  | 19,978  | 23,834  | 5    | 12   | 23   | 47   |
| 72,990  | 72,367  | 71,637  | 70,226  | 71,622  | -1   | -2   | -4   | -2   |
| 112,776 | 98,860  | 85,057  | 74,580  | 66,064  | -12  | -25  | -34  | -41  |
| 63,377  | 63,051  | 66,231  | 69,356  | 74,003  | -1   | 5    | 9    | 17   |
| 17,791  | 17,183  | 16,565  | 16,509  | 15,999  | -3   | -7   | -7   | -10  |
| 139,135 | 140,416 | 145,783 | 155,385 | 165,767 | 1    | 5    | 12   | 19   |
| 36,110  | 35,164  | 34,885  | 35,476  | 36,941  | -3   | -3   | -2   | 2    |
| 203,688 | 209,993 | 221,860 | 226,533 | 247,078 | 3    | 9    | 11   | 21   |
| 124,371 | 131,559 | 138,370 | 151,556 | 155,861 | 6    | 11   | 22   | 25   |
| 30,753  | 25,715  | 22,989  | 20,194  | 17,958  | -16  | -25  | -34  | -42  |
| 457,702 | 451,632 | 459,731 | 481,344 | 504,354 | -1   | 0    | 5    | 10   |
| 36,613  | 37,994  | 39,481  | 41,864  | 44,690  | 4    | 8    | 14   | 22   |
| 41,854  | 39,205  | 36,802  | 35,106  | 34,626  | -6   | -12  | -16  | -17  |
| 12,800  | 12,804  | 13,250  | 13,536  | 14,531  | 0    | 4    | 6    | 14   |

#### Wealth

|    | Country/Year                     | 1990       | 1995       | 2000       | 2005       | 2010       |
|----|----------------------------------|------------|------------|------------|------------|------------|
| 38 | El Salvador                      | 124,670    | 141,924    | 158,415    | 179,374    | 197,736    |
| 39 | Estonia                          | 155,224    | 151,263    | 158,574    | 171,436    | 187,151    |
| 40 | Fiji                             | 31,323     | 35,706     | 37,120     | 39,495     | 42,475     |
| 41 | Finland                          | 1,964,452  | 2,047,532  | 2,054,868  | 2,269,919  | 2,423,299  |
| 42 | France                           | 19,443,346 | 20,942,433 | 22,745,060 | 24,694,656 | 26,686,007 |
| 43 | Gabon                            | 237,028    | 242,441    | 249,860    | 258,767    | 273,794    |
| 44 | Gambia                           | 5,993      | 6,871      | 7,516      | 8,984      | 10,772     |
| 45 | Germany                          | 25,747,425 | 28,904,348 | 30,993,376 | 34,922,304 | 35,855,483 |
| 46 | Ghana                            | 160,056    | 171,435    | 183,337    | 199,512    | 227,090    |
| 47 | Greece                           | 1,809,046  | 1,938,570  | 2,054,842  | 2,207,873  | 2,455,237  |
| 48 | Guatemala                        | 277,525    | 311,995    | 355,921    | 402,973    | 471,936    |
| 49 | Guyana                           | 183,697    | 184,445    | 185,345    | 185,892    | 187,665    |
| 50 | Haiti                            | 33,386     | 38,513     | 44,518     | 49,811     | 56,231     |
| 51 | Honduras                         | 171,768    | 179,487    | 194,201    | 208,394    | 235,566    |
| 52 | Hungary                          | 1,146,512  | 1,232,464  | 1,292,072  | 1,389,011  | 1,425,079  |
| 53 | Iceland                          | 192,428    | 202,233    | 213,916    | 232,418    | 242,865    |
| 54 | India                            | 9,287,027  | 10,195,737 | 11,502,578 | 12,996,926 | 15,088,491 |
| 55 | Indonesia                        | 4,197,870  | 4,434,701  | 4,776,139  | 5,044,358  | 5,440,254  |
| 56 | Iran (Islamic Republic of)       | 4,486,366  | 4,632,574  | 4,850,363  | 5,143,439  | 5,241,747  |
| 57 | Iraq                             | 1,183,949  | 1,191,866  | 1,189,261  | 1,189,287  | 1,198,590  |
| 58 | Ireland                          | 1,126,848  | 1,219,540  | 1,436,974  | 1,741,586  | 1,925,414  |
| 59 | Israel                           | 927,772    | 1,172,166  | 1,393,529  | 1,562,390  | 1,816,553  |
| 60 | Italy                            | 15,739,344 | 16,408,562 | 17,141,802 | 18,613,646 | 19,661,610 |
| 61 | Jamaica                          | 141,185    | 156,950    | 167,812    | 177,966    | 186,801    |
| 62 | Japan                            | 44,161,278 | 48,711,574 | 51,895,150 | 53,747,667 | 54,693,320 |
| 63 | Jordan                           | 102,394    | 141,480    | 158,391    | 177,113    | 212,282    |
| 64 | Kazakhstan*                      | 1,492,558  | 1,500,502  | 1,473,015  | 1,482,121  | 1,543,152  |
| 65 | Kenya                            | 215,582    | 255,182    | 285,111    | 331,645    | 387,396    |
| 66 | Kuwait                           | 1,433,968  | 1,399,021  | 1,390,094  | 1,394,859  | 1,422,032  |
| 67 | Kyrgyzstan*                      | 34,661     | 35,631     | 37,162     | 38,081     | 42,871     |
| 68 | Lao People's Democratic Republic | 135,058    | 134,923    | 136,633    | 139,337    | 147,464    |
| 69 | Latvia                           | 173,979    | 166,434    | 169,806    | 191,803    | 206,576    |
| 70 | Lesotho                          | 19,976     | 23,651     | 27,375     | 29,134     | 31,567     |
| 71 | Liberia                          | 37,984     | 36,642     | 36,694     | 36,147     | 37,005     |
| 72 | Lithuania*                       | 271,569    | 269,588    | 272,111    | 292,331    | 311,714    |
| 73 | Luxembourg                       | 178,050    | 197,065    | 227,539    | 263,673    | 318,999    |
| 74 | Malawi                           | 45,907     | 45,908     | 47,559     | 50,435     | 56,302     |

Inclusive wealth change percentage change with respect to the base year, 1990

Inclusive wealth per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 23,378  | 24,754  | 26,668  | 29,646  | 31,929  | 6    | 14   | 27   | 37   |
| 99,018  | 104,992 | 115,684 | 127,381 | 139,546 | 6    | 17   | 29   | 41   |
| 43,004  | 46,034  | 45,730  | 48,015  | 49,354  | 7    | 6    | 12   | 15   |
| 393,959 | 400,864 | 397,201 | 432,832 | 451,725 | 2    | 1    | 10   | 15   |
| 342,866 | 362,047 | 385,197 | 404,854 | 425,022 | 6    | 12   | 18   | 24   |
| 255,131 | 222,969 | 202,271 | 188,780 | 181,867 | -13  | -21  | -26  | -29  |
| 6,203   | 6,101   | 5,795   | 5,975   | 6,232   | -2   | -7   | -4   | 0    |
| 325,513 | 352,796 | 376,366 | 423,092 | 435,655 | 8    | 16   | 30   | 34   |
| 10,819  | 10,086  | 9,566   | 9,220   | 9,310   | -7   | -12  | -15  | -14  |
| 178,047 | 181,653 | 187,027 | 197,429 | 216,142 | 2    | 5    | 11   | 21   |
| 31,102  | 31,151  | 31,674  | 31,687  | 32,799  | 0    | 2    | 2    | 5    |
| 253,400 | 253,365 | 252,823 | 249,106 | 248,731 | 0    | 0    | -2   | -2   |
| 4,686   | 4,889   | 5,149   | 5,329   | 5,627   | 4    | 10   | 14   | 20   |
| 35,131  | 32,194  | 31,231  | 30,293  | 30,993  | -8   | -11  | -14  | -12  |
| 110,493 | 119,293 | 126,543 | 137,704 | 142,741 | 8    | 15   | 25   | 29   |
| 755,232 | 756,075 | 760,700 | 783,229 | 758,631 | 0    | 1    | 4    | 0    |
| 10,628  | 10,571  | 10,914  | 11,400  | 12,321  | -1   | 3    | 7    | 16   |
| 22,772  | 22,240  | 22,382  | 22,192  | 22,680  | -2   | -2   | -3   | 0    |
| 81,763  | 77,523  | 74,230  | 73,760  | 70,860  | -5   | -9   | -10  | -13  |
| 68,146  | 58,747  | 49,849  | 43,469  | 37,844  | -14  | -27  | -36  | -44  |
| 319,110 | 337,733 | 377,775 | 418,848 | 430,751 | 6    | 18   | 31   | 35   |
| 206,174 | 219,833 | 231,677 | 236,562 | 244,871 | 7    | 12   | 15   | 19   |
| 276,943 | 288,031 | 300,806 | 317,254 | 324,712 | 4    | 9    | 15   | 17   |
| 59,700  | 63,761  | 65,002  | 66,364  | 68,149  | 7    | 9    | 11   | 14   |
| 361,234 | 391,299 | 412,783 | 425,243 | 432,236 | 8    | 14   | 18   | 20   |
| 29,979  | 32,287  | 32,813  | 33,155  | 34,310  | 8    | 9    | 11   | 14   |
| 90,384  | 94,218  | 98,485  | 97,689  | 96,288  | 4    | 9    | 8    | 7    |
| 9,194   | 9,304   | 9,122   | 9,312   | 9,562   | 1    | -1   | 1    | 4    |
| 686,869 | 859,397 | 716,253 | 616,100 | 519,609 | 25   | 4    | -10  | -24  |
| 7,804   | 7,759   | 7,500   | 7,552   | 8,037   | -1   | -4   | -3   | 3    |
| 32,215  | 28,138  | 25,697  | 24,218  | 23,781  | -13  | -20  | -25  | -26  |
| 65,310  | 66,785  | 71,199  | 83,192  | 91,727  | 2    | 9    | 27   | 40   |
| 12,187  | 13,179  | 13,939  | 14,104  | 14,538  | 8    | 14   | 16   | 19   |
| 17,857  | 17,493  | 12,888  | 11,358  | 9,265   | -2   | -28  | -36  | -48  |
| 73,435  | 74,285  | 77,745  | 85,583  | 93,788  | 1    | 6    | 17   | 28   |
| 467,048 | 483,569 | 522,489 | 576,715 | 628,634 | 4    | 12   | 23   | 35   |
| 4,894   | 4,645   | 4,235   | 3,933   | 3,778   | -5   | -13  | -20  | -23  |

#### Wealth

Inclusive wealth in millions of constant 2005 US\$

|     | Country/Year        | 1990       | 1995       | 2000       | 2005       | 2010       |
|-----|---------------------|------------|------------|------------|------------|------------|
| 75  | Malaysia            | 1,287,401  | 1,542,098  | 1,776,498  | 1,954,619  | 2,185,440  |
| 76  | Maldives            | 5,207      | 6,504      | 8,411      | 10,671     | 14,041     |
| 77  | Mali                | 121,481    | 127,946    | 135,739    | 143,418    | 156,351    |
| 78  | Malta               | 48,488     | 53,989     | 61,926     | 69,087     | 76,097     |
| 79  | Mauritania          | 38,274     | 43,458     | 48,735     | 59,433     | 70,112     |
| 80  | Mauritius           | 50,993     | 56,343     | 61,929     | 70,701     | 80,604     |
| 81  | Mexico              | 5,659,273  | 6,337,459  | 7,064,330  | 8,067,346  | 9,107,473  |
| 82  | Mongolia            | 232,447    | 226,291    | 233,566    | 221,330    | 229,253    |
| 83  | Morocco             | 635,221    | 734,898    | 812,980    | 913,325    | 1,019,743  |
| 84  | Mozambique          | 267,929    | 268,814    | 271,756    | 276,555    | 283,993    |
| 85  | Myanmar             | 338,584    | 328,768    | 319,137    | 314,189    | 317,520    |
| 86  | Namibia             | 148,256    | 159,015    | 169,200    | 184,862    | 204,607    |
| 87  | Nepal               | 153,020    | 150,987    | 153,106    | 161,165    | 175,158    |
| 88  | Netherlands         | 5,103,242  | 5,463,776  | 5,895,026  | 6,346,816  | 6,839,819  |
| 89  | New Zealand         | 867,155    | 931,930    | 1,010,962  | 1,121,647  | 1,224,854  |
| 90  | Nicaragua           | 81,535     | 84,403     | 91,494     | 96,521     | 101,188    |
| 91  | Niger               | 52,084     | 56,293     | 62,432     | 70,249     | 83,881     |
| 92  | Nigeria             | 1,604,302  | 1,623,550  | 1,644,292  | 1,711,318  | 1,814,508  |
| 93  | Norway              | 2,602,873  | 2,666,385  | 2,770,344  | 2,950,411  | 3,178,995  |
| 94  | Pakistan            | 1,051,539  | 1,224,868  | 1,408,752  | 1,630,292  | 1,853,840  |
| 95  | Panama              | 132,533    | 151,945    | 168,641    | 189,994    | 217,620    |
| 96  | Papua New Guinea    | 319,177    | 314,209    | 309,720    | 306,782    | 304,566    |
| 97  | Paraguay            | 202,482    | 210,508    | 222,735    | 231,373    | 245,811    |
| 98  | Peru                | 1,732,583  | 1,789,353  | 1,888,314  | 1,936,395  | 2,074,525  |
| 99  | Philippines         | 796,723    | 904,962    | 1,047,595  | 1,165,500  | 1,330,369  |
| 100 | Poland              | 3,301,177  | 3,479,992  | 3,715,899  | 3,841,872  | 4,145,075  |
| 101 | Portugal            | 2,214,539  | 2,384,715  | 2,620,983  | 2,766,700  | 2,912,219  |
| 102 | Qatar               | 436,878    | 440,189    | 453,075    | 514,920    | 763,008    |
| 103 | Republic of Korea   | 5,268,564  | 6,413,398  | 7,439,254  | 8,444,159  | 9,397,391  |
| 104 | Republic of Moldova | 67,954     | 65,801     | 62,070     | 57,817     | 53,635     |
| 105 | Romania             | 1,182,697  | 1,219,401  | 1,264,036  | 1,248,543  | 1,343,364  |
| 106 | Russian Federation* | 19,691,845 | 19,550,393 | 19,535,785 | 19,486,481 | 19,464,667 |
| 107 | Rwanda              | 25,556     | 21,304     | 30,430     | 35,205     | 43,397     |
| 108 | Saudi Arabia        | 5,758,030  | 6,229,676  | 6,368,480  | 6,823,590  | 7,469,646  |
| 109 | Senegal             | 113,069    | 119,274    | 129,294    | 139,978    | 158,149    |
| 110 | Serbia              | 478,857    | 527,318    | 532,370    | 539,616    | 570,721    |
| 111 | Sierra Leone        | 35,269     | 34,227     | 34,945     | 40,278     | 43,677     |

Inclusive wealth change percentage change with respect to the base year, 1990

Inclusive wealth per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 70,703  | 74,423  | 75,870  | 74,889  | 76,949  | 5    | 7    | 6    | 9    |
| 23,724  | 26,152  | 30,783  | 36,144  | 44,451  | 10   | 30   | 52   | 87   |
| 14,007  | 13,023  | 12,017  | 10,884  | 10,173  | -7   | -14  | -22  | -27  |
| 131,930 | 139,659 | 155,821 | 168,794 | 182,700 | 6    | 18   | 28   | 38   |
| 19,180  | 18,960  | 18,441  | 19,504  | 20,265  | -1   | -4   | 2    | 6    |
| 48,128  | 49,550  | 51,779  | 56,258  | 62,042  | 3    | 8    | 17   | 29   |
| 67,127  | 68,682  | 70,672  | 75,761  | 80,296  | 2    | 5    | 13   | 20   |
| 106,017 | 98,126  | 96,860  | 86,887  | 83,183  | -7   | -9   | -18  | -22  |
| 25,633  | 27,291  | 28,235  | 30,051  | 31,915  | 6    | 10   | 17   | 25   |
| 19,778  | 16,871  | 14,931  | 13,315  | 12,141  | -15  | -25  | -33  | -39  |
| 8,622   | 7,803   | 7,099   | 6,783   | 6,620   | -10  | -18  | -21  | -23  |
| 104,787 | 96,294  | 89,248  | 88,878  | 89,611  | -8   | -15  | -15  | -14  |
| 8,019   | 6,992   | 6,275   | 5,907   | 5,847   | -13  | -22  | -26  | -27  |
| 342,690 | 354,281 | 371,625 | 389,245 | 411,715 | 3    | 8    | 14   | 20   |
| 255,197 | 253,582 | 262,041 | 271,315 | 280,407 | -1   | 3    | 6    | 10   |
| 19,786  | 18,202  | 18,033  | 17,794  | 17,482  | -8   | -9   | -10  | -12  |
| 6,688   | 6,133   | 5,716   | 5,406   | 5,408   | -8   | -15  | -19  | -19  |
| 16,446  | 14,758  | 13,294  | 12,239  | 11,454  | -10  | -19  | -26  | -30  |
| 613,670 | 611,683 | 616,885 | 638,162 | 651,018 | 0    | 1    | 4    | 6    |
| 9,402   | 9,618   | 9,748   | 10,276  | 10,679  | 2    | 4    | 9    | 14   |
| 54,858  | 56,761  | 57,048  | 58,671  | 61,880  | 3    | 4    | 7    | 13   |
| 76,769  | 66,632  | 57,581  | 50,330  | 44,409  | -13  | -25  | -34  | -42  |
| 47,712  | 43,898  | 41,683  | 39,230  | 38,083  | -8   | -13  | -18  | -20  |
| 79,896  | 75,097  | 73,015  | 70,264  | 71,347  | -6   | -9   | -12  | -11  |
| 12,928  | 13,067  | 13,551  | 13,624  | 14,265  | 1    | 5    | 5    | 10   |
| 86,745  | 90,644  | 97,015  | 100,665 | 108,293 | 4    | 12   | 16   | 25   |
| 223,117 | 235,529 | 253,573 | 262,404 | 272,793 | 6    | 14   | 18   | 22   |
| 922,224 | 877,971 | 766,681 | 627,197 | 433,825 | -5   | -17  | -32  | -53  |
| 122,581 | 143,598 | 161,766 | 179,494 | 195,033 | 17   | 32   | 46   | 59   |
| 15,572  | 15,166  | 15,114  | 15,349  | 15,012  | -3   | -3   | -1   | -4   |
| 50,964  | 53,763  | 56,960  | 57,347  | 62,522  | 5    | 12   | 13   | 23   |
| 132,450 | 131,477 | 133,116 | 135,470 | 136,156 | -1   | 1    | 2    | 3    |
| 3,595   | 3,825   | 3,758   | 3,826   | 4,085   | 6    | 5    | 6    | 14   |
| 356,776 | 336,888 | 317,705 | 283,830 | 272,137 | -6   | -11  | -20  | -24  |
| 15,614  | 14,252  | 13,601  | 12,875  | 12,719  | -9   | -13  | -18  | -19  |
| 50,044  | 51,678  | 52,535  | 54,749  | 57,905  | 3    | 5    | 9    | 16   |
| 8,858   | 8,780   | 8,434   | 7,816   | 7,444   | -1   | -5   | -12  | -16  |

#### Wealth

|     | Country/Year                       | 1990        | 1995        | 2000        | 2005        | 2010        |
|-----|------------------------------------|-------------|-------------|-------------|-------------|-------------|
| 112 | Singapore                          | 560,076     | 698,776     | 885,471     | 1,051,266   | 1,368,578   |
| 113 | Slovakia**                         | 528,665     | 546,157     | 578,849     | 623,258     | 666,294     |
| 114 | Slovenia*                          | 366,092     | 401,279     | 431,280     | 468,501     | 495,111     |
| 115 | South Africa                       | 2,616,273   | 3,013,284   | 3,127,813   | 3,433,802   | 3,628,541   |
| 116 | Spain                              | 9,309,175   | 10,741,110  | 12,256,939  | 14,018,657  | 16,074,035  |
| 117 | Sri Lanka                          | 291,846     | 322,911     | 369,254     | 394,298     | 442,899     |
| 118 | Sudan (former)                     | 520,644     | 521,368     | 534,550     | 571,482     | 624,194     |
| 119 | Swaziland                          | 41,244      | 48,706      | 54,142      | 57,458      | 60,387      |
| 120 | Sweden                             | 3,491,613   | 3,603,607   | 3,739,798   | 3,993,066   | 4,337,750   |
| 121 | Switzerland                        | 3,714,812   | 3,880,368   | 4,020,050   | 4,293,789   | 4,651,636   |
| 122 | Syrian Arab Republic               | 296,073     | 334,935     | 373,837     | 411,842     | 460,871     |
| 123 | Tajikistan*                        | 30,644      | 30,782      | 30,419      | 30,837      | 31,826      |
| 124 | Thailand                           | 1,437,028   | 1,678,861   | 1,850,642   | 2,016,731   | 2,225,034   |
| 125 | Тодо                               | 39,345      | 40,896      | 46,338      | 51,814      | 57,221      |
| 126 | Trinidad and Tobago                | 174,561     | 175,416     | 182,804     | 191,142     | 184,628     |
| 127 | Tunisia                            | 292,994     | 332,547     | 373,827     | 411,675     | 467,016     |
| 128 | Turkey                             | 3,423,105   | 3,945,817   | 4,371,834   | 4,798,599   | 5,500,100   |
| 129 | Uganda                             | 65,758      | 71,707      | 81,331      | 93,607      | 116,156     |
| 130 | Ukraine*                           | 2,016,108   | 2,053,045   | 1,958,753   | 1,923,031   | 1,912,677   |
| 131 | United Arab Emirates               | 1,758,990   | 1,967,716   | 2,238,375   | 2,677,592   | 4,004,059   |
| 132 | United Kingdom                     | 19,766,855  | 20,099,689  | 21,505,927  | 23,612,677  | 25,377,131  |
| 133 | United Republic of Tanzania        | 231,635     | 234,884     | 234,544     | 240,914     | 256,542     |
| 134 | United States of America           | 104,292,941 | 110,603,864 | 121,230,967 | 133,576,817 | 143,824,201 |
| 135 | Uruguay                            | 226,990     | 238,897     | 260,196     | 263,264     | 284,856     |
| 136 | Venezuela (Bolivarian Republic of) | 3,419,948   | 3,520,694   | 3,670,473   | 3,827,194   | 4,042,649   |
| 137 | Vietnam                            | 451,088     | 511,221     | 613,708     | 721,443     | 861,705     |
| 138 | Yemen                              | 201,889     | 236,196     | 280,911     | 325,030     | 375,183     |
| 139 | Zambia                             | 401,431     | 400,225     | 402,409     | 407,626     | 418,101     |
| 140 | Zimbabwe                           | 157,088     | 158,800     | 157,766     | 158,456     | 150,414     |

Inclusive wealth change percentage change with respect to the base year, 1990

Inclusive wealth per capita in millions of constant 2005 US\$

| - |         |         |         |         |         |      |      |      |      |
|---|---------|---------|---------|---------|---------|------|------|------|------|
|   | 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|   | 185,663 | 200,701 | 225,926 | 246,429 | 269,065 | 8    | 22   | 33   | 45   |
|   | 99,469  | 101,726 | 107,098 | 115,088 | 121,985 | 2    | 8    | 16   | 23   |
|   | 189,165 | 204,088 | 217,225 | 234,007 | 243,936 | 8    | 15   | 24   | 29   |
|   | 71,106  | 72,780  | 69,879  | 71,848  | 72,379  | 2    | -2   | 1    | 2    |
|   | 239,377 | 272,431 | 304,230 | 323,044 | 348,852 | 14   | 27   | 35   | 46   |
|   | 16,834  | 17,714  | 19,699  | 19,871  | 21,232  | 5    | 17   | 18   | 26   |
|   | 19,651  | 17,298  | 15,636  | 14,878  | 14,332  | -12  | -20  | -24  | -27  |
|   | 47,794  | 50,547  | 50,893  | 52,003  | 50,914  | 6    | 6    | 9    | 7    |
|   | 407,954 | 408,251 | 422,092 | 442,232 | 462,462 | 0    | 3    | 8    | 13   |
|   | 556,631 | 552,861 | 560,840 | 579,067 | 606,921 | -1   | 1    | 4    | 9    |
|   | 24,024  | 23,635  | 23,382  | 22,281  | 22,580  | -2   | -3   | -7   | -6   |
|   | 5,655   | 5,330   | 4,928   | 4,779   | 4,627   | -6   | -13  | -16  | -18  |
|   | 25,179  | 28,145  | 29,303  | 30,237  | 32,190  | 12   | 16   | 20   | 28   |
|   | 10,734  | 10,010  | 9,667   | 9,581   | 9,493   | -7   | -10  | -11  | -12  |
|   | 143,613 | 139,059 | 141,483 | 145,312 | 137,631 | -3   | -1   | 1    | -4   |
|   | 35,665  | 37,216  | 39,533  | 41,532  | 44,559  | 4    | 11   | 16   | 25   |
|   | 63,238  | 67,032  | 68,709  | 70,419  | 75,600  | 6    | 9    | 11   | 20   |
|   | 3,715   | 3,442   | 3,359   | 3,292   | 3,475   | -7   | -10  | -11  | -6   |
|   | 39,022  | 40,160  | 40,063  | 40,982  | 42,085  | 3    | 3    | 5    | 8    |
|   | 972,547 | 837,847 | 737,888 | 657,990 | 533,044 | -14  | -24  | -32  | -45  |
|   | 345,487 | 346,563 | 365,287 | 392,219 | 409,074 | 0    | 6    | 14   | 18   |
|   | 9,091   | 7,844   | 6,891   | 6,204   | 5,721   | -14  | -24  | -32  | -37  |
|   | 411,673 | 415,299 | 429,142 | 450,026 | 463,375 | 1    | 4    | 9    | 13   |
|   | 73,008  | 74,113  | 78,394  | 79,236  | 84,557  | 2    | 7    | 9    | 16   |
|   | 173,732 | 159,778 | 150,749 | 143,533 | 139,499 | -8   | -13  | -17  | -20  |
|   | 6,722   | 6,908   | 7,792   | 8,675   | 9,809   | 3    | 16   | 29   | 46   |
|   | 16,897  | 15,592  | 15,850  | 15,741  | 15,599  | -8   | -6   | -7   | -8   |
|   | 51,072  | 44,871  | 39,446  | 35,562  | 31,944  | -12  | -23  | -30  | -37  |
|   | 15,005  | 13,590  | 12,612  | 12,605  | 11,965  | -9   | -16  | -16  | -20  |

#### Human capital

Human capital level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995       | 2000       | 2005       | 2010       |
|----|----------------------------------|-----------|------------|------------|------------|------------|
| 1  | Afghanistan                      | 39,480    | 61,321     | 71,117     | 92,528     | 111,250    |
| 2  | Albania                          | 42,720    | 41,466     | 41,309     | 45,641     | 48,460     |
| 3  | Algeria                          | 330,494   | 404,177    | 471,665    | 535,711    | 607,230    |
| 4  | Argentina                        | 1,209,100 | 1,305,938  | 1,442,112  | 1,640,718  | 1,754,180  |
| 5  | Armenia                          | 50,623    | 45,856     | 45,901     | 47,248     | 47,814     |
| 6  | Australia                        | 3,938,197 | 4,204,911  | 4,477,842  | 4,927,504  | 5,663,742  |
| 7  | Austria                          | 1,912,493 | 2,120,114  | 2,155,198  | 2,328,310  | 2,442,916  |
| 8  | Bahrain                          | 36,959    | 47,078     | 57,385     | 67,496     | 131,505    |
| 9  | Bangladesh                       | 351,223   | 404,298    | 484,267    | 552,709    | 609,936    |
| 10 | Barbados                         | 43,915    | 45,217     | 46,069     | 50,187     | 51,267     |
| 11 | Belgium                          | 2,605,881 | 2,763,022  | 2,884,003  | 3,096,223  | 3,173,437  |
| 12 | Belize                           | 4,292     | 5,062      | 6,084      | 7,382      | 8,417      |
| 13 | Benin                            | 31,003    | 36,814     | 44,364     | 52,716     | 64,989     |
| 14 | Bolivia (Plurinational State of) | 62,960    | 71,971     | 85,782     | 102,001    | 115,672    |
| 15 | Botswana                         | 27,720    | 35,083     | 40,050     | 47,166     | 52,377     |
| 16 | Brazil                           | 5,385,277 | 6,455,777  | 7,464,481  | 8,367,825  | 8,968,120  |
| 17 | Bulgaria                         | 210,179   | 202,913    | 199,625    | 196,728    | 198,201    |
| 18 | Burundi                          | 13,252    | 14,856     | 15,532     | 18,849     | 22,951     |
| 19 | Cambodia                         | 19,842    | 25,079     | 30,378     | 35,157     | 38,950     |
| 20 | Cameroon                         | 100,230   | 117,636    | 139,841    | 156,365    | 179,829    |
| 21 | Canada                           | 6,541,784 | 6,825,762  | 7,384,326  | 8,538,772  | 9,105,623  |
| 22 | Central African Republic         | 4,280     | 4,988      | 5,639      | 6,322      | 6,888      |
| 23 | Chile                            | 661,306   | 730,906    | 816,619    | 920,373    | 1,068,195  |
| 24 | China                            | 9,210,965 | 10,464,623 | 11,597,065 | 12,427,482 | 13,446,810 |
| 25 | Colombia                         | 976,550   | 1,150,156  | 1,336,597  | 1,509,853  | 1,717,331  |
| 26 | Congo                            | 2,514     | 2,868      | 3,342      | 3,831      | 4,511      |
| 27 | Costa Rica                       | 134,876   | 156,077    | 191,409    | 221,220    | 251,580    |
| 28 | Côte d'Ivoire                    | 120,374   | 150,227    | 171,259    | 192,434    | 210,304    |
| 29 | Croatia*                         | 517,580   | 540,959    | 528,355    | 537,436    | 542,251    |
| 30 | Cuba                             | 244,605   | 249,111    | 256,195    | 269,802    | 278,918    |
| 31 | Cyprus                           | 125,474   | 143,466    | 168,537    | 186,875    | 216,058    |
| 32 | Czech Republic**                 | 861,255   | 917,084    | 929,790    | 1,006,725  | 1,016,335  |
| 33 | Democratic Republic of the Congo | 109,259   | 134,420    | 154,080    | 182,079    | 213,679    |
| 34 | Denmark                          | 1,830,664 | 1,790,251  | 1,788,566  | 1,845,875  | 1,934,085  |
| 35 | Dominican Republic               | 200,150   | 230,567    | 253,255    | 287,492    | 321,863    |
| 36 | Ecuador                          | 121,991   | 142,567    | 159,212    | 179,493    | 203,225    |
| 37 | Egypt                            | 502,731   | 564,326    | 647,236    | 731,445    | 849,042    |

Human capital change percentage change with respect to the base year, 1990

Human capital per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010    | 2005    | 2000    | 1995    | 1990    |
|------|------|------|------|---------|---------|---------|---------|---------|
| 17   | 11   | 3    | 2    | 3,542   | 3,351   | 3,111   | 3,099   | 3,029   |
| 16   | 12   | 4    | 2    | 15,124  | 14,527  | 13,447  | 13,201  | 12,987  |
| 31   | 25   | 18   | 9    | 17,120  | 16,289  | 15,447  | 14,286  | 13,063  |
| 17   | 15   | 5    | 1    | 43,407  | 42,416  | 39,049  | 37,468  | 37,041  |
| 8    | 8    | 4    | 0    | 15,463  | 15,410  | 14,922  | 14,227  | 14,281  |
| 10   | 5    | 1    | 1    | 254,340 | 241,503 | 233,655 | 232,079 | 230,356 |
| 17   | 13   | 8    | 7    | 291,044 | 282,830 | 269,241 | 267,164 | 249,331 |
| 39   | 24   | 20   | 12   | 104,218 | 93,123  | 89,918  | 84,208  | 74,984  |
| 23   | 18   | 12   | 3    | 4,102   | 3,931   | 3,737   | 3,441   | 3,337   |
| 11   | 10   | 2    | 1    | 187,564 | 185,531 | 172,215 | 171,656 | 169,228 |
| 13   | 14   | 8    | 5    | 296,249 | 297,308 | 283,421 | 274,113 | 261,923 |
| 20   | 16   | 8    | 2    | 27,009  | 26,276  | 24,285  | 22,975  | 22,563  |
| 13   | 6    | 5    | 0    | 7,343   | 6,906   | 6,807   | 6,514   | 6,495   |
| 23   | 18   | 9    | 2    | 11,649  | 11,152  | 10,326  | 9,633   | 9,456   |
| 30   | 25   | 14   | 10   | 26,098  | 25,146  | 22,783  | 22,125  | 20,052  |
| 28   | 25   | 19   | 11   | 46,003  | 44,991  | 42,795  | 39,888  | 35,986  |
| 11   | 7    | 5    | 2    | 26,447  | 25,420  | 24,934  | 24,282  | 23,833  |
| 16   | 10   | 3    | 3    | 2,738   | 2,599   | 2,437   | 2,441   | 2,366   |
| 32   | 26   | 17   | 8    | 2,755   | 2,632   | 2,441   | 2,245   | 2,082   |
| 12   | 8    | 8    | 3    | 9,175   | 8,908   | 8,919   | 8,439   | 8,228   |
| 13   | 12   | 2    | -1   | 267,682 | 264,494 | 240,788 | 232,945 | 236,158 |
| 7    | 8    | 4    | 3    | 1,565   | 1,573   | 1,524   | 1,499   | 1,458   |
| 24   | 13   | 6    | 1    | 62,418  | 56,459  | 52,959  | 50,724  | 50,145  |
| 25   | 18   | 14   | 7    | 10,025  | 9,504   | 9,138   | 8,620   | 8,043   |
| 26   | 19   | 14   | 7    | 37,096  | 35,080  | 33,613  | 31,551  | 29,411  |
| 6    | 3    | 1    | 0    | 1,116   | 1,084   | 1,066   | 1,049   | 1,052   |
| 23   | 17   | 11   | 2    | 54,000  | 51,334  | 48,839  | 44,993  | 43,930  |
| 11   | 11   | 7    | 6    | 10,655  | 10,678  | 10,328  | 10,235  | 9,616   |
| 8    | 6    | 3    | 2    | 123,146 | 120,990 | 117,268 | 115,862 | 113,802 |
| 7    | 4    | 0    | -1   | 24,775  | 23,973  | 23,072  | 22,853  | 23,141  |
| 20   | 11   | 9    | 2    | 195,767 | 180,982 | 178,668 | 167,714 | 163,663 |
| 16   | 18   | 9    | 6    | 96,859  | 98,499  | 90,774  | 88,870  | 83,506  |
| 8    | 6    | 3    | 2    | 3,239   | 3,171   | 3,105   | 3,050   | 3,001   |
| -2   | -4   | -6   | -4   | 348,475 | 340,602 | 334,969 | 342,084 | 356,089 |
| 17   | 12   | 6    | 5    | 32,422  | 31,032  | 29,476  | 29,126  | 27,819  |
| 18   | 12   | 8    | 5    | 14,050  | 13,369  | 12,897  | 12,523  | 11,889  |
| 18   | 11   | 8    | 3    | 10,466  | 9,857   | 9,568   | 9,093   | 8,844   |

#### Human capital

Human capital level in millions of constant 2005 US\$

|    | Country/Year                     | 1990       | 1995       | 2000       | 2005       | 2010       |
|----|----------------------------------|------------|------------|------------|------------|------------|
| 38 | El Salvador                      | 92,966     | 106,169    | 117,112    | 132,107    | 145,210    |
| 39 | Estonia                          | 107,181    | 102,068    | 105,380    | 108,193    | 111,841    |
| 40 | Fiji                             | 20,815     | 24,607     | 25,521     | 27,041     | 29,286     |
| 41 | Finland                          | 1,304,527  | 1,367,531  | 1,321,571  | 1,475,176  | 1,555,587  |
| 42 | France                           | 14,563,212 | 15,530,642 | 16,745,322 | 17,938,170 | 19,118,064 |
| 43 | Gabon                            | 47,646     | 58,089     | 69,020     | 80,896     | 97,620     |
| 44 | Gambia                           | 3,308      | 4,106      | 4,661      | 5,658      | 6,589      |
| 45 | Germany                          | 17,572,938 | 19,835,706 | 21,046,418 | 24,337,379 | 24,576,140 |
| 46 | Ghana                            | 62,036     | 73,103     | 84,082     | 97,187     | 113,336    |
| 47 | Greece                           | 1,055,664  | 1,161,390  | 1,231,362  | 1,342,330  | 1,413,804  |
| 48 | Guatemala                        | 199,192    | 230,123    | 265,597    | 303,490    | 362,912    |
| 49 | Guyana                           | 8,593      | 8,668      | 8,856      | 8,983      | 9,944      |
| 50 | Haiti                            | 26,449     | 31,563     | 36,863     | 41,020     | 46,341     |
| 51 | Honduras                         | 69,581     | 80,868     | 99,708     | 113,657    | 138,451    |
| 52 | Hungary                          | 804,809    | 880,945    | 918,322    | 971,444    | 968,933    |
| 53 | Iceland                          | 87,149     | 94,952     | 101,448    | 112,108    | 125,767    |
| 54 | India                            | 5,954,841  | 6,636,022  | 7,579,171  | 8,479,881  | 9,354,665  |
| 55 | Indonesia                        | 1,779,692  | 1,987,999  | 2,260,719  | 2,452,002  | 2,703,511  |
| 56 | Iran (Islamic Republic of)       | 817,985    | 1,013,433  | 1,277,430  | 1,576,454  | 1,633,140  |
| 57 | Iraq                             | 99,340     | 117,674    | 143,166    | 165,540    | 190,166    |
| 58 | Ireland                          | 807,136    | 875,099    | 998,721    | 1,171,091  | 1,242,911  |
| 59 | Israel                           | 735,676    | 923,177    | 1,076,119  | 1,200,168  | 1,385,779  |
| 60 | Italy                            | 11,029,940 | 11,201,349 | 11,370,113 | 12,167,855 | 12,674,011 |
| 61 | Jamaica                          | 80,436     | 89,090     | 96,549     | 101,750    | 108,062    |
| 62 | Japan                            | 30,190,110 | 32,024,691 | 33,097,452 | 33,566,644 | 33,644,994 |
| 63 | Jordan                           | 67,127     | 100,266    | 112,892    | 126,396    | 152,675    |
| 64 | Kazakhstan*                      | 332,612    | 341,786    | 352,790    | 379,926    | 410,350    |
| 65 | Kenya                            | 143,493    | 180,514    | 206,607    | 247,204    | 288,468    |
| 66 | Kuwait                           | 110,822    | 95,524     | 115,047    | 138,186    | 164,920    |
| 67 | Kyrgyzstan*                      | 11,869     | 12,693     | 14,243     | 14,875     | 16,173     |
| 68 | Lao People's Democratic Republic | 9,928      | 11,795     | 13,925     | 15,668     | 18,185     |
| 69 | Latvia                           | 122,740    | 115,832    | 115,443    | 126,415    | 125,943    |
| 70 | Lesotho                          | 16,213     | 18,137     | 20,347     | 21,632     | 23,289     |
| 71 | Liberia                          | 2,749      | 2,884      | 4,242      | 4,864      | 6,461      |
| 72 | Lithuania*                       | 188,663    | 187,406    | 186,943    | 197,469    | 200,365    |
| 73 | Luxembourg                       | 130,302    | 139,026    | 154,553    | 170,127    | 202,291    |
| 74 | Malawi                           | 14,116     | 14,702     | 16,885     | 20,150     | 23,486     |

Human capital change percentage change with respect to the base year, 1990

#### Human capital per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 17,433  | 18,518  | 19,715  | 21,834  | 23,447  | 6    | 13   | 25   | 35   |
| 68,371  | 70,846  | 76,877  | 80,390  | 83,393  | 4    | 12   | 18   | 22   |
| 28,577  | 31,724  | 31,441  | 32,875  | 34,029  | 11   | 10   | 15   | 19   |
| 261,615 | 267,734 | 255,457 | 281,289 | 289,975 | 2    | -2   | 8    | 11   |
| 256,809 | 268,490 | 283,589 | 294,085 | 304,489 | 5    | 10   | 15   | 19   |
| 51,284  | 53,424  | 55,874  | 59,016  | 64,844  | 4    | 9    | 15   | 26   |
| 3,424   | 3,646   | 3,594   | 3,763   | 3,812   | 6    | 5    | 10   | 11   |
| 222,166 | 242,107 | 255,576 | 294,853 | 298,608 | 9    | 15   | 33   | 34   |
| 4,194   | 4,301   | 4,387   | 4,491   | 4,646   | 3    | 5    | 7    | 11   |
| 103,899 | 108,828 | 112,076 | 120,032 | 124,462 | 5    | 8    | 16   | 20   |
| 22,323  | 22,976  | 23,636  | 23,865  | 25,222  | 3    | 6    | 7    | 13   |
| 11,854  | 11,908  | 12,080  | 12,038  | 13,180  | 0    | 2    | 2    | 11   |
| 3,712   | 4,007   | 4,264   | 4,388   | 4,637   | 8    | 15   | 18   | 25   |
| 14,231  | 14,505  | 16,035  | 16,522  | 18,216  | 2    | 13   | 16   | 28   |
| 77,562  | 85,269  | 89,939  | 96,307  | 97,052  | 10   | 16   | 24   | 25   |
| 342,040 | 354,992 | 360,754 | 377,793 | 392,854 | 4    | 5    | 10   | 15   |
| 6,815   | 6,880   | 7,192   | 7,438   | 7,639   | 1    | 6    | 9    | 12   |
| 9,654   | 9,970   | 10,594  | 10,787  | 11,271  | 3    | 10   | 12   | 17   |
| 14,908  | 16,959  | 19,550  | 22,607  | 22,077  | 14   | 31   | 52   | 48   |
| 5,718   | 5,800   | 6,001   | 6,051   | 6,004   | 1    | 5    | 6    | 5    |
| 228,571 | 242,345 | 262,560 | 281,645 | 278,062 | 6    | 15   | 23   | 22   |
| 163,485 | 173,136 | 178,907 | 181,718 | 186,803 | 6    | 9    | 11   | 14   |
| 194,079 | 196,625 | 199,524 | 207,391 | 209,312 | 1    | 3    | 7    | 8    |
| 34,012  | 36,193  | 37,398  | 37,943  | 39,424  | 6    | 10   | 12   | 16   |
| 246,951 | 257,254 | 263,263 | 265,574 | 265,893 | 4    | 7    | 8    | 8    |
| 19,653  | 22,881  | 23,387  | 23,661  | 24,676  | 16   | 19   | 20   | 26   |
| 20,142  | 21,461  | 23,587  | 25,042  | 25,605  | 7    | 17   | 24   | 27   |
| 6,120   | 6,582   | 6,611   | 6,941   | 7,120   | 8    | 8    | 13   | 16   |
| 53,083  | 58,679  | 59,278  | 61,036  | 60,262  | 11   | 12   | 15   | 14   |
| 2,672   | 2,764   | 2,875   | 2,950   | 3,032   | 3    | 8    | 10   | 13   |
| 2,368   | 2,460   | 2,619   | 2,723   | 2,933   | 4    | 11   | 15   | 24   |
| 46,075  | 46,480  | 48,404  | 54,831  | 55,923  | 1    | 5    | 19   | 21   |
| 9,891   | 10,107  | 10,360  | 10,472  | 10,726  | 2    | 5    | 6    | 8    |
| 1,292   | 1,377   | 1,490   | 1,528   | 1,618   | 7    | 15   | 18   | 25   |
| 51,016  | 51,640  | 53,412  | 57,811  | 60,285  | 1    | 5    | 13   | 18   |
| 341,799 | 341,150 | 354,895 | 372,110 | 398,644 | 0    | 4    | 9    | 17   |
| 1,505   | 1,488   | 1,504   | 1,571   | 1,576   | -1   | 0    | 4    | 5    |

#### Human capital

Human capital level in millions of constant 2005 US\$

|     | Country/Year        | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|---------------------|-----------|-----------|-----------|-----------|-----------|
| 75  | Malaysia            | 722,253   | 908,587   | 1,078,764 | 1,223,142 | 1,403,269 |
| 76  | Maldives            | 4,430     | 5,323     | 6,610     | 7,837     | 9,499     |
| 77  | Mali                | 36,437    | 41,792    | 48,511    | 56,049    | 69,024    |
| 78  | Malta               | 41,628    | 44,604    | 49,155    | 53,543    | 58,450    |
| 79  | Mauritania          | 27,487    | 32,992    | 38,658    | 47,698    | 55,448    |
| 80  | Mauritius           | 42,119    | 44,710    | 46,981    | 52,099    | 57,015    |
| 81  | Mexico              | 3,303,020 | 3,771,351 | 4,239,892 | 4,933,308 | 5,525,249 |
| 82  | Mongolia            | 12,800    | 13,977    | 15,548    | 17,591    | 19,563    |
| 83  | Morocco             | 467,171   | 549,131   | 602,563   | 666,207   | 706,976   |
| 84  | Mozambique          | 27,655    | 32,156    | 37,213    | 42,850    | 48,938    |
| 85  | Myanmar             | 47,259    | 53,335    | 58,554    | 63,957    | 68,550    |
| 86  | Namibia             | 69,509    | 80,338    | 90,451    | 105,109   | 120,738   |
| 87  | Nepal               | 37,056    | 41,214    | 48,612    | 57,640    | 66,392    |
| 88  | Netherlands         | 3,711,231 | 3,921,464 | 4,100,906 | 4,325,578 | 4,581,766 |
| 89  | New Zealand         | 484,972   | 539,080   | 579,080   | 658,248   | 705,712   |
| 90  | Nicaragua           | 26,559    | 30,692    | 36,869    | 41,713    | 46,591    |
| 91  | Niger               | 28,153    | 34,456    | 42,392    | 49,739    | 61,341    |
| 92  | Nigeria             | 558,170   | 632,976   | 715,751   | 827,020   | 963,107   |
| 93  | Norway              | 1,538,519 | 1,613,671 | 1,693,401 | 1,852,887 | 1,982,162 |
| 94  | Pakistan            | 683,751   | 820,901   | 974,190   | 1,169,587 | 1,346,483 |
| 95  | Panama              | 74,718    | 91,658    | 102,642   | 119,752   | 132,099   |
| 96  | Papua New Guinea    | 13,522    | 15,732    | 18,117    | 20,706    | 23,973    |
| 97  | Paraguay            | 50,657    | 59,572    | 67,422    | 80,063    | 94,100    |
| 98  | Peru                | 386,127   | 446,428   | 530,178   | 570,603   | 654,704   |
| 99  | Philippines         | 509,938   | 585,668   | 682,538   | 756,472   | 863,279   |
| 100 | Poland              | 2,661,481 | 2,814,946 | 2,920,790 | 2,942,842 | 3,044,674 |
| 101 | Portugal            | 1,791,067 | 1,888,218 | 2,010,059 | 2,051,338 | 2,130,564 |
| 102 | Qatar               | 60,764    | 68,418    | 82,281    | 126,648   | 302,140   |
| 103 | Republic of Korea   | 4,093,798 | 4,591,453 | 4,938,518 | 5,245,909 | 5,499,675 |
| 104 | Republic of Moldova | 28,414    | 29,005    | 28,869    | 26,960    | 22,969    |
| 105 | Romania             | 723,290   | 767,448   | 804,769   | 763,252   | 767,888   |
| 106 | Russian Federation* | 5,669,166 | 5,691,289 | 6,177,924 | 6,458,518 | 6,390,872 |
| 107 | Rwanda              | 20,108    | 16,094    | 24,782    | 28,561    | 34,148    |
| 108 | Saudi Arabia        | 1,433,460 | 1,763,841 | 2,005,961 | 2,536,731 | 3,118,153 |
| 109 | Senegal             | 37,707    | 43,714    | 51,898    | 59,645    | 71,729    |
| 110 | Serbia              | 395,030   | 443,777   | 449,425   | 448,405   | 458,295   |
| 111 | Sierra Leone        | 17,618    | 17,207    | 18,674    | 24,455    | 28,128    |

Human capital change percentage change with respect to the base year, 1990

Human capital per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010    | 2005    | 2000    | 1995    | 1990    |
|------|------|------|------|---------|---------|---------|---------|---------|
| 25   | 18   | 16   | 11   | 49,409  | 46,863  | 46,072  | 43,849  | 39,666  |
| 49   | 31   | 20   | 6    | 30,071  | 26,544  | 24,193  | 21,403  | 20,187  |
| 7    | 1    | 2    | 1    | 4,491   | 4,254   | 4,295   | 4,254   | 4,201   |
| 24   | 15   | 9    | 2    | 140,331 | 130,818 | 123,685 | 115,382 | 113,266 |
| 16   | 14   | 6    | 5    | 16,027  | 15,653  | 14,628  | 14,394  | 13,774  |
| 10   | 4    | -1   | -1   | 43,886  | 41,456  | 39,281  | 39,319  | 39,752  |
| 24   | 18   | 8    | 4    | 48,714  | 46,329  | 42,416  | 40,872  | 39,179  |
| 22   | 18   | 10   | 4    | 7,098   | 6,906   | 6,448   | 6,061   | 5,838   |
| 17   | 16   | 11   | 8    | 22,127  | 21,920  | 20,927  | 20,392  | 18,852  |
| 2    | 1    | 0    | -1   | 2,092   | 2,063   | 2,045   | 2,018   | 2,041   |
| 19   | 15   | 8    | 5    | 1,429   | 1,381   | 1,302   | 1,266   | 1,203   |
| 8    | 3    | -3   | -1   | 52,879  | 50,534  | 47,710  | 48,650  | 49,129  |
| 14   | 9    | 3    | -2   | 2,216   | 2,113   | 1,992   | 1,908   | 1,942   |
| 11   | 6    | 4    | 2    | 275,794 | 265,284 | 258,523 | 254,275 | 249,214 |
| 13   | 12   | 5    | 3    | 161,559 | 159,223 | 150,097 | 146,686 | 142,723 |
| 25   | 19   | 13   | 3    | 8,049   | 7,690   | 7,267   | 6,619   | 6,445   |
| 9    | 6    | 7    | 4    | 3,954   | 3,828   | 3,881   | 3,754   | 3,615   |
| 6    | 3    | 1    | 1    | 6,079   | 5,915   | 5,787   | 5,754   | 5,722   |
| 12   | 10   | 4    | 2    | 405,922 | 400,772 | 377,077 | 370,185 | 362,731 |
| 27   | 21   | 10   | 5    | 7,757   | 7,372   | 6,741   | 6,446   | 6,113   |
| 21   | 20   | 12   | 11   | 37,562  | 36,980  | 34,722  | 34,240  | 30,927  |
| 7    | 4    | 4    | 3    | 3,496   | 3,397   | 3,368   | 3,336   | 3,252   |
| 22   | 14   | 6    | 4    | 14,579  | 13,575  | 12,618  | 12,423  | 11,936  |
| 26   | 16   | 15   | 5    | 22,517  | 20,705  | 20,500  | 18,736  | 17,806  |
| 12   | 7    | 7    | 2    | 9,257   | 8,843   | 8,829   | 8,457   | 8,274   |
| 14   | 10   | 9    | 5    | 79,544  | 77,108  | 76,256  | 73,322  | 69,936  |
| 11   | 8    | 8    | 3    | 199,574 | 194,556 | 194,468 | 186,492 | 180,451 |
| 34   | 20   | 9    | 6    | 171,788 | 154,264 | 139,233 | 136,462 | 128,269 |
| 20   | 17   | 13   | 8    | 114,140 | 111,510 | 107,388 | 102,804 | 95,248  |
| -1   | 10   | 8    | 3    | 6,429   | 7,157   | 7,030   | 6,685   | 6,511   |
| 15   | 12   | 16   | 9    | 35,738  | 35,057  | 36,264  | 33,837  | 31,167  |
| 17   | 18   | 10   | 0    | 44,704  | 44,900  | 42,096  | 38,274  | 38,132  |
| 14   | 10   | 8    | 2    | 3,214   | 3,104   | 3,060   | 2,889   | 2,828   |
| 28   | 19   | 13   | 7    | 113,602 | 105,516 | 100,072 | 95,385  | 88,819  |
| 11   | 5    | 5    | 0    | 5,769   | 5,486   | 5,460   | 5,223   | 5,207   |
| 13   | 10   | 7    | 5    | 46,498  | 45,495  | 44,350  | 43,491  | 41,283  |
| 8    | 7    | 2    | 0    | 4,794   | 4,745   | 4,507   | 4,414   | 4,425   |

#### Human capital

Human capital level in millions of constant 2005 US\$

|     | Country/Year                       | 1990       | 1995       | 2000       | 2005       | 2010       |
|-----|------------------------------------|------------|------------|------------|------------|------------|
| 112 | Singapore                          | 429,721    | 510,905    | 604,867    | 701,993    | 907,291    |
| 113 | Slovakia**                         | 360,909    | 373,975    | 386,533    | 412,498    | 426,543    |
| 114 | Slovenia*                          | 269,040    | 300,332    | 314,073    | 329,523    | 329,473    |
| 115 | South Africa                       | 1,470,352  | 1,868,184  | 1,962,459  | 2,227,970  | 2,297,274  |
| 116 | Spain                              | 6,745,216  | 7,787,640  | 8,778,299  | 9,765,963  | 11,050,147 |
| 117 | Sri Lanka                          | 235,388    | 260,333    | 297,495    | 310,668    | 336,914    |
| 118 | Sudan (former)                     | 111,466    | 125,975    | 148,000    | 171,486    | 201,691    |
| 119 | Swaziland                          | 28,930     | 33,551     | 37,660     | 41,849     | 45,462     |
| 120 | Sweden                             | 2,522,960  | 2,581,896  | 2,634,171  | 2,766,056  | 2,953,824  |
| 121 | Switzerland                        | 2,711,651  | 2,788,798  | 2,817,812  | 2,970,641  | 3,181,542  |
| 122 | Syrian Arab Republic               | 188,240    | 227,706    | 267,644    | 300,041    | 337,908    |
| 123 | Tajikistan*                        | 11,166     | 11,669     | 12,745     | 14,310     | 15,756     |
| 124 | Thailand                           | 847,717    | 886,318    | 959,879    | 1,063,295  | 1,170,214  |
| 125 | Тодо                               | 23,258     | 26,356     | 32,837     | 39,350     | 44,899     |
| 126 | Trinidad and Tobago                | 70,034     | 77,649     | 88,812     | 98,724     | 100,828    |
| 127 | Tunisia                            | 204,989    | 236,098    | 267,044    | 291,620    | 328,994    |
| 128 | Turkey                             | 2,585,417  | 2,950,242  | 3,173,236  | 3,423,735  | 3,835,905  |
| 129 | Uganda                             | 36,804     | 42,756     | 50,754     | 59,884     | 74,525     |
| 130 | Ukraine*                           | 745,666    | 782,291    | 757,268    | 756,514    | 749,004    |
| 131 | United Arab Emirates               | 422,414    | 634,804    | 901,039    | 1,319,613  | 2,541,624  |
| 132 | United Kingdom                     | 15,996,765 | 15,957,772 | 16,768,300 | 18,116,941 | 19,079,056 |
| 133 | United Republic of Tanzania        | 29,379     | 35,446     | 39,754     | 46,840     | 53,811     |
| 134 | United States of America           | 77,709,588 | 81,467,032 | 87,203,943 | 93,882,940 | 99,641,223 |
| 135 | Uruguay                            | 156,431    | 163,165    | 176,921    | 177,808    | 190,024    |
| 136 | Venezuela (Bolivarian Republic of) | 947,252    | 1,089,164  | 1,298,414  | 1,516,600  | 1,696,277  |
| 137 | Vietnam                            | 308,545    | 349,104    | 410,208    | 462,888    | 517,070    |
| 138 | Yemen                              | 114,725    | 148,476    | 189,488    | 229,656    | 276,793    |
| 139 | Zambia                             | 27,859     | 32,989     | 37,446     | 43,326     | 50,141     |
| 140 | Zimbabwe                           | 52,511     | 60,821     | 66,540     | 73,792     | 73,197     |

Human capital change percentage change with respect to the base year, 1990

Human capital per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 142,451 | 146,741 | 154,330 | 164,555 | 178,375 | 3    | 8    | 16   | 25   |
| 67,906  | 69,656  | 71,516  | 76,170  | 78,091  | 3    | 5    | 12   | 15   |
| 139,017 | 152,747 | 158,191 | 164,590 | 162,328 | 10   | 14   | 18   | 17   |
| 39,962  | 45,123  | 43,844  | 46,617  | 45,824  | 13   | 10   | 17   | 15   |
| 173,447 | 197,521 | 217,886 | 225,046 | 239,819 | 14   | 26   | 30   | 38   |
| 13,577  | 14,281  | 15,871  | 15,657  | 16,151  | 5    | 17   | 15   | 19   |
| 4,207   | 4,180   | 4,329   | 4,465   | 4,631   | -1   | 3    | 6    | 10   |
| 33,525  | 34,819  | 35,401  | 37,876  | 38,331  | 4    | 6    | 13   | 14   |
| 294,779 | 292,501 | 297,305 | 306,341 | 314,917 | -1   | 1    | 4    | 7    |
| 406,317 | 397,338 | 393,115 | 400,625 | 415,111 | -2   | -3   | -1   | 2    |
| 15,274  | 16,068  | 16,740  | 16,232  | 16,556  | 5    | 10   | 6    | 8    |
| 2,061   | 2,020   | 2,065   | 2,218   | 2,291   | -2   | 0    | 8    | 11   |
| 14,853  | 14,859  | 15,199  | 15,942  | 16,930  | 0    | 2    | 7    | 14   |
| 6,345   | 6,451   | 6,850   | 7,276   | 7,449   | 2    | 8    | 15   | 17   |
| 57,618  | 61,556  | 68,737  | 75,053  | 75,163  | 7    | 19   | 30   | 30   |
| 24,953  | 26,422  | 28,240  | 29,421  | 31,390  | 6    | 13   | 18   | 26   |
| 47,763  | 50,119  | 49,872  | 50,243  | 52,726  | 5    | 4    | 5    | 10   |
| 2,079   | 2,053   | 2,096   | 2,106   | 2,230   | -1   | 1    | 1    | 7    |
| 14,432  | 15,303  | 15,489  | 16,122  | 16,480  | 6    | 7    | 12   | 14   |
| 233,553 | 270,297 | 297,031 | 324,281 | 338,356 | 16   | 27   | 39   | 45   |
| 279,593 | 275,147 | 284,816 | 300,932 | 307,550 | -2   | 2    | 8    | 10   |
| 1,153   | 1,184   | 1,168   | 1,206   | 1,200   | 3    | 1    | 5    | 4    |
| 306,741 | 305,895 | 308,691 | 316,296 | 321,026 | 0    | 1    | 3    | 5    |
| 50,314  | 50,619  | 53,304  | 53,516  | 56,407  | 1    | 6    | 6    | 12   |
| 48,120  | 49,429  | 53,327  | 56,878  | 58,533  | 3    | 11   | 18   | 22   |
| 4,598   | 4,717   | 5,208   | 5,566   | 5,886   | 3    | 13   | 21   | 28   |
| 9,602   | 9,802   | 10,692  | 11,122  | 11,508  | 2    | 11   | 16   | 20   |
| 3,544   | 3,699   | 3,671   | 3,780   | 3,831   | 4    | 4    | 7    | 8    |
| 5,016   | 5,205   | 5,319   | 5,870   | 5,822   | 4    | 6    | 17   | 16   |

## Produced capital

Produced capital level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010       |
|----|----------------------------------|-----------|-----------|-----------|-----------|------------|
| 1  | Afghanistan                      | 25,451    | 25,065    | 24,444    | 26,994    | 31,927     |
| 2  | Albania                          | 41,198    | 37,168    | 36,242    | 41,945    | 50,506     |
| 3  | Algeria                          | 295,715   | 309,446   | 327,508   | 360,431   | 430,264    |
| 4  | Argentina                        | 466,667   | 506,843   | 574,605   | 598,862   | 739,816    |
| 5  | Armenia                          | 15,116    | 15,826    | 14,991    | 16,780    | 23,485     |
| 6  | Australia                        | 1,320,821 | 1,543,146 | 1,885,364 | 2,386,429 | 3,103,463  |
| 7  | Austria                          | 747,277   | 867,386   | 997,936   | 1,117,769 | 1,225,301  |
| 8  | Bahrain                          | 17,401    | 20,991    | 23,414    | 30,132    | 45,240     |
| 9  | Bangladesh                       | 53,417    | 66,472    | 91,401    | 129,805   | 185,330    |
| 10 | Barbados                         | 11,464    | 11,729    | 13,130    | 14,007    | 15,010     |
| 11 | Belgium                          | 817,777   | 931,848   | 1,060,901 | 1,196,425 | 1,354,303  |
| 12 | Belize                           | 1,350     | 1,887     | 2,394     | 2,946     | 3,499      |
| 13 | Benin                            | 6,537     | 7,244     | 8,481     | 10,500    | 13,066     |
| 14 | Bolivia (Plurinational State of) | 12,743    | 14,657    | 18,763    | 20,852    | 24,969     |
| 15 | Botswana                         | 9,332     | 13,887    | 20,166    | 28,013    | 37,144     |
| 16 | Brazil                           | 1,899,175 | 2,067,899 | 2,316,349 | 2,506,169 | 2,903,836  |
| 17 | Bulgaria                         | 95,413    | 90,613    | 85,813    | 94,908    | 120,397    |
| 18 | Burundi                          | 2,756     | 3,455     | 3,453     | 3,681     | 3,885      |
| 19 | Cambodia                         | 4,025     | 4,514     | 5,963     | 9,126     | 15,498     |
| 20 | Cameroon                         | 32,426    | 33,765    | 36,238    | 42,452    | 49,927     |
| 21 | Canada                           | 1,889,256 | 2,143,343 | 2,505,853 | 3,017,978 | 3,658,851  |
| 22 | Central African Republic         | 3,002     | 3,105     | 3,080     | 3,044     | 3,290      |
| 23 | Chile                            | 88,636    | 124,775   | 179,493   | 237,604   | 333,443    |
| 24 | China                            | 1,567,556 | 2,421,380 | 3,963,916 | 6,596,072 | 11,734,004 |
| 25 | Colombia                         | 276,942   | 338,743   | 382,932   | 420,501   | 531,831    |
| 26 | Congo                            | 15,538    | 19,636    | 20,857    | 22,106    | 29,980     |
| 27 | Costa Rica                       | 22,709    | 28,592    | 36,548    | 45,967    | 60,066     |
| 28 | Côte d'Ivoire                    | 30,072    | 29,124    | 30,961    | 32,483    | 33,540     |
| 29 | Croatia*                         | 94,315    | 93,542    | 107,005   | 130,837   | 165,199    |
| 30 | Cuba                             | 110,893   | 105,653   | 100,707   | 97,455    | 105,523    |
| 31 | Cyprus                           | 29,362    | 34,854    | 39,409    | 45,633    | 55,325     |
| 32 | Czech Republic**                 | 389,981   | 412,377   | 464,267   | 523,615   | 604,732    |
| 33 | Democratic Republic of the Congo | 21,160    | 19,591    | 17,641    | 18,259    | 21,675     |
| 34 | Denmark                          | 478,058   | 532,838   | 631,858   | 736,144   | 845,459    |
| 35 | Dominican Republic               | 32,871    | 39,692    | 56,115    | 70,911    | 92,528     |
| 36 | Ecuador                          | 83,572    | 93,913    | 100,664   | 115,064   | 138,340    |
| 37 | Egypt                            | 79,630    | 101,449   | 138,013   | 177,679   | 252,061    |

Produced capital change percentage change with respect to the base year, 1990

Produced capital per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010    | 2005    | 2000    | 1995    | 1990   |
|------|------|------|------|---------|---------|---------|---------|--------|
| -48  | -50  | -45  | -35  | 1,016   | 978     | 1,069   | 1,267   | 1,953  |
| 26   | 7    | -6   | -6   | 15,762  | 13,351  | 11,798  | 11,833  | 12,524 |
| 4    | -6   | -8   | -6   | 12,131  | 10,959  | 10,726  | 10,938  | 11,689 |
| 28   | 8    | 9    | 2    | 18,307  | 15,482  | 15,559  | 14,541  | 14,296 |
| 78   | 28   | 14   | 15   | 7,595   | 5,473   | 4,873   | 4,910   | 4,264  |
| 80   | 51   | 27   | 10   | 139,366 | 116,962 | 98,379  | 85,170  | 77,258 |
| 50   | 39   | 28   | 12   | 145,980 | 135,780 | 124,669 | 109,303 | 97,422 |
| 2    | 18   | 4    | 6    | 35,852  | 41,572  | 36,689  | 37,546  | 35,305 |
| 146  | 82   | 39   | 11   | 1,246   | 923     | 705     | 566     | 507    |
| 24   | 17   | 11   | 1    | 54,916  | 51,783  | 49,080  | 44,526  | 44,175 |
| 54   | 40   | 27   | 12   | 126,428 | 114,884 | 104,258 | 92,446  | 82,197 |
| 58   | 48   | 35   | 21   | 11,230  | 10,485  | 9,556   | 8,564   | 7,094  |
| 8    | 0    | -5   | -6   | 1,476   | 1,375   | 1,301   | 1,282   | 1,370  |
| 31   | 19   | 18   | 3    | 2,515   | 2,280   | 2,259   | 1,962   | 1,914  |
| 174  | 121  | 70   | 30   | 18,507  | 14,935  | 11,472  | 8,758   | 6,751  |
| 17   | 6    | 5    | 1    | 14,896  | 13,475  | 13,280  | 12,777  | 12,691 |
| 48   | 13   | -1   | 0    | 16,065  | 12,263  | 10,718  | 10,843  | 10,819 |
| -6   | 3    | 10   | 15   | 463     | 508     | 542     | 568     | 492    |
| 160  | 62   | 13   | -4   | 1,096   | 683     | 479     | 404     | 422    |
| -4   | -9   | -13  | -9   | 2,547   | 2,418   | 2,311   | 2,422   | 2,662  |
| 58   | 37   | 20   | 7    | 107,561 | 93,484  | 81,711  | 73,146  | 68,202 |
| -27  | -26  | -19  | -9   | 748     | 758     | 832     | 933     | 1,023  |
| 190  | 117  | 73   | 29   | 19,484  | 14,575  | 11,640  | 8,659   | 6,721  |
| 539  | 269  | 128  | 46   | 8,748   | 5,044   | 3,123   | 1,995   | 1,369  |
| 38   | 17   | 15   | 11   | 11,488  | 9,770   | 9,630   | 9,293   | 8,341  |
| 14   | -4   | 2    | 10   | 7,416   | 6,257   | 6,651   | 7,185   | 6,504  |
| 74   | 44   | 26   | 11   | 12,893  | 10,667  | 9,325   | 8,242   | 7,396  |
| -29  | -25  | -22  | -17  | 1,699   | 1,803   | 1,867   | 1,984   | 2,402  |
| 81   | 42   | 15   | -3   | 37,517  | 29,455  | 23,750  | 20,035  | 20,737 |
| -11  | -17  | -14  | -8   | 9,373   | 8,659   | 9,069   | 9,692   | 10,491 |
| 31   | 15   | 9    | 6    | 50,129  | 44,194  | 41,779  | 40,745  | 38,299 |
| 52   | 35   | 20   | 6    | 57,632  | 51,231  | 45,326  | 39,962  | 37,812 |
| -43  | -45  | -39  | -24  | 329     | 318     | 355     | 445     | 581    |
| 64   | 46   | 27   | 9    | 152,331 | 135,834 | 118,336 | 101,816 | 92,989 |
| 104  | 68   | 43   | 10   | 9,321   | 7,654   | 6,531   | 5,014   | 4,569  |
| 17   | 5    | 0    | 1    | 9,564   | 8,570   | 8,154   | 8,249   | 8,145  |
| 122  | 71   | 46   | 17   | 3,107   | 2,394   | 2,040   | 1,635   | 1,401  |

## Produced capital

#### Produced capital level in millions of constant 2005 US\$

|    | Country/Year                     | 1990       | 1995       | 2000       | 2005       | 2010       |
|----|----------------------------------|------------|------------|------------|------------|------------|
| 38 | El Salvador                      | 23,942     | 27,968     | 33,493     | 39,359     | 44,902     |
| 39 | Estonia                          | 30,788     | 31,246     | 34,552     | 44,631     | 56,961     |
| 40 | Fiji                             | 4,753      | 5,208      | 5,793      | 6,674      | 7,437      |
| 41 | Finland                          | 496,242    | 514,190    | 564,731    | 631,803    | 704,663    |
| 42 | France                           | 4,611,452  | 5,142,790  | 5,729,333  | 6,479,550  | 7,290,809  |
| 43 | Gabon                            | 32,901     | 33,220     | 35,523     | 36,801     | 39,183     |
| 44 | Gambia                           | 356        | 396        | 439        | 867        | 1,660      |
| 45 | Germany                          | 6,580,914  | 7,527,739  | 8,446,071  | 9,124,010  | 9,856,172  |
| 46 | Ghana                            | 34,449     | 37,893     | 40,787     | 46,875     | 61,339     |
| 47 | Greece                           | 523,442    | 551,922    | 611,918    | 727,129    | 844,548    |
| 48 | Guatemala                        | 42,021     | 47,813     | 58,703     | 70,172     | 82,604     |
| 49 | Guyana                           | 3,418      | 4,111      | 4,848      | 5,285      | 6,113      |
| 50 | Haiti                            | 6,283      | 6,344      | 7,096      | 8,281      | 9,431      |
| 51 | Honduras                         | 15,581     | 19,808     | 25,500     | 30,823     | 38,695     |
| 52 | Hungary                          | 291,623    | 304,260    | 329,291    | 374,393    | 415,982    |
| 53 | Iceland                          | 28,655     | 30,739     | 36,330     | 44,454     | 52,969     |
| 54 | India                            | 848,080    | 1,114,016  | 1,515,896  | 2,171,823  | 3,456,372  |
| 55 | Indonesia                        | 301,769    | 471,588    | 640,925    | 787,249    | 1,025,167  |
| 56 | Iran (Islamic Republic of)       | 466,200    | 488,275    | 526,761    | 635,378    | 782,036    |
| 57 | Iraq                             | 32,169     | 28,621     | 25,157     | 28,929     | 44,031     |
| 58 | Ireland                          | 279,911    | 313,381    | 407,400    | 540,659    | 651,173    |
| 59 | Israel                           | 185,414    | 242,412    | 310,956    | 356,248    | 425,063    |
| 60 | Italy                            | 4,384,260  | 4,888,355  | 5,439,510  | 6,112,982  | 6,648,672  |
| 61 | Jamaica                          | 30,668     | 36,868     | 41,299     | 46,988     | 50,565     |
| 62 | Japan                            | 13,570,113 | 16,288,603 | 18,399,084 | 19,795,727 | 20,668,040 |
| 63 | Jordan                           | 27,492     | 33,401     | 37,917     | 43,404     | 52,451     |
| 64 | Kazakhstan*                      | 273,576    | 284,303    | 260,495    | 265,334    | 326,750    |
| 65 | Kenya                            | 30,411     | 33,326     | 38,261     | 44,654     | 59,364     |
| 66 | Kuwait                           | 49,530     | 58,821     | 71,836     | 98,390     | 149,139    |
| 67 | Kyrgyzstan*                      | 9,062      | 9,070      | 8,831      | 8,903      | 10,551     |
| 68 | Lao People's Democratic Republic | 1,249      | 1,765      | 3,140      | 6,106      | 11,588     |
| 69 | Latvia                           | 33,141     | 31,445     | 34,147     | 44,339     | 57,622     |
| 70 | Lesotho                          | 3,613      | 5,364      | 6,880      | 7,354      | 8,130      |
| 71 | Liberia                          | 3,409      | 2,901      | 2,566      | 2,367      | 2,598      |
| 72 | Lithuania*                       | 68,473     | 67,526     | 70,265     | 79,512     | 96,027     |
| 73 | Luxembourg                       | 46,734     | 56,914     | 71,750     | 92,310     | 115,473    |
| 74 | Malawi                           | 8,344      | 8,605      | 8,412      | 8,610      | 11,751     |

Produced capital change percentage change with respect to the base year, 1990

Produced capital per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 4,490   | 4,878   | 5,638   | 6,505   | 7,250   | 9    | 26   | 45   | 61   |
| 19,640  | 21,688  | 25,206  | 33,162  | 42,472  | 10   | 28   | 69   | 116  |
| 6,525   | 6,714   | 7,137   | 8,114   | 8,641   | 3    | 9    | 24   | 32   |
| 99,518  | 100,668 | 109,161 | 120,473 | 131,356 | 1    | 10   | 21   | 32   |
| 81,319  | 88,907  | 97,029  | 106,228 | 116,119 | 9    | 19   | 31   | 43   |
| 35,413  | 30,552  | 28,757  | 26,848  | 26,027  | -14  | -19  | -24  | -27  |
| 369     | 352     | 339     | 577     | 960     | -5   | -8   | 56   | 160  |
| 83,199  | 91,881  | 102,564 | 110,539 | 119,755 | 10   | 23   | 33   | 44   |
| 2,329   | 2,229   | 2,128   | 2,166   | 2,515   | -4   | -9   | -7   | 8    |
| 51,517  | 51,718  | 55,695  | 65,020  | 74,348  | 0    | 8    | 26   | 44   |
| 4,709   | 4,774   | 5,224   | 5,518   | 5,741   | 1    | 11   | 17   | 22   |
| 4,715   | 5,647   | 6,613   | 7,082   | 8,103   | 20   | 40   | 50   | 72   |
| 882     | 805     | 821     | 886     | 944     | -9   | -7   | 0    | 7    |
| 3,187   | 3,553   | 4,101   | 4,481   | 5,091   | 11   | 29   | 41   | 60   |
| 28,105  | 29,450  | 32,250  | 37,117  | 41,666  | 5    | 15   | 32   | 48   |
| 112,463 | 114,920 | 129,192 | 149,807 | 165,457 | 2    | 15   | 33   | 47   |
| 971     | 1,155   | 1,438   | 1,905   | 2,822   | 19   | 48   | 96   | 191  |
| 1,637   | 2,365   | 3,003   | 3,463   | 4,274   | 44   | 83   | 112  | 161  |
| 8,496   | 8,171   | 8,062   | 9,112   | 10,572  | -4   | -5   | 7    | 24   |
| 1,852   | 1,411   | 1,054   | 1,057   | 1,390   | -24  | -43  | -43  | -25  |
| 79,268  | 86,786  | 107,104 | 130,027 | 145,680 | 9    | 35   | 64   | 84   |
| 41,204  | 45,463  | 51,697  | 53,940  | 57,298  | 10   | 25   | 31   | 39   |
| 77,144  | 85,809  | 95,453  | 104,190 | 109,803 | 11   | 24   | 35   | 42   |
| 12,968  | 14,978  | 15,997  | 17,522  | 18,447  | 16   | 23   | 35   | 42   |
| 111,002 | 130,846 | 146,349 | 156,621 | 163,337 | 18   | 32   | 41   | 47   |
| 8,049   | 7,622   | 7,855   | 8,125   | 8,477   | -5   | -2   | 1    | 5    |
| 16,567  | 17,852  | 17,417  | 17,489  | 20,388  | 8    | 5    | 6    | 23   |
| 1,297   | 1,215   | 1,224   | 1,254   | 1,465   | -6   | -6   | -3   | 13   |
| 23,725  | 36,133  | 37,014  | 43,458  | 54,495  | 52   | 56   | 83   | 130  |
| 2,040   | 1,975   | 1,782   | 1,766   | 1,978   | -3   | -13  | -13  | -3   |
| 298     | 368     | 591     | 1,061   | 1,869   | 24   | 98   | 256  | 527  |
| 12,441  | 12,618  | 14,318  | 19,232  | 25,586  | 1    | 15   | 55   | 106  |
| 2,204   | 2,989   | 3,503   | 3,560   | 3,744   | 36   | 59   | 61   | 70   |
| 1,602   | 1,385   | 901     | 744     | 650     | -14  | -44  | -54  | -59  |
| 18,516  | 18,607  | 20,075  | 23,278  | 28,892  | 0    | 8    | 26   | 56   |
| 122,588 | 139,659 | 164,757 | 201,903 | 227,556 | 14   | 34   | 65   | 86   |
| 889     | 871     | 749     | 671     | 789     | -2   | -16  | -25  | -11  |

## Produced capital

## Produced capital level in millions of constant 2005 US\$

|     | Country/Year        | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|---------------------|-----------|-----------|-----------|-----------|-----------|
| 75  | Malaysia            | 119,981   | 208,507   | 299,254   | 365,961   | 450,159   |
| 76  | Maldives            | 773       | 1,178     | 1,797     | 2,831     | 4,539     |
| 77  | Mali                | 6,107     | 7,561     | 8,997     | 10,820    | 13,718    |
| 78  | Malta               | 6,702     | 9,251     | 12,662    | 15,430    | 17,522    |
| 79  | Mauritania          | 4,819     | 4,903     | 4,932     | 6,995     | 10,228    |
| 80  | Mauritius           | 7,514     | 10,348    | 13,696    | 17,422    | 22,463    |
| 81  | Mexico              | 1,359,588 | 1,592,960 | 1,899,978 | 2,265,913 | 2,757,829 |
| 82  | Mongolia            | 10,681    | 11,578    | 12,553    | 13,190    | 23,047    |
| 83  | Morocco             | 113,846   | 131,147   | 155,911   | 193,023   | 258,846   |
| 84  | Mozambique          | 6,210     | 7,249     | 9,729     | 13,275    | 18,672    |
| 85  | Myanmar             | 2,630     | 3,336     | 5,199     | 9,028     | 22,713    |
| 86  | Namibia             | 15,408    | 16,788    | 18,418    | 21,003    | 26,709    |
| 87  | Nepal               | 9,269     | 12,357    | 16,204    | 20,543    | 25,786    |
| 88  | Netherlands         | 1,267,929 | 1,433,637 | 1,699,609 | 1,940,251 | 2,190,994 |
| 89  | New Zealand         | 210,571   | 231,818   | 269,984   | 329,724   | 386,182   |
| 90  | Nicaragua           | 18,472    | 18,372    | 20,428    | 22,444    | 24,796    |
| 91  | Niger               | 11,618    | 10,990    | 10,776    | 11,442    | 13,975    |
| 92  | Nigeria             | 220,009   | 206,845   | 195,235   | 189,725   | 209,513   |
| 93  | Norway              | 627,418   | 661,494   | 751,299   | 838,474   | 993,996   |
| 94  | Pakistan            | 165,804   | 208,457   | 246,979   | 282,919   | 343,841   |
| 95  | Panama              | 18,003    | 22,280    | 29,409    | 34,220    | 50,042    |
| 96  | Papua New Guinea    | 7,300     | 8,011     | 9,156     | 11,170    | 13,020    |
| 97  | Paraguay            | 18,847    | 23,580    | 26,493    | 27,618    | 30,909    |
| 98  | Peru                | 168,947   | 181,997   | 211,372   | 230,280   | 304,429   |
| 99  | Philippines         | 180,541   | 212,159   | 255,815   | 297,862   | 352,574   |
| 100 | Poland              | 428,126   | 471,573   | 618,486   | 742,648   | 960,224   |
| 101 | Portugal            | 393,042   | 465,610   | 579,805   | 684,693   | 751,653   |
| 102 | Qatar               | 25,373    | 28,560    | 39,192    | 72,109    | 164,883   |
| 103 | Republic of Korea   | 841,614   | 1,475,163 | 2,133,910 | 2,808,704 | 3,494,869 |
| 104 | Republic of Moldova | 35,957    | 33,009    | 29,210    | 26,561    | 26,143    |
| 105 | Romania             | 274,003   | 274,100   | 287,342   | 322,425   | 416,065   |
| 106 | Russian Federation* | 3,415,470 | 3,445,515 | 3,163,169 | 3,097,099 | 3,402,771 |
| 107 | Rwanda              | 3,115     | 3,280     | 3,584     | 4,425     | 6,936     |
| 108 | Saudi Arabia        | 365,249   | 426,815   | 494,168   | 601,057   | 858,912   |
| 109 | Senegal             | 13,024    | 15,050    | 18,576    | 23,518    | 30,893    |
| 110 | Serbia              | 62,521    | 61,545    | 60,258    | 66,004    | 81,970    |

Produced capital change percentage change with respect to the base year, 1990

Produced capital per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010    | 2005    | 2000    | 1995    | 1990    |
|------|------|------|------|---------|---------|---------|---------|---------|
| 141  | 113  | 94   | 53   | 15,850  | 14,021  | 12,780  | 10,063  | 6,589   |
| 308  | 172  | 87   | 34   | 14,370  | 9,588   | 6,578   | 4,735   | 3,522   |
| 27   | 17   | 13   | 9    | 893     | 821     | 797     | 770     | 704     |
| 131  | 107  | 75   | 31   | 42,068  | 37,700  | 31,861  | 23,931  | 18,234  |
| 22   | -5   | -23  | -11  | 2,956   | 2,296   | 1,866   | 2,139   | 2,415   |
| 144  | 95   | 61   | 28   | 17,291  | 13,863  | 11,451  | 9,101   | 7,092   |
| 51   | 32   | 18   | 7    | 24,315  | 21,279  | 19,007  | 17,264  | 16,127  |
| 72   | 6    | 7    | 3    | 8,362   | 5,178   | 5,206   | 5,021   | 4,871   |
| 76   | 38   | 18   | 6    | 8,101   | 6,351   | 5,415   | 4,870   | 4,594   |
| 74   | 39   | 17   | -1   | 798     | 639     | 535     | 455     | 458     |
| 607  | 191  | 73   | 18   | 474     | 195     | 116     | 79      | 67      |
| 7    | -7   | -11  | -7   | 11,697  | 10,098  | 9,715   | 10,166  | 10,890  |
| 77   | 55   | 37   | 18   | 861     | 753     | 664     | 572     | 486     |
| 55   | 40   | 26   | 9    | 131,884 | 118,994 | 107,144 | 92,959  | 85,143  |
| 43   | 29   | 13   | 2    | 88,409  | 79,757  | 69,980  | 63,079  | 61,970  |
| -4   | -8   | -10  | -12  | 4,284   | 4,138   | 4,026   | 3,962   | 4,483   |
| -40  | -41  | -34  | -20  | 901     | 881     | 987     | 1,197   | 1,492   |
| -41  | -40  | -30  | -17  | 1,322   | 1,357   | 1,578   | 1,880   | 2,255   |
| 38   | 23   | 13   | 3    | 203,558 | 181,358 | 167,295 | 151,750 | 147,924 |
| 34   | 20   | 15   | 10   | 1,981   | 1,783   | 1,709   | 1,637   | 1,482   |
| 91   | 42   | 34   | 12   | 14,229  | 10,567  | 9,948   | 8,323   | 7,452   |
| 8    | 4    | -3   | -3   | 1,898   | 1,832   | 1,702   | 1,699   | 1,756   |
| 8    | 5    | 12   | 11   | 4,789   | 4,683   | 4,958   | 4,917   | 4,441   |
| 34   | 7    | 5    | -2   | 10,470  | 8,356   | 8,173   | 7,638   | 7,791   |
| 29   | 19   | 13   | 5    | 3,781   | 3,482   | 3,309   | 3,063   | 2,929   |
| 123  | 73   | 44   | 9    | 25,086  | 19,459  | 16,147  | 12,283  | 11,250  |
| 78   | 64   | 42   | 16   | 70,409  | 64,939  | 56,095  | 45,986  | 39,599  |
| 75   | 64   | 24   | 6    | 93,748  | 87,832  | 66,320  | 56,963  | 53,560  |
| 270  | 205  | 137  | 69   | 72,532  | 59,704  | 46,402  | 33,029  | 19,581  |
| -11  | -14  | -14  | -8   | 7,317   | 7,051   | 7,113   | 7,608   | 8,240   |
| 64   | 25   | 10   | 2    | 19,364  | 14,809  | 12,948  | 12,085  | 11,807  |
| 4    | -6   | -6   | 1    | 23,803  | 21,531  | 21,554  | 23,171  | 22,973  |
| 49   | 10   | 1    | 34   | 653     | 481     | 443     | 589     | 438     |
| 38   | 10   | 9    | 2    | 31,292  | 25,001  | 24,653  | 23,081  | 22,631  |
| 38   | 20   | 9    | 0    | 2,485   | 2,163   | 1,954   | 1,798   | 1,798   |
| 27   | 2    | -9   | -8   | 8,317   | 6,697   | 5,946   | 6,031   | 6,534   |

## Produced capital

Produced capital level in millions of constant 2005 US\$

|     | Country/Year                       | 1990       | 1995       | 2000       | 2005       | 2010       |
|-----|------------------------------------|------------|------------|------------|------------|------------|
| 111 | Sierra Leone                       | 1,138      | 1,157      | 1,060      | 1,239      | 1,565      |
| 112 | Singapore                          | 130,321    | 187,836    | 280,570    | 349,238    | 461,252    |
| 113 | Slovakia**                         | 152,085    | 156,514    | 176,657    | 195,095    | 224,086    |
| 114 | Slovenia*                          | 77,935     | 81,396     | 97,081     | 117,971    | 143,632    |
| 115 | South Africa                       | 453,800    | 470,590    | 512,459    | 576,926    | 726,246    |
| 116 | Spain                              | 2,344,858  | 2,732,645  | 3,251,642  | 4,029,463  | 4,804,877  |
| 117 | Sri Lanka                          | 35,309     | 42,106     | 51,793     | 63,524     | 85,686     |
| 118 | Sudan (former)                     | 8,234      | 9,541      | 16,528     | 36,239     | 67,470     |
| 119 | Swaziland                          | 9,927      | 12,659     | 13,881     | 12,910     | 12,133     |
| 120 | Sweden                             | 760,258    | 814,967    | 900,452    | 1,012,899  | 1,166,991  |
| 121 | Switzerland                        | 920,651    | 1,008,601  | 1,120,851  | 1,241,788  | 1,388,524  |
| 122 | Syrian Arab Republic               | 42,567     | 49,821     | 57,484     | 71,061     | 89,093     |
| 123 | Tajikistan*                        | 13,367     | 13,024     | 11,605     | 10,475     | 10,032     |
| 124 | Thailand                           | 308,166    | 519,936    | 634,845    | 711,558    | 824,125    |
| 125 | Тодо                               | 7,544      | 6,996      | 6,831      | 7,032      | 7,654      |
| 126 | Trinidad and Tobago                | 49,845     | 46,840     | 47,590     | 53,317     | 55,227     |
| 127 | Tunisia                            | 66,189     | 76,267     | 88,500     | 103,585    | 123,402    |
| 128 | Turkey                             | 499,362    | 664,497    | 870,502    | 1,052,519  | 1,367,017  |
| 129 | Uganda                             | 9,133      | 10,973     | 14,442     | 19,428     | 29,208     |
| 130 | Ukraine*                           | 584,144    | 592,168    | 529,811    | 503,488    | 509,202    |
| 131 | United Arab Emirates               | 281,987    | 317,808    | 362,856    | 426,769    | 579,019    |
| 132 | United Kingdom                     | 3,502,941  | 3,929,715  | 4,591,999  | 5,406,740  | 6,247,208  |
| 133 | United Republic of Tanzania        | 11,822     | 18,330     | 23,007     | 31,649     | 49,562     |
| 134 | United States of America           | 17,158,834 | 19,966,588 | 25,108,673 | 30,919,448 | 35,543,091 |
| 135 | Uruguay                            | 33,979     | 39,033     | 46,291     | 48,909     | 59,070     |
| 136 | Venezuela (Bolivarian Republic of) | 495,509    | 519,957    | 538,862    | 546,101    | 644,299    |
| 137 | Vietnam                            | 22,952     | 39,526     | 73,387     | 126,666    | 218,664    |
| 138 | Yemen                              | 16,334     | 20,165     | 29,106     | 38,859     | 45,956     |
| 139 | Zambia                             | 14,278     | 13,021     | 15,342     | 20,083     | 28,694     |
| 140 | Zimbabwe                           | 3,409      | 4.030      | 4.430      | 4.980      | 4.598      |

Produced capital change percentage change with respect to the base year, 1990

Produced capital per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010    | 2005    | 2000    | 1995    | 1990    |
|------|------|------|------|---------|---------|---------|---------|---------|
| -7   | -16  | -10  | 4    | 267     | 240     | 256     | 297     | 286     |
| 110  | 89   | 66   | 25   | 90,683  | 81,865  | 71,587  | 53,950  | 43,201  |
| 43   | 26   | 14   | 2    | 41,025  | 36,025  | 32,685  | 29,152  | 28,615  |
| 76   | 46   | 21   | 3    | 70,766  | 58,924  | 48,898  | 41,397  | 40,270  |
| 17   | -2   | -7   | -8   | 14,486  | 12,071  | 11,449  | 11,366  | 12,334  |
| 73   | 54   | 34   | 15   | 104,279 | 92,855  | 80,709  | 69,309  | 60,296  |
| 102  | 57   | 36   | 13   | 4,108   | 3,201   | 2,763   | 2,310   | 2,037   |
| 398  | 204  | 56   | 2    | 1,549   | 943     | 483     | 317     | 311     |
| -11  | 2    | 13   | 14   | 10,230  | 11,684  | 13,048  | 13,137  | 11,504  |
| 40   | 26   | 14   | 4    | 124,417 | 112,179 | 101,629 | 92,327  | 88,827  |
| 31   | 21   | 13   | 4    | 181,167 | 167,470 | 156,371 | 143,702 | 137,951 |
| 26   | 11   | 4    | 2    | 4,365   | 3,844   | 3,595   | 3,516   | 3,454   |
| -41  | -34  | -24  | -9   | 1,458   | 1,623   | 1,880   | 2,255   | 2,467   |
| 121  | 98   | 86   | 61   | 11,923  | 10,668  | 10,052  | 8,716   | 5,400   |
| -38  | -37  | -31  | -17  | 1,270   | 1,300   | 1,425   | 1,712   | 2,058   |
| 0    | -1   | -10  | -9   | 41,169  | 40,533  | 36,833  | 37,132  | 41,008  |
| 46   | 30   | 16   | 6    | 11,774  | 10,450  | 9,359   | 8,535   | 8,057   |
| 104  | 67   | 48   | 22   | 18,790  | 15,446  | 13,681  | 11,289  | 9,225   |
| 69   | 32   | 16   | 2    | 874     | 683     | 596     | 527     | 516     |
| -1   | -5   | -4   | 2    | 11,204  | 10,730  | 10,836  | 11,583  | 11,306  |
| -51  | -33  | -23  | -13  | 77,082  | 104,874 | 119,617 | 135,322 | 155,911 |
| 64   | 47   | 27   | 11   | 100,704 | 89,809  | 77,997  | 67,757  | 61,225  |
| 138  | 76   | 46   | 32   | 1,105   | 815     | 676     | 612     | 464     |
| 69   | 54   | 31   | 11   | 114,513 | 104,169 | 88,881  | 74,971  | 67,731  |
| 60   | 35   | 28   | 11   | 17,535  | 14,720  | 13,947  | 12,109  | 10,929  |
| -12  | -19  | -12  | -6   | 22,233  | 20,481  | 22,131  | 23,597  | 25,172  |
| 628  | 345  | 172  | 56   | 2,489   | 1,523   | 932     | 534     | 342     |
| 40   | 38   | 20   | -3   | 1,911   | 1,882   | 1,642   | 1,331   | 1,367   |
| 21   | -4   | -17  | -20  | 2,192   | 1,752   | 1,504   | 1,460   | 1,817   |
| 12   | 22   | 9    | 6    | 366     | 396     | 354     | 345     | 326     |

## Natural capital

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 1  | Afghanistan                      | 8,472     | 8,416     | 8,368     | 8,360     | 8,347     |
| 2  | Albania                          | 30,646    | 30,430    | 30,429    | 29,752    | 30,685    |
| 3  | Algeria                          | 540,899   | 507,941   | 468,762   | 423,063   | 371,680   |
| 4  | Argentina                        | 696,148   | 675,818   | 650,429   | 642,205   | 635,909   |
| 5  | Armenia                          | 2,791     | 2,626     | 2,461     | 2,294     | 2,026     |
| 6  | Australia                        | 3,005,926 | 2,974,354 | 2,927,251 | 2,860,514 | 2,717,359 |
| 7  | Austria                          | 60,663    | 61,456    | 62,411    | 62,729    | 60,822    |
| 8  | Bahrain                          | 4,849     | 3,931     | 3,025     | 2,140     | 1,283     |
| 9  | Bangladesh                       | 41,415    | 38,657    | 38,160    | 37,188    | 36,855    |
| 10 | Barbados                         | 467       | 441       | 395       | 339       | 284       |
| 11 | Belgium                          | 3,369     | 3,907     | 4,445     | 4,874     | 5,021     |
| 12 | Belize                           | 13,805    | 13,481    | 13,091    | 12,696    | 12,308    |
| 13 | Benin                            | 36,726    | 34,478    | 32,231    | 30,626    | 28,999    |
| 14 | Bolivia (Plurinational State of) | 938,941   | 921,240   | 902,539   | 882,454   | 859,094   |
| 15 | Botswana                         | 67,952    | 65,268    | 62,593    | 59,870    | 57,173    |
| 16 | Brazil                           | 5,061,443 | 4,956,971 | 4,829,135 | 4,705,799 | 4,567,813 |
| 17 | Bulgaria                         | 70,918    | 69,862    | 68,093    | 68,169    | 68,216    |
| 18 | Burundi                          | 5,559     | 5,112     | 4,860     | 4,836     | 4,711     |
| 19 | Cambodia                         | 98,497    | 94,347    | 90,519    | 87,515    | 84,563    |
| 20 | Cameroon                         | 247,702   | 235,200   | 223,070   | 211,603   | 200,549   |
| 21 | Canada                           | 4,750,302 | 4,661,655 | 4,559,587 | 4,450,099 | 4,344,908 |
| 22 | Central African Republic         | 181,088   | 179,914   | 178,739   | 177,574   | 176,399   |
| 23 | Chile                            | 271,477   | 264,898   | 259,105   | 259,238   | 250,413   |
| 24 | China                            | 7,792,499 | 7,714,637 | 7,464,322 | 7,099,247 | 6,788,988 |
| 25 | Colombia                         | 1,170,032 | 1,149,137 | 1,129,043 | 1,092,203 | 1,066,561 |
| 26 | Congo                            | 251,360   | 247,651   | 242,520   | 237,569   | 232,601   |
| 27 | Costa Rica                       | 36,996    | 34,050    | 31,613    | 31,697    | 33,128    |
| 28 | Côte d'Ivoire                    | 72,258    | 72,842    | 72,459    | 72,584    | 71,938    |
| 29 | Croatia*                         | 20,902    | 21,099    | 21,472    | 21,944    | 22,478    |
| 30 | Cuba                             | 26,193    | 28,556    | 30,475    | 32,002    | 31,441    |
| 31 | Cyprus                           | 1,324     | 1,313     | 1,333     | 1,402     | 1,304     |
| 32 | Czech Republic**                 | 31,489    | 28,138    | 23,255    | 18,656    | 14,373    |
| 33 | Democratic Republic of the Congo | 989,199   | 979,165   | 969,137   | 959,191   | 949,252   |
| 34 | Denmark                          | 44,339    | 40,465    | 34,308    | 26,598    | 19,693    |
| 35 | Dominican Republic               | 30,398    | 30,509    | 29,851    | 29,437    | 29,260    |
| 36 | Ecuador                          | 223,882   | 209,852   | 194,445   | 176,787   | 159,292   |
| 37 | Egypt                            | 145,246   | 128,885   | 111,100   | 95,301    | 77,702    |

Natural capital change percentage change with respect to the base year, 1990

Natural capital per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 650     | 425     | 366     | 303     | 266     | -35  | -44  | -53  | -59  |
| 9,316   | 9,688   | 9,906   | 9,470   | 9,576   | 4    | 6    | 2    | 3    |
| 21,380  | 17,954  | 15,352  | 12,864  | 10,479  | -16  | -28  | -40  | -51  |
| 21,326  | 19,389  | 17,612  | 16,603  | 15,736  | -9   | -17  | -22  | -26  |
| 787     | 815     | 800     | 748     | 655     | 3    | 2    | -5   | -17  |
| 175,825 | 164,162 | 152,745 | 140,197 | 122,028 | -7   | -13  | -20  | -31  |
| 7,909   | 7,744   | 7,797   | 7,620   | 7,246   | -2   | -1   | -4   | -8   |
| 9,837   | 7,031   | 4,739   | 2,953   | 1,017   | -29  | -52  | -70  | -90  |
| 393     | 329     | 294     | 265     | 248     | -16  | -25  | -33  | -37  |
| 1,799   | 1,673   | 1,477   | 1,253   | 1,039   | -7   | -18  | -30  | -42  |
| 339     | 388     | 437     | 468     | 469     | 14   | 29   | 38   | 38   |
| 72,565  | 61,192  | 52,255  | 45,189  | 39,496  | -16  | -28  | -38  | -46  |
| 7,694   | 6,101   | 4,945   | 4,012   | 3,277   | -21  | -36  | -48  | -57  |
| 141,015 | 123,310 | 108,645 | 96,478  | 86,516  | -13  | -23  | -32  | -39  |
| 49,156  | 41,162  | 35,606  | 31,919  | 28,488  | -16  | -28  | -35  | -42  |
| 33,822  | 30,627  | 27,686  | 25,302  | 23,431  | -9   | -18  | -25  | -31  |
| 8,042   | 8,360   | 8,505   | 8,808   | 9,102   | 4    | 6    | 10   | 13   |
| 992     | 840     | 762     | 667     | 562     | -15  | -23  | -33  | -43  |
| 10,333  | 8,447   | 7,272   | 6,552   | 5,981   | -18  | -30  | -37  | -42  |
| 20,335  | 16,872  | 14,228  | 12,055  | 10,233  | -17  | -30  | -41  | -50  |
| 171,486 | 159,089 | 148,679 | 137,845 | 127,729 | -7   | -13  | -20  | -26  |
| 61,704  | 54,065  | 48,287  | 44,196  | 40,081  | -12  | -22  | -28  | -35  |
| 20,585  | 18,384  | 16,803  | 15,902  | 14,632  | -11  | -18  | -23  | -29  |
| 6,805   | 6,355   | 5,882   | 5,429   | 5,061   | -7   | -14  | -20  | -26  |
| 35,238  | 31,524  | 28,393  | 25,376  | 23,038  | -11  | -19  | -28  | -35  |
| 105,220 | 90,625  | 77,340  | 67,240  | 57,533  | -14  | -26  | -36  | -45  |
| 12,050  | 9,816   | 8,066   | 7,355   | 7,111   | -19  | -33  | -39  | -41  |
| 5,772   | 4,963   | 4,370   | 4,028   | 3,645   | -14  | -24  | -30  | -37  |
| 4,596   | 4,519   | 4,766   | 4,940   | 5,105   | -2   | 4    | 7    | 11   |
| 2,478   | 2,620   | 2,744   | 2,844   | 2,793   | 6    | 11   | 15   | 13   |
| 1,726   | 1,534   | 1,413   | 1,358   | 1,182   | -11  | -18  | -21  | -32  |
| 3,053   | 2,727   | 2,270   | 1,825   | 1,370   | -11  | -26  | -40  | -55  |
| 27,171  | 22,220  | 19,529  | 16,705  | 14,390  | -18  | -28  | -39  | -47  |
| 8,624   | 7,732   | 6,425   | 4,908   | 3,548   | -10  | -25  | -43  | -59  |
| 4,225   | 3,854   | 3,474   | 3,177   | 2,947   | -9   | -18  | -25  | -30  |
| 21,820  | 18,433  | 15,751  | 13,167  | 11,012  | -16  | -28  | -40  | -50  |
| 2,555   | 2,077   | 1,642   | 1,284   | 958     | -19  | -36  | -50  | -63  |

## Natural capital

Natural capital level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 38 | El Salvador                      | 7,762     | 7,786     | 7,810     | 7,908     | 7,624     |
| 39 | Estonia                          | 17,255    | 17,949    | 18,643    | 18,612    | 18,349    |
| 40 | Fiji                             | 5,756     | 5,891     | 5,806     | 5,780     | 5,751     |
| 41 | Finland                          | 163,683   | 165,811   | 168,566   | 162,940   | 163,050   |
| 42 | France                           | 268,682   | 269,002   | 270,405   | 276,935   | 277,134   |
| 43 | Gabon                            | 156,482   | 151,131   | 145,317   | 141,070   | 136,992   |
| 44 | Gambia                           | 2,328     | 2,369     | 2,416     | 2,459     | 2,523     |
| 45 | Germany                          | 1,593,574 | 1,540,903 | 1,500,887 | 1,460,914 | 1,423,171 |
| 46 | Ghana                            | 63,571    | 60,439    | 58,468    | 55,450    | 52,414    |
| 47 | Greece                           | 229,940   | 225,258   | 211,562   | 138,414   | 196,886   |
| 48 | Guatemala                        | 36,312    | 34,058    | 31,621    | 29,312    | 26,419    |
| 49 | Guyana                           | 171,686   | 171,665   | 171,641   | 171,624   | 171,608   |
| 50 | Haiti                            | 653       | 606       | 559       | 510       | 459       |
| 51 | Honduras                         | 86,606    | 78,811    | 68,993    | 63,915    | 58,420    |
| 52 | Hungary                          | 50,079    | 47,258    | 44,459    | 43,175    | 40,164    |
| 53 | Iceland                          | 76,624    | 76,542    | 76,139    | 75,856    | 64,129    |
| 54 | India                            | 2,484,106 | 2,445,698 | 2,407,511 | 2,345,221 | 2,277,454 |
| 55 | Indonesia                        | 2,116,409 | 1,975,114 | 1,874,495 | 1,805,107 | 1,711,576 |
| 56 | Iran (Islamic Republic of)       | 3,202,181 | 3,130,867 | 3,046,171 | 2,931,607 | 2,826,571 |
| 57 | Iraq                             | 1,052,440 | 1,045,572 | 1,020,938 | 994,817   | 964,393   |
| 58 | Ireland                          | 39,801    | 31,059    | 30,854    | 29,836    | 31,330    |
| 59 | Israel                           | 6,682     | 6,578     | 6,453     | 5,974     | 5,711     |
| 60 | Italy                            | 325,144   | 318,858   | 332,179   | 332,809   | 338,927   |
| 61 | Jamaica                          | 30,081    | 30,992    | 29,964    | 29,229    | 28,174    |
| 62 | Japan                            | 401,054   | 398,280   | 398,614   | 385,297   | 380,285   |
| 63 | Jordan                           | 7,775     | 7,813     | 7,581     | 7,313     | 7,155     |
| 64 | Kazakhstan*                      | 886,369   | 874,413   | 859,730   | 836,861   | 806,053   |
| 65 | Kenya                            | 41,677    | 41,343    | 40,243    | 39,787    | 39,565    |
| 66 | Kuwait                           | 1,273,616 | 1,244,676 | 1,203,211 | 1,158,284 | 1,107,973 |
| 67 | Kyrgyzstan*                      | 13,730    | 13,868    | 14,088    | 14,303    | 16,147    |
| 68 | Lao People's Democratic Republic | 123,881   | 121,363   | 119,569   | 117,563   | 117,691   |
| 69 | Latvia                           | 18,098    | 19,157    | 20,216    | 21,049    | 23,011    |
| 70 | Lesotho                          | 150       | 149       | 149       | 149       | 148       |
| 71 | Liberia                          | 31,827    | 30,856    | 29,886    | 28,916    | 27,946    |
| 72 | Lithuania*                       | 14,433    | 14,656    | 14,903    | 15,350    | 15,322    |
| 73 | Luxembourg                       | 1,015     | 1,125     | 1,235     | 1,235     | 1,235     |
| 74 | Malawi                           | 23,447    | 22,600    | 22,263    | 21,675    | 21,066    |

#### Natural capital change percentage change with respect to the base year, 1990

Natural capital per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 1,456   | 1,358   | 1,315   | 1,307   | 1,231   | -7   | -10  | -10  | -15  |
| 11,007  | 12,458  | 13,600  | 13,829  | 13,681  | 13   | 24   | 26   | 24   |
| 7,902   | 7,595   | 7,153   | 7,027   | 6,683   | -4   | -9   | -11  | -15  |
| 32,826  | 32,462  | 32,583  | 31,070  | 30,394  | -1   | -1   | -5   | -7   |
| 4,738   | 4,650   | 4,579   | 4,540   | 4,414   | -2   | -3   | -4   | -7   |
| 168,433 | 138,994 | 117,640 | 102,916 | 90,996  | -17  | -30  | -39  | -46  |
| 2,410   | 2,104   | 1,862   | 1,635   | 1,460   | -13  | -23  | -32  | -39  |
| 20,147  | 18,808  | 18,226  | 17,699  | 17,292  | -7   | -10  | -12  | -14  |
| 4,297   | 3,556   | 3,051   | 2,562   | 2,149   | -17  | -29  | -40  | -50  |
| 22,631  | 21,108  | 19,256  | 12,377  | 17,332  | -7   | -15  | -45  | -23  |
| 4,069   | 3,401   | 2,814   | 2,305   | 1,836   | -16  | -31  | -43  | -55  |
| 236,831 | 235,810 | 234,130 | 229,987 | 227,448 | 0    | -1   | -3   | -4   |
| 92      | 77      | 65      | 55      | 46      | -16  | -29  | -41  | -50  |
| 17,713  | 14,136  | 11,095  | 9,291   | 7,686   | -20  | -37  | -48  | -57  |
| 4,826   | 4,574   | 4,354   | 4,280   | 4,023   | -5   | -10  | -11  | -17  |
| 300,729 | 286,163 | 270,754 | 255,629 | 200,320 | -5   | -10  | -15  | -33  |
| 2,843   | 2,536   | 2,284   | 2,057   | 1,860   | -11  | -20  | -28  | -35  |
| 11,481  | 9,905   | 8,784   | 7,941   | 7,135   | -14  | -23  | -31  | -38  |
| 58,359  | 52,393  | 46,619  | 42,041  | 38,211  | -10  | -20  | -28  | -35  |
| 60,576  | 51,536  | 42,793  | 36,361  | 30,450  | -15  | -29  | -40  | -50  |
| 11,271  | 8,601   | 8,111   | 7,175   | 7,009   | -24  | -28  | -36  | -38  |
| 1,485   | 1,234   | 1,073   | 905     | 770     | -17  | -28  | -39  | -48  |
| 5,721   | 5,597   | 5,829   | 5,672   | 5,597   | -2   | 2    | -1   | -2   |
| 12,720  | 12,591  | 11,606  | 10,900  | 10,279  | -1   | -9   | -14  | -19  |
| 3,281   | 3,199   | 3,171   | 3,048   | 3,005   | -2   | -3   | -7   | -8   |
| 2,276   | 1,783   | 1,571   | 1,369   | 1,156   | -22  | -31  | -40  | -49  |
| 53,675  | 54,905  | 57,481  | 55,159  | 50,295  | 2    | 7    | 3    | -6   |
| 1,778   | 1,507   | 1,288   | 1,117   | 977     | -15  | -28  | -37  | -45  |
| 610,061 | 764,585 | 619,961 | 511,606 | 404,853 | 25   | 2    | -16  | -34  |
| 3,091   | 3,020   | 2,843   | 2,837   | 3,027   | -2   | -8   | -8   | -2   |
| 29,549  | 25,310  | 22,488  | 20,434  | 18,980  | -14  | -24  | -31  | -36  |
| 6,794   | 7,687   | 8,476   | 9,130   | 10,218  | 13   | 25   | 34   | 50   |
| 91      | 83      | 76      | 72      | 68      | -9   | -17  | -21  | -25  |
| 14,962  | 14,731  | 10,496  | 9,086   | 6,997   | -2   | -30  | -39  | -53  |
| 3,903   | 4,038   | 4,258   | 4,494   | 4,610   | 3    | 9    | 15   | 18   |
| 2,661   | 2,761   | 2,837   | 2,702   | 2,435   | 4    | 7    | 2    | -9   |
| 2,499   | 2,287   | 1,983   | 1,690   | 1,414   | -9   | -21  | -32  | -43  |

## Natural capital

Natural capital level in millions of constant 2005 US\$

|     | Country/Year        | 1990       | 1995       | 2000       | 2005      | 2010      |
|-----|---------------------|------------|------------|------------|-----------|-----------|
| 75  | Malaysia            | 445,167    | 425,004    | 398,481    | 365,515   | 332,012   |
| 76  | Maldives            | 3          | 3          | 3          | 3         | 3         |
| 77  | Mali                | 78,936     | 78,594     | 78,231     | 76,549    | 73,609    |
| 78  | Malta               | 158        | 134        | 109        | 113       | 125       |
| 79  | Mauritania          | 5,968      | 5,563      | 5,146      | 4,740     | 4,436     |
| 80  | Mauritius           | 1,360      | 1,285      | 1,253      | 1,180     | 1,125     |
| 81  | Mexico              | 996,664    | 973,148    | 924,459    | 868,124   | 824,395   |
| 82  | Mongolia            | 208,966    | 200,736    | 205,465    | 190,548   | 186,644   |
| 83  | Morocco             | 54,203     | 54,621     | 54,506     | 54,096    | 53,922    |
| 84  | Mozambique          | 234,064    | 229,408    | 224,813    | 220,430   | 216,384   |
| 85  | Myanmar             | 288,694    | 272,096    | 255,384    | 241,204   | 226,257   |
| 86  | Namibia             | 63,339     | 61,889     | 60,332     | 58,750    | 57,160    |
| 87  | Nepal               | 106,695    | 97,416     | 88,290     | 82,982    | 82,981    |
| 88  | Netherlands         | 124,082    | 108,675    | 94,511     | 80,987    | 67,059    |
| 89  | New Zealand         | 171,612    | 161,032    | 161,897    | 133,675   | 132,960   |
| 90  | Nicaragua           | 36,505     | 35,339     | 34,198     | 32,364    | 29,802    |
| 91  | Niger               | 12,313     | 10,846     | 9,263      | 9,068     | 8,565     |
| 92  | Nigeria             | 826,123    | 783,729    | 733,306    | 694,573   | 641,889   |
| 93  | Norway              | 436,937    | 391,220    | 325,644    | 259,050   | 202,837   |
| 94  | Pakistan            | 201,983    | 195,509    | 187,584    | 177,785   | 163,516   |
| 95  | Panama              | 39,813     | 38,007     | 36,590     | 36,023    | 35,479    |
| 96  | Papua New Guinea    | 298,355    | 290,466    | 282,447    | 274,907   | 267,573   |
| 97  | Paraguay            | 132,978    | 127,356    | 128,820    | 123,693   | 120,802   |
| 98  | Peru                | 1,177,508  | 1,160,928  | 1,146,764  | 1,135,512 | 1,115,393 |
| 99  | Philippines         | 106,244    | 107,136    | 109,241    | 111,167   | 114,515   |
| 100 | Poland              | 211,569    | 193,474    | 176,624    | 156,382   | 140,177   |
| 101 | Portugal            | 30,430     | 30,887     | 31,120     | 30,669    | 30,002    |
| 102 | Qatar               | 350,742    | 343,211    | 331,602    | 316,163   | 295,984   |
| 103 | Republic of Korea   | 333,151    | 346,782    | 366,825    | 389,546   | 402,847   |
| 104 | Republic of Moldova | 3,583      | 3,787      | 3,991      | 4,296     | 4,522     |
| 105 | Romania             | 185,404    | 177,854    | 171,925    | 162,866   | 159,411   |
| 106 | Russian Federation* | 10,607,209 | 10,413,589 | 10,194,693 | 9,930,864 | 9,671,023 |
| 107 | Rwanda              | 2,333      | 1,929      | 2,063      | 2,220     | 2,314     |
| 108 | Saudi Arabia        | 3,959,322  | 4,039,020  | 3,868,351  | 3,685,802 | 3,492,581 |
| 109 | Senegal             | 62,338     | 60,509     | 58,820     | 56,816    | 55,527    |
| 110 | Serbia              | 21,306     | 21,996     | 22,687     | 25,208    | 30,456    |
| 111 | Sierra Leone        | 16,514     | 15,862     | 15,211     | 14,584    | 13,984    |

Natural capital change percentage change with respect to the base year, 1990

Natural capital per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 24,448  | 20,511  | 17,018  | 14,004  | 11,690  | -16  | -30  | -43  | -52  |
| 15      | 13      | 12      | 11      | 11      | -12  | -20  | -26  | -31  |
| 9,102   | 8,000   | 6,926   | 5,809   | 4,789   | -12  | -24  | -36  | -47  |
| 430     | 346     | 275     | 276     | 301     | -20  | -36  | -36  | -30  |
| 2,991   | 2,427   | 1,947   | 1,556   | 1,282   | -19  | -35  | -48  | -57  |
| 1,284   | 1,130   | 1,047   | 939     | 866     | -12  | -18  | -27  | -33  |
| 11,822  | 10,546  | 9,248   | 8,153   | 7,268   | -11  | -22  | -31  | -39  |
| 95,307  | 87,044  | 85,207  | 74,803  | 67,723  | -9   | -11  | -22  | -29  |
| 2,187   | 2,028   | 1,893   | 1,780   | 1,688   | -7   | -13  | -19  | -23  |
| 17,278  | 14,398  | 12,352  | 10,613  | 9,251   | -17  | -29  | -39  | -46  |
| 7,352   | 6,458   | 5,681   | 5,207   | 4,717   | -12  | -23  | -29  | -36  |
| 44,768  | 37,478  | 31,823  | 28,246  | 25,034  | -16  | -29  | -37  | -44  |
| 5,592   | 4,511   | 3,618   | 3,042   | 2,770   | -19  | -35  | -46  | -50  |
| 8,332   | 7,047   | 5,958   | 4,967   | 4,037   | -15  | -28  | -40  | -52  |
| 50,504  | 43,818  | 41,964  | 32,335  | 30,439  | -13  | -17  | -36  | -40  |
| 8,859   | 7,621   | 6,740   | 5,966   | 5,149   | -14  | -24  | -33  | -42  |
| 1,581   | 1,182   | 848     | 698     | 552     | -25  | -46  | -56  | -65  |
| 8,469   | 7,124   | 5,929   | 4,968   | 4,052   | -16  | -30  | -41  | -52  |
| 103,015 | 89,748  | 72,513  | 56,032  | 41,538  | -13  | -30  | -46  | -60  |
| 1,806   | 1,535   | 1,298   | 1,121   | 942     | -15  | -28  | -38  | -48  |
| 16,479  | 14,198  | 12,378  | 11,124  | 10,088  | -14  | -25  | -32  | -39  |
| 71,760  | 61,597  | 52,511  | 45,100  | 39,015  | -14  | -27  | -37  | -46  |
| 31,334  | 26,558  | 24,108  | 20,973  | 18,716  | -15  | -23  | -33  | -40  |
| 54,299  | 48,723  | 44,342  | 41,203  | 38,361  | -10  | -18  | -24  | -29  |
| 1,724   | 1,547   | 1,413   | 1,299   | 1,228   | -10  | -18  | -25  | -29  |
| 5,559   | 5,039   | 4,611   | 4,098   | 3,662   | -9   | -17  | -26  | -34  |
| 3,066   | 3,051   | 3,011   | 2,909   | 2,810   | 0    | -2   | -5   | -8   |
| 740,395 | 684,545 | 561,128 | 385,101 | 168,288 | -8   | -24  | -48  | -77  |
| 7,751   | 7,765   | 7,977   | 8,280   | 8,361   | 0    | 3    | 7    | 8    |
| 821     | 873     | 972     | 1,140   | 1,266   | 6    | 18   | 39   | 54   |
| 7,989   | 7,842   | 7,747   | 7,481   | 7,419   | -2   | -3   | -6   | -7   |
| 71,346  | 70,032  | 69,466  | 69,040  | 67,649  | -2   | -3   | -3   | -5   |
| 328     | 346     | 255     | 241     | 218     | 6    | -22  | -26  | -34  |
| 245,326 | 218,422 | 192,981 | 153,312 | 127,243 | -11  | -21  | -38  | -48  |
| 8,608   | 7,230   | 6,188   | 5,226   | 4,466   | -16  | -28  | -39  | -48  |
| 2,227   | 2,156   | 2,239   | 2,558   | 3,090   | -3   | 1    | 15   | 39   |
| 4,148   | 4,069   | 3,671   | 2,830   | 2,383   | -2   | -11  | -32  | -43  |

## Natural capital

Natural capital level in millions of constant 2005 US\$

|     | Country/Year                       | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|------------------------------------|-----------|-----------|-----------|-----------|-----------|
| 112 | Singapore                          | 34        | 34        | 34        | 34        | 34        |
| 113 | Slovakia**                         | 15,671    | 15,668    | 15,659    | 15,665    | 15,665    |
| 114 | Slovenia*                          | 19,117    | 19,551    | 20,126    | 21,008    | 22,006    |
| 115 | South Africa                       | 692,121   | 674,510   | 652,895   | 628,906   | 605,021   |
| 116 | Spain                              | 219,100   | 220,826   | 226,998   | 223,231   | 219,010   |
| 117 | Sri Lanka                          | 21,149    | 20,471    | 19,967    | 20,105    | 20,299    |
| 118 | Sudan (former)                     | 400,945   | 385,852   | 370,022   | 363,757   | 355,033   |
| 119 | Swaziland                          | 2,387     | 2,497     | 2,601     | 2,699     | 2,791     |
| 120 | Sweden                             | 208,395   | 206,744   | 205,174   | 214,112   | 216,935   |
| 121 | Switzerland                        | 82,510    | 82,969    | 81,387    | 81,360    | 81,570    |
| 122 | Syrian Arab Republic               | 65,267    | 57,408    | 48,709    | 40,740    | 33,870    |
| 123 | Tajikistan*                        | 6,111     | 6,088     | 6,069     | 6,052     | 6,038     |
| 124 | Thailand                           | 281,145   | 272,607   | 255,917   | 241,878   | 230,696   |
| 125 | Тодо                               | 8,543     | 7,544     | 6,671     | 5,432     | 4,669     |
| 126 | Trinidad and Tobago                | 54,682    | 50,927    | 46,402    | 39,101    | 28,572    |
| 127 | Tunisia                            | 21,816    | 20,182    | 18,283    | 16,470    | 14,619    |
| 128 | Turkey                             | 338,325   | 331,078   | 328,096   | 322,346   | 297,177   |
| 129 | Uganda                             | 19,822    | 17,978    | 16,135    | 14,296    | 12,423    |
| 130 | Ukraine*                           | 686,298   | 678,586   | 671,674   | 663,029   | 654,471   |
| 131 | United Arab Emirates               | 1,054,588 | 1,015,103 | 974,480   | 931,210   | 883,415   |
| 132 | United Kingdom                     | 267,148   | 212,203   | 145,628   | 88,995    | 50,867    |
| 133 | United Republic of Tanzania        | 190,434   | 181,108   | 171,782   | 162,425   | 153,169   |
| 134 | United States of America           | 9,424,518 | 9,170,245 | 8,918,350 | 8,774,429 | 8,639,887 |
| 135 | Uruguay                            | 36,580    | 36,700    | 36,984    | 36,547    | 35,762    |
| 136 | Venezuela (Bolivarian Republic of) | 1,977,186 | 1,911,574 | 1,833,197 | 1,764,493 | 1,702,074 |
| 137 | Vietnam                            | 119,591   | 122,592   | 130,113   | 131,889   | 125,972   |
| 138 | Yemen                              | 70,830    | 67,555    | 62,317    | 56,515    | 52,434    |
| 139 | Zambia                             | 359,294   | 354,215   | 349,621   | 344,216   | 339,266   |
| 140 | Zimbabwe                           | 101,168   | 93,949    | 86,796    | 79,684    | 72,619    |

Natural capital change percentage change with respect to the base year, 1990

Natural capital per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010    | 2005    | 2000    | 1995    | 1990    |
|------|------|------|------|---------|---------|---------|---------|---------|
| -41  | -29  | -23  | -13  | 7       | 8       | 9       | 10      | 11      |
| -3   | -2   | -2   | -1   | 2,868   | 2,893   | 2,897   | 2,918   | 2,949   |
| 10   | 6    | 3    | 1    | 10,842  | 10,493  | 10,137  | 9,944   | 9,878   |
| -36  | -30  | -22  | -13  | 12,068  | 13,159  | 14,586  | 16,292  | 18,811  |
| -16  | -9   | 0    | -1   | 4,753   | 5,144   | 5,634   | 5,601   | 5,634   |
| -20  | -17  | -13  | -8   | 973     | 1,013   | 1,065   | 1,123   | 1,220   |
| -46  | -37  | -28  | -15  | 8,152   | 9,470   | 10,823  | 12,802  | 15,133  |
| -15  | -12  | -12  | -6   | 2,353   | 2,443   | 2,445   | 2,591   | 2,766   |
| -5   | -3   | -5   | -4   | 23,128  | 23,713  | 23,157  | 23,422  | 24,349  |
| -14  | -11  | -8   | -4   | 10,643  | 10,972  | 11,354  | 11,821  | 12,363  |
| -69  | -58  | -42  | -24  | 1,659   | 2,204   | 3,046   | 4,051   | 5,296   |
| -22  | -17  | -13  | -7   | 878     | 938     | 983     | 1,054   | 1,128   |
| -32  | -26  | -18  | -7   | 3,338   | 3,626   | 4,052   | 4,570   | 4,926   |
| -67  | -57  | -40  | -21  | 775     | 1,004   | 1,392   | 1,847   | 2,331   |
| -53  | -34  | -20  | -10  | 21,299  | 29,726  | 35,913  | 40,372  | 44,987  |
| -47  | -37  | -27  | -15  | 1,395   | 1,662   | 1,934   | 2,259   | 2,656   |
| -35  | -24  | -17  | -10  | 4,085   | 4,730   | 5,156   | 5,624   | 6,250   |
| -67  | -55  | -40  | -23  | 372     | 503     | 666     | 863     | 1,120   |
| 8    | 6    | 3    | 0    | 14,400  | 14,130  | 13,738  | 13,274  | 13,283  |
| -80  | -61  | -45  | -26  | 117,605 | 228,835 | 321,240 | 432,228 | 583,083 |
| -82  | -68  | -47  | -22  | 820     | 1,478   | 2,474   | 3,659   | 4,669   |
| -54  | -44  | -32  | -19  | 3,416   | 4,183   | 5,047   | 6,048   | 7,474   |
| -25  | -21  | -15  | -7   | 27,836  | 29,561  | 31,570  | 34,433  | 37,201  |
| -10  | -7   | -5   | -3   | 10,616  | 11,000  | 11,143  | 11,385  | 11,765  |
| -42  | -34  | -25  | -14  | 58,733  | 66,175  | 75,291  | 86,752  | 100,440 |
| -20  | -11  | -7   | -7   | 1,434   | 1,586   | 1,652   | 1,656   | 1,782   |
| -63  | -54  | -41  | -25  | 2,180   | 2,737   | 3,516   | 4,460   | 5,928   |
| -43  | -34  | -25  | -13  | 25,921  | 30,030  | 34,271  | 39,713  | 45,711  |
| -40  | -34  | -28  | -17  | 5,776   | 6.339   | 6,938   | 8,040   | 9,663   |

#### Renewable resources

Renewable resources level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 1  | Afghanistan                      | 5,482     | 5,482     | 5,482     | 5,482     | 5,482     |
| 2  | Albania                          | 14,477    | 14,516    | 14,647    | 14,091    | 15,154    |
| 3  | Algeria                          | 67,763    | 68,913    | 69,150    | 70,481    | 70,210    |
| 4  | Argentina                        | 578,665   | 572,279   | 565,862   | 577,985   | 591,816   |
| 5  | Armenia                          | 2,791     | 2,626     | 2,461     | 2,294     | 2,026     |
| 6  | Australia                        | 1,626,609 | 1,625,957 | 1,616,272 | 1,592,871 | 1,498,281 |
| 7  | Austria                          | 56,846    | 58,270    | 59,865    | 60,895    | 59,703    |
| 8  | Bahrain                          |           |           |           |           |           |
| 9  | Bangladesh                       | 38,625    | 36,067    | 35,843    | 35,286    | 35,542    |
| 10 | Barbados                         | 345       | 345       | 329       | 296       | 263       |
| 11 | Belgium                          | 3,369     | 3,907     | 4,445     | 4,874     | 5,021     |
| 12 | Belize                           | 13,805    | 13,481    | 13,091    | 12,696    | 12,308    |
| 13 | Benin                            | 36,726    | 34,478    | 32,231    | 30,626    | 28,999    |
| 14 | Bolivia (Plurinational State of) | 916,165   | 899,264   | 881,500   | 862,932   | 841,746   |
| 15 | Botswana                         | 66,153    | 63,629    | 61,123    | 58,587    | 56,081    |
| 16 | Brazil                           | 4,662,966 | 4,575,207 | 4,470,724 | 4,379,204 | 4,284,272 |
| 17 | Bulgaria                         | 20,804    | 22,065    | 22,756    | 25,140    | 27,652    |
| 18 | Burundi                          | 5,559     | 5,112     | 4,860     | 4,836     | 4,711     |
| 19 | Cambodia                         | 98,497    | 94,347    | 90,519    | 87,515    | 84,563    |
| 20 | Cameroon                         | 238,983   | 228,730   | 218,490   | 208,296   | 198,622   |
| 21 | Canada                           | 2,634,350 | 2,615,960 | 2,597,020 | 2,578,415 | 2,566,897 |
| 22 | Central African Republic         | 181,088   | 179,914   | 178,739   | 177,574   | 176,399   |
| 23 | Chile                            | 207,026   | 203,539   | 202,738   | 209,006   | 206,852   |
| 24 | China                            | 4,929,045 | 4,994,483 | 4,911,710 | 4,754,842 | 4,751,033 |
| 25 | Colombia                         | 976,405   | 966,507   | 962,638   | 941,167   | 934,492   |
| 26 | Congo                            | 220,844   | 220,073   | 219,345   | 218,612   | 217,952   |
| 27 | Costa Rica                       | 36,996    | 34,050    | 31,613    | 31,697    | 33,128    |
| 28 | Côte d'Ivoire                    | 69,992    | 70,633    | 70,492    | 70,938    | 71,117    |
| 29 | Croatia*                         | 17,727    | 18,491    | 19,446    | 20,405    | 21,376    |
| 30 | Cuba                             | 13,863    | 16,732    | 19,601    | 22,393    | 23,144    |
| 31 | Cyprus                           | 1,324     | 1,313     | 1,333     | 1,402     | 1,304     |
| 32 | Czech Republic**                 | 215       | 226       | 245       | 252       | 272       |
| 33 | Democratic Republic of the Congo | 985,906   | 976,326   | 966,747   | 957,169   | 947,585   |
| 34 | Denmark                          | 6,995     | 6,935     | 6,838     | 7,326     | 7,156     |
| 35 | Dominican Republic               | 29,086    | 29,373    | 28,885    | 28,609    | 28,494    |
| 36 | Ecuador                          | 124,550   | 117,063   | 108,989   | 99,962    | 92,041    |

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Empty cells represent either missing data or unavailability of the resource in the related countries.

Renewable resources change percentage change with respect to the base year, 1990

# Renewable resources per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|--------|--------|------|------|------|------|
| 421     | 277     | 240     | 199    | 175    | -34  | -43  | -53  | -59  |
| 4,401   | 4,621   | 4,768   | 4,485  | 4,729  | 5    | 8    | 2    | 7    |
| 2,678   | 2,436   | 2,265   | 2,143  | 1,980  | -9   | -15  | -20  | -26  |
| 17,727  | 16,419  | 15,322  | 14,942 | 14,644 | -7   | -14  | -16  | -17  |
| 787     | 815     | 800     | 748    | 655    | 3    | 2    | -5   | -17  |
| 95,145  | 89,741  | 84,337  | 78,068 | 67,283 | -6   | -11  | -18  | -29  |
| 7,411   | 7,343   | 7,479   | 7,397  | 7,113  | -1   | 1    | 0    | -4   |
|         |         |         |        |        |      |      |      |      |
| 367     | 307     | 277     | 251    | 239    | -16  | -25  | -32  | -35  |
| 1,329   | 1,309   | 1,228   | 1,094  | 963    | -1   | -8   | -18  | -28  |
| 339     | 388     | 437     | 468    | 469    | 14   | 29   | 38   | 38   |
| 72,565  | 61,192  | 52,255  | 45,189 | 39,496 | -16  | -28  | -38  | -46  |
| 7,694   | 6,101   | 4,945   | 4,012  | 3,277  | -21  | -36  | -48  | -57  |
| 137,594 | 120,368 | 106,112 | 94,344 | 84,769 | -13  | -23  | -31  | -38  |
| 47,855  | 40,128  | 34,770  | 31,235 | 27,944 | -16  | -27  | -35  | -42  |
| 31,159  | 28,269  | 25,631  | 23,546 | 21,977 | -9   | -18  | -24  | -29  |
| 2,359   | 2,640   | 2,842   | 3,248  | 3,690  | 12   | 20   | 38   | 56   |
| 992     | 840     | 762     | 667    | 562    | -15  | -23  | -33  | -43  |
| 10,333  | 8,447   | 7,272   | 6,552  | 5,981  | -18  | -30  | -37  | -42  |
| 19,620  | 16,408  | 13,936  | 11,866 | 10,134 | -16  | -29  | -40  | -48  |
| 95,100  | 89,276  | 84,684  | 79,868 | 75,460 | -6   | -11  | -16  | -21  |
| 61,704  | 54,065  | 48,287  | 44,196 | 40,081 | -12  | -22  | -28  | -35  |
| 15,698  | 14,125  | 13,148  | 12,821 | 12,087 | -10  | -16  | -18  | -23  |
| 4,304   | 4,114   | 3,870   | 3,636  | 3,542  | -4   | -10  | -16  | -18  |
| 29,407  | 26,514  | 24,209  | 21,867 | 20,186 | -10  | -18  | -26  | -31  |
| 92,446  | 80,533  | 69,949  | 61,874 | 53,910 | -13  | -24  | -33  | -42  |
| 12,050  | 9,816   | 8,066   | 7,355  | 7,111  | -19  | -33  | -39  | -41  |
| 5,591   | 4,812   | 4,251   | 3,936  | 3,603  | -14  | -24  | -30  | -36  |
| 3,898   | 3,960   | 4,316   | 4,594  | 4,854  | 2    | 11   | 18   | 25   |
| 1,312   | 1,535   | 1,765   | 1,990  | 2,056  | 17   | 35   | 52   | 57   |
| 1,726   | 1,534   | 1,413   | 1,358  | 1,182  | -11  | -18  | -21  | -32  |
| 21      | 22      | 24      | 25     | 26     | 5    | 15   | 19   | 25   |
| 27,081  | 22,155  | 19,481  | 16,669 | 14,365 | -18  | -28  | -38  | -47  |
| 1,361   | 1,325   | 1,281   | 1,352  | 1,289  | -3   | -6   | -1   | -5   |
| 4,043   | 3,711   | 3,362   | 3,088  | 2,870  | -8   | -17  | -24  | -29  |
| 12,139  | 10,283  | 8,829   | 7,445  | 6,363  | -15  | -27  | -39  | -48  |
|         |         |         |        |        |      |      |      |      |

#### Renewable resources

Renewable resources level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 37 | Egypt                            | 7,978     | 9,891     | 9,915     | 10,614    | 11,060    |
| 38 | El Salvador                      | 7,762     | 7,786     | 7,810     | 7,908     | 7,624     |
| 39 | Estonia                          | 17,255    | 17,949    | 18,643    | 18,612    | 18,349    |
| 40 | Fiji                             | 5,756     | 5,891     | 5,806     | 5,780     | 5,751     |
| 41 | Finland                          | 163,683   | 165,811   | 168,566   | 162,940   | 163,050   |
| 42 | France                           | 266,093   | 267,182   | 269,106   | 276,059   | 276,568   |
| 43 | Gabon                            | 118,073   | 118,073   | 118,073   | 118,073   | 118,073   |
| 44 | Gambia                           | 2,328     | 2,369     | 2,416     | 2,459     | 2,523     |
| 45 | Germany                          | 150,677   | 151,532   | 154,266   | 155,259   | 154,286   |
| 46 | Ghana                            | 60,756    | 57,824    | 56,201    | 53,540    | 50,945    |
| 47 | Greece                           | 159,978   | 160,089   | 151,545   | 84,152    | 147,907   |
| 48 | Guatemala                        | 34,702    | 32,549    | 30,395    | 28,375    | 25,748    |
| 49 | Guyana                           | 169,938   | 169,938   | 169,938   | 169,938   | 169,938   |
| 50 | Haiti                            | 653       | 606       | 559       | 510       | 459       |
| 51 | Honduras                         | 86,606    | 78,811    | 68,993    | 63,915    | 58,420    |
| 52 | Hungary                          | 25,782    | 24,945    | 24,001    | 24,192    | 22,372    |
| 53 | Iceland                          | 76,624    | 76,542    | 76,139    | 75,856    | 64,129    |
| 54 | India                            | 1,006,461 | 1,006,703 | 1,014,782 | 1,007,976 | 1,011,246 |
| 55 | Indonesia                        | 1,622,714 | 1,528,164 | 1,480,161 | 1,461,770 | 1,431,679 |
| 56 | Iran (Islamic Republic of)       | 140,919   | 144,815   | 142,924   | 121,221   | 122,979   |
| 57 | Iraq                             | 5,375     | 5,423     | 5,470     | 5,518     | 5,518     |
| 58 | Ireland                          | 37,957    | 29,618    | 29,738    | 28,998    | 30,741    |
| 59 | Israel                           | 4,334     | 4,289     | 4,238     | 3,886     | 3,902     |
| 60 | Italy                            | 307,596   | 304,528   | 321,218   | 324,797   | 333,699   |
| 61 | Jamaica                          | 25,651    | 26,674    | 25,763    | 25,159    | 24,224    |
| 62 | Japan                            | 386,489   | 385,980   | 388,184   | 376,360   | 372,748   |
| 63 | Jordan                           | 2,459     | 2,609     | 2,518     | 2,404     | 2,382     |
| 64 | Kazakhstan*                      | 12,041    | 11,890    | 11,701    | 12,325    | 12,242    |
| 65 | Kenya                            | 41,677    | 41,343    | 40,243    | 39,787    | 39,565    |
| 66 | Kuwait                           |           |           |           |           |           |
| 67 | Kyrgyzstan*                      | 5,506     | 5,713     | 5,973     | 6,227     | 8,107     |
| 68 | Lao People's Democratic Republic | 123,881   | 121,363   | 119,569   | 117,563   | 117,691   |
| 69 | Latvia                           | 18,098    | 19,157    | 20,216    | 21,049    | 23,011    |
| 70 | Lesotho                          | 150       | 149       | 149       | 149       | 148       |
| 71 | Liberia                          | 31,827    | 30,856    | 29,886    | 28,916    | 27,946    |
| 72 | Lithuania*                       | 14,141    | 14,382    | 14,684    | 15,219    | 15,243    |

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Renewable resources change percentage change with respect to the base year, 1990

# Renewable resources per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 140     | 159     | 147     | 143     | 136     | 14   | 4    | 2    | -3   |
| 1,456   | 1,358   | 1,315   | 1,307   | 1,231   | -7   | -10  | -10  | -15  |
| 11,007  | 12,458  | 13,600  | 13,829  | 13,681  | 13   | 24   | 26   | 24   |
| 7,902   | 7,595   | 7,153   | 7,027   | 6,683   | -4   | -9   | -11  | -15  |
| 32,826  | 32,462  | 32,583  | 31,070  | 30,394  | -1   | -1   | -5   | -7   |
| 4,692   | 4,619   | 4,557   | 4,526   | 4,405   | -2   | -3   | -4   | -6   |
| 127,091 | 108,590 | 95,584  | 86,139  | 78,430  | -15  | -25  | -32  | -38  |
| 2,410   | 2,104   | 1,862   | 1,635   | 1,460   | -13  | -23  | -32  | -39  |
| 1,905   | 1,850   | 1,873   | 1,881   | 1,875   | -3   | -2   | -1   | -2   |
| 4,107   | 3,402   | 2,932   | 2,474   | 2,089   | -17  | -29  | -40  | -49  |
| 15,745  | 15,001  | 13,793  | 7,525   | 13,021  | -5   | -12  | -52  | -17  |
| 3,889   | 3,250   | 2,705   | 2,231   | 1,789   | -16  | -30  | -43  | -54  |
| 234,419 | 233,438 | 231,807 | 227,727 | 225,235 | 0    | -1   | -3   | -4   |
| 92      | 77      | 65      | 55      | 46      | -16  | -29  | -41  | -50  |
| 17,713  | 14,136  | 11,095  | 9,291   | 7,686   | -20  | -37  | -48  | -57  |
| 2,485   | 2,415   | 2,351   | 2,398   | 2,241   | -3   | -5   | -3   | -10  |
| 300,729 | 286,163 | 270,754 | 255,629 | 200,320 | -5   | -10  | -15  | -33  |
| 1,152   | 1,044   | 963     | 884     | 826     | -9   | -16  | -23  | -28  |
| 8,803   | 7,664   | 6,936   | 6,431   | 5,969   | -13  | -21  | -27  | -32  |
| 2,568   | 2,423   | 2,187   | 1,738   | 1,662   | -6   | -15  | -32  | -35  |
| 309     | 267     | 229     | 202     | 174     | -14  | -26  | -35  | -44  |
| 10,749  | 8,202   | 7,818   | 6,974   | 6,877   | -24  | -27  | -35  | -36  |
| 963     | 804     | 704     | 588     | 526     | -16  | -27  | -39  | -45  |
| 5,412   | 5,346   | 5,637   | 5,536   | 5,511   | -1   | 4    | 2    | 2    |
| 10,847  | 10,836  | 9,979   | 9,382   | 8,838   | 0    | -8   | -14  | -19  |
| 3,161   | 3,101   | 3,088   | 2,978   | 2,946   | -2   | -2   | -6   | -7   |
| 720     | 595     | 522     | 450     | 385     | -17  | -28  | -38  | -47  |
| 729     | 747     | 782     | 812     | 764     | 2    | 7    | 11   | 5    |
| 1,778   | 1,507   | 1,288   | 1,117   | 977     | -15  | -28  | -37  | -45  |
|         |         |         |         |         |      |      |      |      |
| 1,240   | 1,244   | 1,205   | 1,235   | 1,520   | 0    | -3   | 0    | 23   |
| 29,549  | 25,310  | 22,488  | 20,434  | 18,980  | -14  | -24  | -31  | -36  |
| 6,794   | 7,687   | 8,476   | 9,130   | 10,218  | 13   | 25   | 34   | 50   |
| 91      | 83      | 76      | 72      | 68      | -9   | -17  | -21  | -25  |
| 14,962  | 14,731  | 10,496  | 9,086   | 6,997   | -2   | -30  | -39  | -53  |
| 3,824   | 3,963   | 4,195   | 4,456   | 4,586   | 4    | 10   | 17   | 20   |
|         |         |         |         |         |      |      |      |      |

#### Renewable resources

Renewable resources level in millions of constant 2005 US\$

|     | Country/Year        | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|---------------------|-----------|-----------|-----------|-----------|-----------|
| 73  | Luxembourg          | 1,015     | 1,125     | 1,235     | 1,235     | 1,235     |
| 74  | Malawi              | 23,447    | 22,600    | 22,263    | 21,675    | 21,066    |
| 75  | Malaysia            | 210,322   | 207,661   | 203,659   | 196,207   | 188,417   |
| 76  | Maldives            | 3         | 3         | 3         | 3         | 3         |
| 77  | Mali                | 78,936    | 78,594    | 78,231    | 76,549    | 73,609    |
| 78  | Malta               | 158       | 134       | 109       | 113       | 125       |
| 79  | Mauritania          | 1,860     | 1,619     | 1,379     | 1,134     | 1,002     |
| 80  | Mauritius           | 1,360     | 1,285     | 1,253     | 1,180     | 1,125     |
| 81  | Mexico              | 779,881   | 784,405   | 767,731   | 746,122   | 736,605   |
| 82  | Mongolia            | 169,414   | 161,621   | 166,730   | 152,248   | 149,355   |
| 83  | Могоссо             | 54,116    | 54,539    | 54,432    | 54,031    | 53,868    |
| 84  | Mozambique          | 230,835   | 226,182   | 221,588   | 217,208   | 213,164   |
| 85  | Myanmar             | 268,976   | 252,897   | 236,818   | 225,097   | 213,313   |
| 86  | Namibia             | 63,339    | 61,889    | 60,332    | 58,750    | 57,160    |
| 87  | Nepal               | 106,680   | 97,401    | 88,276    | 82,968    | 82,968    |
| 88  | Netherlands         | 22,601    | 22,127    | 22,037    | 21,831    | 21,504    |
| 89  | New Zealand         | 152,703   | 144,238   | 147,304   | 120,947   | 122,310   |
| 90  | Nicaragua           | 36,505    | 35,339    | 34,198    | 32,364    | 29,802    |
| 91  | Niger               | 11,717    | 10,257    | 8,681     | 8,494     | 7,998     |
| 92  | Nigeria             | 276,360   | 269,814   | 258,447   | 262,684   | 253,157   |
| 93  | Norway              | 59,888    | 62,408    | 62,241    | 65,283    | 68,046    |
| 94  | Pakistan            | 104,911   | 103,100   | 100,693   | 97,491    | 92,205    |
| 95  | Panama              | 39,813    | 38,007    | 36,590    | 36,023    | 35,479    |
| 96  | Papua New Guinea    | 284,410   | 277,987   | 271,564   | 265,133   | 258,681   |
| 97  | Paraguay            | 132,978   | 127,356   | 128,820   | 123,693   | 120,802   |
| 98  | Peru                | 1,113,028 | 1,099,987 | 1,089,735 | 1,083,220 | 1,069,643 |
| 99  | Philippines         | 99,638    | 100,688   | 102,945   | 105,108   | 109,216   |
| 100 | Poland              | 43,553    | 43,608    | 43,582    | 38,754    | 36,379    |
| 101 | Portugal            | 30,372    | 30,840    | 31,081    | 30,631    | 29,965    |
| 102 | Qatar               | 487       | 519       | 527       | 508       | 527       |
| 103 | Republic of Korea   | 327,088   | 342,069   | 362,716   | 385,885   | 399,542   |
| 104 | Republic of Moldova | 3,583     | 3,787     | 3,991     | 4,296     | 4,522     |
| 105 | Romania             | 151,079   | 151,201   | 151,495   | 148,126   | 150,034   |
| 106 | Russian Federation* | 5,349,101 | 5,344,681 | 5,339,156 | 5,331,594 | 5,358,877 |
| 107 | Rwanda              | 2,333     | 1,929     | 2,063     | 2,220     | 2,314     |
| 108 | Saudi Arabia        | 608,139   | 853,368   | 853,368   | 853,036   | 851,271   |

\* Data for this country are available from 1991, thus all data refer to 1991 instead of 1990.
\*\* Data for this country are available from 1992, thus all data refer to 1992 instead of 1990.
Empty cells represent either missing data or unavailability of the resource in the related countries.
Renewable resources change percentage change with respect to the base year, 1990

# Renewable resources per capita in millions of constant 2005 US\$

|        |        |        |        |        |      |      |      | -    |
|--------|--------|--------|--------|--------|------|------|------|------|
| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
| 2,661  | 2,761  | 2,837  | 2,702  | 2,435  | 4    | 7    | 2    | -9   |
| 2,499  | 2,287  | 1,983  | 1,690  | 1,414  | -9   | -21  | -32  | -43  |
| 11,551 | 10,022 | 8,698  | 7,517  | 6,634  | -13  | -25  | -35  | -43  |
| 15     | 13     | 12     | 11     | 11     | -12  | -20  | -26  | -31  |
| 9,102  | 8,000  | 6,926  | 5,809  | 4,789  | -12  | -24  | -36  | -47  |
| 430    | 346    | 275    | 276    | 301    | -20  | -36  | -36  | -30  |
| 932    | 706    | 522    | 372    | 290    | -24  | -44  | -60  | -69  |
| 1,284  | 1,130  | 1,047  | 939    | 866    | -12  | -18  | -27  | -33  |
| 9,251  | 8,501  | 7,680  | 7,007  | 6,494  | -8   | -17  | -24  | -30  |
| 77,268 | 70,083 | 69,143 | 59,767 | 54,193 | -9   | -11  | -23  | -30  |
| 2,184  | 2,025  | 1,890  | 1,778  | 1,686  | -7   | -13  | -19  | -23  |
| 17,039 | 14,195 | 12,175 | 10,458 | 9,113  | -17  | -29  | -39  | -47  |
| 6,850  | 6,002  | 5,268  | 4,859  | 4,447  | -12  | -23  | -29  | -35  |
| 44,768 | 37,478 | 31,823 | 28,246 | 25,034 | -16  | -29  | -37  | -44  |
| 5,591  | 4,510  | 3,618  | 3,041  | 2,769  | -19  | -35  | -46  | -50  |
| 1,518  | 1,435  | 1,389  | 1,339  | 1,294  | -5   | -8   | -12  | -15  |
| 44,939 | 39,248 | 38,181 | 29,256 | 28,000 | -13  | -15  | -35  | -38  |
| 8,859  | 7,621  | 6,740  | 5,966  | 5,149  | -14  | -24  | -33  | -42  |
| 1,505  | 1,118  | 795    | 654    | 516    | -26  | -47  | -57  | -66  |
| 2,833  | 2,453  | 2,089  | 1,879  | 1,598  | -13  | -26  | -34  | -44  |
| 14,120 | 14,317 | 13,860 | 14,120 | 13,935 | 1    | -2   | 0    | -1   |
| 938    | 810    | 697    | 615    | 531    | -14  | -26  | -34  | -43  |
| 16,479 | 14,198 | 12,378 | 11,124 | 10,088 | -14  | -25  | -32  | -39  |
| 68,406 | 58,950 | 50,488 | 43,497 | 37,718 | -14  | -26  | -36  | -45  |
| 31,334 | 26,558 | 24,108 | 20,973 | 18,716 | -15  | -23  | -33  | -40  |
| 51,326 | 46,165 | 42,137 | 39,306 | 36,787 | -10  | -18  | -23  | -28  |
| 1,617  | 1,454  | 1,332  | 1,229  | 1,171  | -10  | -18  | -24  | -28  |
| 1,144  | 1,136  | 1,138  | 1,015  | 950    | -1   | -1   | -11  | -17  |
| 3,060  | 3,046  | 3,007  | 2,905  | 2,807  | 0    | -2   | -5   | -8   |
| 1,029  | 1,036  | 893    | 619    | 300    | 1    | -13  | -40  | -71  |
| 7,610  | 7,659  | 7,887  | 8,203  | 8,292  | 1    | 4    | 8    | 9    |
| 821    | 873    | 972    | 1,140  | 1,266  | 6    | 18   | 39   | 54   |
| 6,510  | 6,666  | 6,827  | 6,804  | 6,983  | 2    | 5    | 5    | 7    |
| 35,979 | 35,943 | 36,381 | 37,065 | 37,486 | 0    | 1    | 3    | 4    |
| 328    | 346    | 255    | 241    | 218    | 6    | -22  | -26  | -34  |
| 37,681 | 46,148 | 42,572 | 35,482 | 31,014 | 22   | 13   | -6   | -18  |
|        |        |        |        |        |      |      |      |      |

## Renewable resources

Renewable resources level in millions of constant 2005 US\$

|     | Country/Year                       | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|------------------------------------|-----------|-----------|-----------|-----------|-----------|
| 109 | Senegal                            | 61,427    | 59,637    | 57,984    | 56,014    | 54,742    |
| 110 | Serbia                             | 21,306    | 21,996    | 22,687    | 25,208    | 30,456    |
| 111 | Sierra Leone                       | 16,514    | 15,862    | 15,211    | 14,584    | 13,984    |
| 112 | Singapore                          | 34        | 34        | 34        | 34        | 34        |
| 113 | Slovakia**                         | 9,202     | 9,437     | 9,829     | 10,229    | 10,545    |
| 114 | Slovenia*                          | 13,512    | 14,323    | 15,337    | 16,644    | 18,056    |
| 115 | South Africa                       | 157,076   | 159,270   | 159,801   | 158,866   | 158,056   |
| 116 | Spain                              | 197,370   | 202,767   | 211,693   | 210,291   | 207,440   |
| 117 | Sri Lanka                          | 21,149    | 20,471    | 19,967    | 20,105    | 20,299    |
| 118 | Sudan (former)                     | 350,775   | 335,682   | 320,590   | 318,175   | 315,766   |
| 119 | Swaziland                          | 1,496     | 1,609     | 1,723     | 1,834     | 1,941     |
| 120 | Sweden                             | 197,792   | 196,433   | 195,169   | 204,407   | 207,554   |
| 121 | Switzerland                        | 82,510    | 82,969    | 81,387    | 81,360    | 81,570    |
| 122 | Syrian Arab Republic               | 1,084     | 1,085     | 1,087     | 1,083     | 1,084     |
| 123 | Tajikistan*                        | 1,471     | 1,465     | 1,458     | 1,452     | 1,455     |
| 124 | Thailand                           | 204,895   | 200,838   | 191,602   | 187,719   | 190,830   |
| 125 | Тодо                               | 8,129     | 7,182     | 6,357     | 5,145     | 4,400     |
| 126 | Trinidad and Tobago                | 5,444     | 5,352     | 5,190     | 4,979     | 4,834     |
| 127 | Tunisia                            | 6,843     | 7,213     | 7,272     | 7,442     | 7,567     |
| 128 | Turkey                             | 275,350   | 273,730   | 277,104   | 276,687   | 258,570   |
| 129 | Uganda                             | 19,822    | 17,978    | 16,135    | 14,296    | 12,423    |
| 130 | Ukraine*                           | 34,794    | 36,967    | 39,683    | 41,085    | 42,618    |
| 131 | United Arab Emirates               |           |           |           |           |           |
| 132 | United Kingdom                     | 7,804     | 7,879     | 7,953     | 8,477     | 9,138     |
| 133 | United Republic of Tanzania        | 188,852   | 179,528   | 170,204   | 160,850   | 151,597   |
| 134 | United States of America           | 4,591,232 | 4,629,866 | 4,671,883 | 4,811,674 | 4,967,448 |
| 135 | Uruguay                            | 36,580    | 36,700    | 36,984    | 36,547    | 35,762    |
| 136 | Venezuela (Bolivarian Republic of) | 531,200   | 520,136   | 509,220   | 500,088   | 490,410   |
| 137 | Vietnam                            | 83,561    | 88,854    | 100,837   | 109,979   | 113,357   |
| 138 | Yemen                              | 30,602    | 30,732    | 30,653    | 30,480    | 30,546    |
| 139 | Zambia                             | 353,043   | 348,416   | 344,170   | 339,172   | 334,847   |
| 140 | Zimbabwe                           | 92,697    | 85,859    | 79,021    | 72,182    | 65,345    |

Renewable resources change percentage change with respect to the base year, 1990

# Renewable resources per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010   | 2005   | 2000   | 1995   | 1990   |
|------|------|------|------|--------|--------|--------|--------|--------|
| -48  | -39  | -28  | -16  | 4,403  | 5,152  | 6,100  | 7,126  | 8,483  |
| 39   | 15   | 1    | -3   | 3,090  | 2,558  | 2,239  | 2,156  | 2,227  |
| -43  | -32  | -11  | -2   | 2,383  | 2,830  | 3,671  | 4,069  | 4,148  |
| -41  | -29  | -23  | -13  | 7      | 8      | 9      | 10     | 11     |
| 12   | 9    | 5    | 2    | 1,931  | 1,889  | 1,819  | 1,758  | 1,731  |
| 27   | 19   | 11   | 4    | 8,896  | 8,313  | 7,725  | 7,285  | 6,982  |
| -26  | -22  | -16  | -10  | 3,153  | 3,324  | 3,570  | 3,847  | 4,269  |
| -11  | -5   | 4    | 1    | 4,502  | 4,846  | 5,254  | 5,143  | 5,075  |
| -20  | -17  | -13  | -8   | 973    | 1,013  | 1,065  | 1,123  | 1,220  |
| -45  | -37  | -29  | -16  | 7,250  | 8,284  | 9,377  | 11,137 | 13,240 |
| -6   | -4   | -7   | -4   | 1,636  | 1,660  | 1,619  | 1,670  | 1,733  |
| -4   | -2   | -5   | -4   | 22,128 | 22,638 | 22,028 | 22,254 | 23,110 |
| -14  | -11  | -8   | -4   | 10,643 | 10,972 | 11,354 | 11,821 | 12,363 |
| -40  | -33  | -23  | -13  | 53     | 59     | 68     | 77     | 88     |
| -22  | -17  | -13  | -7   | 212    | 225    | 236    | 254    | 271    |
| -23  | -22  | -15  | -6   | 2,761  | 2,814  | 3,034  | 3,367  | 3,590  |
| -67  | -57  | -40  | -21  | 730    | 951    | 1,326  | 1,758  | 2,218  |
| -20  | -15  | -10  | -5   | 3,604  | 3,785  | 4,017  | 4,243  | 4,479  |
| -13  | -10  | -8   | -3   | 722    | 751    | 769    | 807    | 833    |
| -30  | -20  | -14  | -9   | 3,554  | 4,060  | 4,355  | 4,650  | 5,087  |
| -67  | -55  | -40  | -23  | 372    | 503    | 666    | 863    | 1,120  |
| 39   | 30   | 21   | 7    | 938    | 876    | 812    | 723    | 673    |
|      |      |      |      |        |        |        |        |        |
| 8    | 3    | -1   | 0    | 147    | 141    | 135    | 136    | 136    |
| -54  | -44  | -33  | -19  | 3,381  | 4,142  | 5,000  | 5,996  | 7,412  |
| -12  | -11  | -9   | -4   | 16,004 | 16,211 | 16,538 | 17,384 | 18,123 |
| -10  | -7   | -5   | -3   | 10,616 | 11,000 | 11,143 | 11,385 | 11,765 |
| -37  | -30  | -22  | -13  | 16,922 | 18,755 | 20,914 | 23,605 | 26,985 |
| 4    | 6    | 3    | -4   | 1,290  | 1,322  | 1,280  | 1,201  | 1,245  |
| -50  | -42  | -32  | -21  | 1,270  | 1,476  | 1,730  | 2,029  | 2,561  |
| -43  | -34  | -25  | -13  | 25,583 | 29,590 | 33,737 | 39,062 | 44,916 |
| -41  | -35  | -29  | -17  | 5,198  | 5,742  | 6,317  | 7,348  | 8,854  |

## Non-renewable resources

#### Non-renewable resources level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 1  | Afghanistan                      | 2,990     | 2,934     | 2,886     | 2,877     | 2,864     |
| 2  | Albania                          | 16,169    | 15,914    | 15,782    | 15,661    | 15,531    |
| 3  | Algeria                          | 473,137   | 439,028   | 399,611   | 352,582   | 301,470   |
| 4  | Argentina                        | 117,483   | 103,540   | 84,567    | 64,220    | 44,094    |
| 5  | Armenia                          |           |           |           |           |           |
| 6  | Australia                        | 1,379,316 | 1,348,397 | 1,310,979 | 1,267,643 | 1,219,079 |
| 7  | Austria                          | 3,816     | 3,186     | 2,546     | 1,834     | 1,120     |
| 8  | Bahrain                          | 4,849     | 3,931     | 3,025     | 2,140     | 1,283     |
| 9  | Bangladesh                       | 2,790     | 2,590     | 2,317     | 1,903     | 1,313     |
| 10 | Barbados                         | 122       | 96        | 66        | 43        | 21        |
| 11 | Belgium                          |           |           |           |           |           |
| 12 | Belize                           |           |           |           |           |           |
| 13 | Benin                            |           |           |           |           |           |
| 14 | Bolivia (Plurinational State of) | 22,776    | 21,976    | 21,039    | 19,522    | 17,348    |
| 15 | Botswana                         | 1,799     | 1,640     | 1,470     | 1,283     | 1,092     |
| 16 | Brazil                           | 398,477   | 381,764   | 358,411   | 326,595   | 283,542   |
| 17 | Bulgaria                         | 50,114    | 47,798    | 45,338    | 43,028    | 40,564    |
| 18 | Burundi                          |           |           |           |           |           |
| 19 | Cambodia                         |           |           |           |           |           |
| 20 | Cameroon                         | 8,719     | 6,470     | 4,580     | 3,308     | 1,927     |
| 21 | Canada                           | 2,115,952 | 2,045,695 | 1,962,567 | 1,871,684 | 1,778,012 |
| 22 | Central African Republic         |           |           |           |           |           |
| 23 | Chile                            | 64,451    | 61,358    | 56,366    | 50,231    | 43,562    |
| 24 | China                            | 2,863,453 | 2,720,154 | 2,552,612 | 2,344,406 | 2,037,955 |
| 25 | Colombia                         | 193,628   | 182,630   | 166,404   | 151,036   | 132,069   |
| 26 | Congo                            | 30,516    | 27,578    | 23,175    | 18,957    | 14,649    |
| 27 | Costa Rica                       |           |           |           |           |           |
| 28 | Côte d'Ivoire                    | 2,266     | 2,210     | 1,966     | 1,645     | 821       |
| 29 | Croatia*                         | NaN       | 2,608     | 2,026     | 1,539     | 1,103     |
| 30 | Cuba                             | 12,330    | 11,824    | 10,873    | 9,610     | 8,297     |
| 31 | Cyprus                           |           |           |           |           |           |
| 32 | Czech Republic**                 | 31,274    | 27,912    | 23,010    | 18,403    | 14,102    |
| 33 | Democratic Republic of the Congo | 3,294     | 2,839     | 2,390     | 2,021     | 1,667     |
| 34 | Denmark                          | 37,343    | 33,530    | 27,470    | 19,272    | 12,537    |
| 35 | Dominican Republic               | 1,312     | 1,136     | 967       | 828       | 766       |
| 36 | Ecuador                          | 99,332    | 92,789    | 85,456    | 76,825    | 67,252    |

Non-renewable resources change percentage change with respect to the base year, 1990

# Non-renewable resources per capita in millions of constant 2005 US\$

|        |        |        |        |        |      | •    |      | <b>3</b> |
|--------|--------|--------|--------|--------|------|------|------|----------|
| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010     |
| 229    | 148    | 126    | 104    | 91     | -35  | -45  | -55  | -60      |
| 4,915  | 5,066  | 5,138  | 4,985  | 4,847  | 3    | 5    | 1    | -1       |
| 18,702 | 15,518 | 13,087 | 10,721 | 8,500  | -17  | -30  | -43  | -55      |
| 3,599  | 2,971  | 2,290  | 1,660  | 1,091  | -17  | -36  | -54  | -70      |
|        |        |        |        |        |      |      |      |          |
| 80,680 | 74,421 | 68,407 | 62,129 | 54,745 | -8   | -15  | -23  | -32      |
| 498    | 401    | 318    | 223    | 133    | -19  | -36  | -55  | -73      |
| 9,837  | 7,031  | 4,739  | 2,953  | 1,017  | -29  | -52  | -70  | -90      |
| 27     | 22     | 18     | 14     | 9      | -17  | -33  | -49  | -67      |
| 470    | 363    | 248    | 159    | 77     | -23  | -47  | -66  | -84      |
|        |        |        |        |        |      |      |      |          |
|        |        |        |        |        |      |      |      |          |
|        |        |        |        |        |      |      |      |          |
| 3,421  | 2,942  | 2,533  | 2,134  | 1,747  | -14  | -26  | -38  | -49      |
| 1,302  | 1,034  | 836    | 684    | 544    | -21  | -36  | -47  | -58      |
| 2,663  | 2,359  | 2,055  | 1,756  | 1,454  | -11  | -23  | -34  | -45      |
| 5,683  | 5,720  | 5,663  | 5,560  | 5,413  | 1    | 0    | -2   | -5       |
|        |        |        |        |        |      |      |      |          |
|        |        |        |        |        |      |      |      |          |
| 716    | 464    | 292    | 188    | 98     | -35  | -59  | -74  | -86      |
| 76,386 | 69,814 | 63,995 | 57,977 | 52,269 | -9   | -16  | -24  | -32      |
|        |        |        |        |        |      |      |      |          |
| 4,887  | 4,258  | 3,655  | 3,081  | 2,545  | -13  | -25  | -37  | -48      |
| 2,500  | 2,241  | 2,011  | 1,793  | 1,519  | -10  | -20  | -28  | -39      |
| 5,832  | 5,010  | 4,185  | 3,509  | 2,853  | -14  | -28  | -40  | -51      |
| 12,774 | 10,092 | 7,391  | 5,365  | 3,623  | -21  | -42  | -58  | -72      |
|        |        |        |        |        |      |      |      |          |
| 181    | 151    | 119    | 91     | 42     | -17  | -35  | -50  | -77      |
| 698    | 559    | 450    | 346    | 250    | -20  | -36  | -50  | -64      |
| 1,166  | 1,085  | 979    | 854    | 737    | -7   | -16  | -27  | -37      |
|        |        |        |        |        |      |      |      |          |
| 3,032  | 2,705  | 2,246  | 1,801  | 1,344  | -11  | -26  | -41  | -56      |
| 90     | 64     | 48     | 35     | 25     | -29  | -47  | -61  | -72      |
| 7,264  | 6,407  | 5,145  | 3,556  | 2,259  | -12  | -29  | -51  | -69      |
| 182    | 143    | 113    | 89     | 77     | -21  | -38  | -51  | -58      |
| 9,681  | 8,150  | 6,922  | 5,722  | 4,649  | -16  | -28  | -41  | -52      |

## Non-renewable resources

#### Non-renewable resources level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 37 | Egypt                            | 137,268   | 118,994   | 101,185   | 84,687    | 66,641    |
| 38 | El Salvador                      |           |           |           |           |           |
| 39 | Estonia                          |           |           |           |           |           |
| 40 | Fiji                             |           |           |           |           |           |
| 41 | Finland                          |           |           |           |           |           |
| 42 | France                           | 2,589     | 1,820     | 1,300     | 876       | 566       |
| 43 | Gabon                            | 38,409    | 33,059    | 27,244    | 22,997    | 18,919    |
| 44 | Gambia                           |           |           |           |           |           |
| 45 | Germany                          | 1,442,897 | 1,389,371 | 1,346,621 | 1,305,655 | 1,268,885 |
| 46 | Ghana                            | 2,816     | 2,616     | 2,267     | 1,910     | 1,469     |
| 47 | Greece                           | 69,962    | 65,168    | 60,016    | 54,262    | 48,979    |
| 48 | Guatemala                        | 1,610     | 1,510     | 1,226     | 937       | 671       |
| 49 | Guyana                           | 1,748     | 1,727     | 1,703     | 1,686     | 1,670     |
| 50 | Haiti                            |           |           |           |           |           |
| 51 | Honduras                         |           |           |           |           |           |
| 52 | Hungary                          | 24,297    | 22,313    | 20,458    | 18,982    | 17,792    |
| 53 | Iceland                          |           |           |           |           |           |
| 54 | India                            | 1,477,645 | 1,438,996 | 1,392,729 | 1,337,246 | 1,266,207 |
| 55 | Indonesia                        | 493,695   | 446,950   | 394,334   | 343,337   | 279,897   |
| 56 | Iran (Islamic Republic of)       | 3,061,262 | 2,986,051 | 2,903,247 | 2,810,386 | 2,703,592 |
| 57 | Iraq                             | 1,047,065 | 1,040,149 | 1,015,468 | 989,299   | 958,875   |
| 58 | Ireland                          | 1,844     | 1,441     | 1,116     | 838       | 589       |
| 59 | Israel                           | 2,348     | 2,288     | 2,216     | 2,088     | 1,809     |
| 60 | Italy                            | 17,548    | 14,330    | 10,960    | 8,012     | 5,227     |
| 61 | Jamaica                          | 4,430     | 4,318     | 4,201     | 4,070     | 3,950     |
| 62 | Japan                            | 14,566    | 12,300    | 10,430    | 8,937     | 7,538     |
| 63 | Jordan                           | 5,315     | 5,204     | 5,064     | 4,909     | 4,774     |
| 64 | Kazakhstan*                      | 874,329   | 862,523   | 848,029   | 824,536   | 793,811   |
| 65 | Kenya                            |           |           |           |           |           |
| 66 | Kuwait                           | 1,273,616 | 1,244,676 | 1,203,211 | 1,158,284 | 1,107,973 |
| 67 | Kyrgyzstan*                      | 8,224     | 8,155     | 8,115     | 8,076     | 8,040     |
| 68 | Lao People's Democratic Republic |           |           |           |           |           |
| 69 | Latvia                           |           |           |           |           |           |
| 70 | Lesotho                          |           |           |           |           |           |
| 71 | Liberia                          |           |           |           |           |           |
| 72 | Lithuania*                       | 292       | 274       | 219       | 130       | 79        |

Non-renewable resources change percentage change with respect to the base year, 1990

### Non-renewable resources per capita in millions of constant 2005 US\$

|         |         |         |         |         |      | withtitoop |      | je year, 1000 |
|---------|---------|---------|---------|---------|------|------------|------|---------------|
| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000       | 2005 | 2010          |
| 2,415   | 1,917   | 1,496   | 1,141   | 822     | -21  | -38        | -53  | -66           |
|         |         |         |         |         |      |            |      |               |
|         |         |         |         |         |      |            |      |               |
|         |         |         |         |         |      |            |      |               |
|         |         |         |         |         |      |            |      |               |
| 46      | 31      | 22      | 14      | 9       | -31  | -52        | -69  | -80           |
| 41,342  | 30,404  | 22,055  | 16,777  | 12,567  | -26  | -47        | -59  | -70           |
|         |         |         |         |         |      |            |      |               |
| 18,242  | 16,958  | 16,353  | 15,818  | 15,417  | -7   | -10        | -13  | -15           |
| 190     | 154     | 118     | 88      | 60      | -19  | -38        | -54  | -68           |
| 6,886   | 6,107   | 5,463   | 4,852   | 4,312   | -11  | -21        | -30  | -37           |
| 180     | 151     | 109     | 74      | 47      | -16  | -40        | -59  | -74           |
| 2,411   | 2,373   | 2,323   | 2,260   | 2,213   | -2   | -4         | -6   | -8            |
|         |         |         |         |         |      |            |      |               |
|         |         |         |         |         |      |            |      |               |
| 2,342   | 2,160   | 2,004   | 1,882   | 1,782   | -8   | -14        | -20  | -24           |
|         |         |         |         |         |      |            |      |               |
| 1,691   | 1,492   | 1,322   | 1,173   | 1,034   | -12  | -22        | -31  | -39           |
| 2,678   | 2,241   | 1,848   | 1,510   | 1,167   | -16  | -31        | -44  | -56           |
| 55,791  | 49,970  | 44,431  | 40,303  | 36,548  | -10  | -20        | -28  | -34           |
| 60,267  | 51,269  | 42,564  | 36,159  | 30,276  | -15  | -29        | -40  | -50           |
| 522     | 399     | 293     | 201     | 132     | -24  | -44        | -61  | -75           |
| 522     | 429     | 368     | 316     | 244     | -18  | -29        | -39  | -53           |
| 309     | 252     | 192     | 137     | 86      | -19  | -38        | -56  | -72           |
| 1,873   | 1,754   | 1,627   | 1,518   | 1,441   | -6   | -13        | -19  | -23           |
| 119     | 99      | 83      | 71      | 60      | -17  | -30        | -41  | -50           |
| 1,556   | 1,188   | 1,049   | 919     | 772     | -24  | -33        | -41  | -50           |
| 52,946  | 54,159  | 56,699  | 54,347  | 49,532  | 2    | 7          | 3    | -6            |
|         |         |         |         |         |      |            |      |               |
| 610,061 | 764,585 | 619,961 | 511,606 | 404,853 | 25   | 2          | -16  | -34           |
| 1,852   | 1,776   | 1,638   | 1,602   | 1,507   | -4   | -12        | -13  | -19           |
|         |         |         |         |         |      |            |      |               |
|         |         |         |         |         |      |            |      |               |
|         |         |         |         |         |      |            |      |               |
|         |         |         |         |         |      |            |      |               |
| 79      | 75      | 63      | 38      | 24      | -4   | -21        | -52  | -70           |

## Non-renewable resources

#### Non-renewable resources level in millions of constant 2005 US\$

|     | Country/Year        | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|---------------------|-----------|-----------|-----------|-----------|-----------|
| 73  | Luxembourg          |           |           |           |           |           |
| 74  | Malawi              |           |           |           |           |           |
| 75  | Malaysia            | 234,845   | 217,342   | 194,822   | 169,308   | 143,595   |
| 76  | Maldives            |           |           |           |           |           |
| 77  | Mali                |           |           |           |           |           |
| 78  | Malta               |           |           |           |           |           |
| 79  | Mauritania          | 4,109     | 3,943     | 3,767     | 3,606     | 3,433     |
| 80  | Mauritius           |           |           |           |           |           |
| 81  | Mexico              | 216,784   | 188,743   | 156,728   | 122,002   | 87,789    |
| 82  | Mongolia            | 39,552    | 39,115    | 38,735    | 38,300    | 37,289    |
| 83  | Могоссо             | 87        | 82        | 74        | 65        | 55        |
| 84  | Mozambique          | 3,229     | 3,226     | 3,225     | 3,223     | 3,220     |
| 85  | Myanmar             | 19,718    | 19,199    | 18,566    | 16,107    | 12,944    |
| 86  | Namibia             |           |           |           |           |           |
| 87  | Nepal               | 15        | 15        | 14        | 13        | 12        |
| 88  | Netherlands         | 101,482   | 86,547    | 72,474    | 59,156    | 45,555    |
| 89  | New Zealand         | 18,909    | 16,794    | 14,593    | 12,728    | 10,650    |
| 90  | Nicaragua           |           |           |           |           |           |
| 91  | Niger               | 596       | 589       | 582       | 574       | 566       |
| 92  | Nigeria             | 549,763   | 513,915   | 474,860   | 431,889   | 388,732   |
| 93  | Norway              | 377,049   | 328,812   | 263,403   | 193,767   | 134,791   |
| 94  | Pakistan            | 97,073    | 92,410    | 86,890    | 80,294    | 71,312    |
| 95  | Panama              |           |           |           |           |           |
| 96  | Papua New Guinea    | 13,944    | 12,479    | 10,883    | 9,774     | 8,893     |
| 97  | Paraguay            |           |           |           |           |           |
| 98  | Peru                | 64,481    | 60,941    | 57,029    | 52,292    | 45,750    |
| 99  | Philippines         | 6,606     | 6,448     | 6,297     | 6,060     | 5,299     |
| 100 | Poland              | 168,016   | 149,866   | 133,042   | 117,628   | 103,798   |
| 101 | Portugal            | 58        | 47        | 39        | 38        | 38        |
| 102 | Qatar               | 350,254   | 342,692   | 331,075   | 315,654   | 295,457   |
| 103 | Republic of Korea   | 6,064     | 4,713     | 4,109     | 3,661     | 3,305     |
| 104 | Republic of Moldova |           |           |           |           |           |
| 105 | Romania             | 34,325    | 26,653    | 20,430    | 14,740    | 9,377     |
| 106 | Russian Federation* | 5,258,107 | 5,068,908 | 4,855,536 | 4,599,270 | 4,312,146 |
| 107 | Rwanda              |           |           |           |           |           |
| 108 | Saudi Arabia        | 3,351,183 | 3,185,652 | 3,014,983 | 2,832,766 | 2,641,310 |

Non-renewable resources change percentage change with respect to the base year, 1990

### Non-renewable resources per capita in millions of constant 2005 US\$

|         |         |         |         |         |      |      |      | , , , , , , , , , , , , , , , , , , , |
|---------|---------|---------|---------|---------|------|------|------|---------------------------------------|
| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010                                  |
|         |         |         |         |         |      |      |      |                                       |
|         |         |         |         |         |      |      |      |                                       |
| 12,897  | 10,489  | 8,320   | 6,487   | 5,056   | -19  | -35  | -50  | -61                                   |
|         |         |         |         |         |      |      |      |                                       |
|         |         |         |         |         |      |      |      |                                       |
|         |         |         |         |         |      |      |      |                                       |
| 2,059   | 1,720   | 1,425   | 1,183   | 992     | -16  | -31  | -43  | -52                                   |
|         |         |         |         |         |      |      |      |                                       |
| 2,571   | 2,045   | 1,568   | 1,146   | 774     | -20  | -39  | -55  | -70                                   |
| 18,039  | 16,961  | 16,064  | 15,035  | 13,530  | -6   | -11  | -17  | -25                                   |
| 4       | 3       | 3       | 2       | 2       | -13  | -27  | -39  | -51                                   |
| 238     | 202     | 177     | 155     | 138     | -15  | -26  | -35  | -42                                   |
| 502     | 456     | 413     | 348     | 270     | -9   | -18  | -31  | -46                                   |
|         |         |         |         |         |      |      |      |                                       |
| 1       | 1       | 1       |         |         | -13  | -28  | -38  | -48                                   |
| 6,815   | 5,612   | 4,569   | 3,628   | 2,742   | -18  | -33  | -47  | -60                                   |
| 5,565   | 4,570   | 3,782   | 3,079   | 2,438   | -18  | -32  | -45  | -56                                   |
|         |         |         |         |         |      |      |      |                                       |
| 77      | 64      | 53      | 44      | 37      | -16  | -30  | -42  | -52                                   |
| 5,636   | 4,671   | 3,839   | 3,089   | 2,454   | -17  | -32  | -45  | -56                                   |
| 88,895  | 75,431  | 58,653  | 41,911  | 27,603  | -15  | -34  | -53  | -69                                   |
| 868     | 726     | 601     | 506     | 411     | -16  | -31  | -42  | -53                                   |
|         |         |         |         |         |      |      |      |                                       |
| 3,354   | 2,646   | 2,023   | 1,603   | 1,297   | -21  | -40  | -52  | -61                                   |
|         |         |         |         |         |      |      |      |                                       |
| 2,973   | 2,558   | 2,205   | 1,897   | 1,573   | -14  | -26  | -36  | -47                                   |
| 107     | 93      | 81      | 71      | 57      | -13  | -24  | -34  | -47                                   |
| 4,415   | 3,904   | 3,473   | 3,082   | 2,712   | -12  | -21  | -30  | -39                                   |
| 6       | 5       | 4       | 4       | 4       | -21  | -35  | -38  | -39                                   |
| 739,366 | 683,509 | 560,235 | 384,482 | 167,988 | -8   | -24  | -48  | -77                                   |
| 141     | 106     | 89      | 78      | 69      | -25  | -37  | -45  | -51                                   |
|         |         |         |         |         |      |      |      |                                       |
| 1,479   | 1,175   | 921     | 677     | 436     | -21  | -38  | -54  | -70                                   |
| 35,367  | 34,088  | 33,085  | 31,974  | 30,164  | -4   | -6   | -10  | -15                                   |
|         |         |         |         |         |      |      |      |                                       |
| 207,644 | 172,273 | 150,409 | 117,830 | 96,229  | -17  | -28  | -43  | -54                                   |

## Non-renewable resources

#### Non-renewable resources level in millions of constant 2005 US\$

|     | Country/Year                       | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|------------------------------------|-----------|-----------|-----------|-----------|-----------|
| 109 | Senegal                            | 911       | 873       | 837       | 802       | 785       |
| 110 | Serbia                             |           |           |           |           |           |
| 111 | Sierra Leone                       |           |           |           |           |           |
| 112 | Singapore                          |           |           |           |           |           |
| 113 | Slovakia**                         | 6,469     | 6,231     | 5,830     | 5,436     | 5,120     |
| 114 | Slovenia*                          | 5,604     | 5,228     | 4,788     | 4,364     | 3,950     |
| 115 | South Africa                       | 535,045   | 515,241   | 493,094   | 470,040   | 446,965   |
| 116 | Spain                              | 21,730    | 18,058    | 15,305    | 12,940    | 11,570    |
| 117 | Sri Lanka                          |           |           |           |           |           |
| 118 | Sudan (former)                     | 50,170    | 50,170    | 49,432    | 45,582    | 39,266    |
| 119 | Swaziland                          | 891       | 887       | 878       | 864       | 850       |
| 120 | Sweden                             | 10,603    | 10,311    | 10,005    | 9,704     | 9,380     |
| 121 | Switzerland                        |           |           |           |           |           |
| 122 | Syrian Arab Republic               | 64,182    | 56,323    | 47,622    | 39,657    | 32,786    |
| 123 | Tajikistan*                        | 4,640     | 4,623     | 4,611     | 4,599     | 4,582     |
| 124 | Thailand                           | 76,250    | 71,769    | 64,316    | 54,159    | 39,866    |
| 125 | Тодо                               | 414       | 362       | 314       | 287       | 268       |
| 126 | Trinidad and Tobago                | 49,238    | 45,575    | 41,212    | 34,122    | 23,738    |
| 127 | Tunisia                            | 14,973    | 12,969    | 11,011    | 9,027     | 7,052     |
| 128 | Turkey                             | 62,975    | 57,348    | 50,992    | 45,659    | 38,607    |
| 129 | Uganda                             |           |           |           |           |           |
| 130 | Ukraine*                           | 651,504   | 641,619   | 631,991   | 621,944   | 611,853   |
| 131 | United Arab Emirates               | 1,054,588 | 1,015,103 | 974,480   | 931,210   | 883,415   |
| 132 | United Kingdom                     | 259,344   | 204,324   | 137,674   | 80,518    | 41,729    |
| 133 | United Republic of Tanzania        | 1,582     | 1,581     | 1,578     | 1,576     | 1,572     |
| 134 | United States of America           | 4,833,286 | 4,540,380 | 4,246,468 | 3,962,755 | 3,672,439 |
| 135 | Uruguay                            |           |           |           |           |           |
| 136 | Venezuela (Bolivarian Republic of) | 1,445,986 | 1,391,438 | 1,323,977 | 1,264,405 | 1,211,664 |
| 137 | Vietnam                            | 36,030    | 33,738    | 29,275    | 21,910    | 12,615    |
| 138 | Yemen                              | 40,228    | 36,823    | 31,664    | 26,034    | 21,888    |
| 139 | Zambia                             | 6,250     | 5,800     | 5,451     | 5,045     | 4,419     |
| 140 | Zimbabwe                           | 8,471     | 8,090     | 7,776     | 7,501     | 7,274     |

Non-renewable resources change percentage change with respect to the base year, 1990

### Non-renewable resources per capita in millions of constant 2005 US\$

| _ |         |         |         |         |         |      |      |      | <b>,</b> |
|---|---------|---------|---------|---------|---------|------|------|------|----------|
|   | 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010     |
|   | 126     | 104     | 88      | 74      | 63      | -17  | -30  | -41  | -50      |
|   |         |         |         |         |         |      |      |      |          |
|   |         |         |         |         |         |      |      |      |          |
|   |         |         |         |         |         |      |      |      |          |
|   | 1,217   | 1,161   | 1,079   | 1,004   | 937     | -5   | -11  | -18  | -23      |
|   | 2,896   | 2,659   | 2,412   | 2,180   | 1,946   | -8   | -17  | -25  | -33      |
|   | 14,542  | 12,445  | 11,016  | 9,835   | 8,916   | -14  | -24  | -32  | -39      |
|   | 559     | 458     | 380     | 298     | 251     | -18  | -32  | -47  | -55      |
|   |         |         |         |         |         |      |      |      |          |
|   | 1,894   | 1,664   | 1,446   | 1,187   | 902     | -12  | -24  | -37  | -52      |
|   | 1,033   | 921     | 825     | 782     | 717     | -11  | -20  | -24  | -31      |
|   | 1,239   | 1,168   | 1,129   | 1,075   | 1,000   | -6   | -9   | -13  | -19      |
|   |         |         |         |         |         |      |      |      |          |
|   | 5,208   | 3,974   | 2,979   | 2,145   | 1,606   | -24  | -43  | -59  | -69      |
|   | 856     | 800     | 747     | 713     | 666     | -7   | -13  | -17  | -22      |
|   | 1,336   | 1,203   | 1,018   | 812     | 577     | -10  | -24  | -39  | -57      |
|   | 113     | 89      | 65      | 53      | 45      | -22  | -42  | -53  | -61      |
|   | 40,508  | 36,129  | 31,896  | 25,941  | 17,696  | -11  | -21  | -36  | -56      |
|   | 1,823   | 1,451   | 1,164   | 911     | 673     | -20  | -36  | -50  | -63      |
|   | 1,163   | 974     | 801     | 670     | 531     | -16  | -31  | -42  | -54      |
|   |         |         |         |         |         |      |      |      |          |
|   | 12,610  | 12,551  | 12,926  | 13,254  | 13,463  | 0    | 3    | 5    | 7        |
|   | 583,083 | 432,228 | 321,240 | 228,835 | 117,605 | -26  | -45  | -61  | -80      |
|   | 4,533   | 3,523   | 2,338   | 1,337   | 673     | -22  | -48  | -70  | -85      |
|   | 62      | 53      | 46      | 41      | 35      | -15  | -25  | -35  | -44      |
|   | 19,078  | 17,048  | 15,032  | 13,351  | 11,832  | -11  | -21  | -30  | -38      |
|   |         |         |         |         |         |      |      |      |          |
|   | 73,456  | 63,147  | 54,377  | 47,420  | 41,811  | -14  | -26  | -35  | -43      |
|   | 537     | 456     | 372     | 263     | 144     | -15  | -31  | -51  | -73      |
|   | 3,367   | 2,431   | 1,787   | 1,261   | 910     | -28  | -47  | -63  | -73      |
|   | 795     | 650     | 534     | 440     | 338     | -18  | -33  | -45  | -58      |
|   | 809     | 692     | 622     | 597     | 579     | -14  | -23  | -26  | -28      |

## Agricultural land

# Agricultural land level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 1  | Afghanistan                      |           |           |           |           |           |
| 2  | Albania                          | 9,380     | 9,430     | 9,573     | 9,012     | 10,052    |
| 3  | Algeria                          | 58,766    | 60,244    | 60,810    | 62,618    | 62,864    |
| 4  | Argentina                        | 257,809   | 258,779   | 259,719   | 278,489   | 296,515   |
| 5  | Armenia                          |           |           |           |           |           |
| 6  | Australia                        | 624,747   | 623,224   | 612,668   | 598,745   | 536,108   |
| 7  | Austria                          | 23,066    | 22,618    | 22,341    | 21,504    | 20,859    |
| 8  | Bahrain                          |           |           |           |           |           |
| 9  | Bangladesh                       | 23,358    | 21,082    | 21,141    | 20,941    | 20,783    |
| 10 | Barbados                         | 310       | 310       | 294       | 261       | 229       |
| 11 | Belgium                          |           |           |           |           |           |
| 12 | Belize                           | 497       | 576       | 588       | 600       | 619       |
| 13 | Benin                            |           |           |           |           |           |
| 14 | Bolivia (Plurinational State of) | 56,024    | 57,675    | 58,463    | 58,405    | 58,410    |
| 15 | Botswana                         | 5,989     | 5,972     | 5,973     | 5,948     | 5,953     |
| 16 | Brazil                           | 290,077   | 310,324   | 313,847   | 325,724   | 328,272   |
| 17 | Bulgaria                         | 5,973     | 5,978     | 5,414     | 5,106     | 4,900     |
| 18 | Burundi                          | 3,928     | 3,965     | 4,196     | 4,225     | 4,103     |
| 19 | Cambodia                         | 16,900    | 17,336    | 18,094    | 20,317    | 21,452    |
| 20 | Cameroon                         | 12,735    | 12,721    | 12,721    | 12,725    | 13,332    |
| 21 | Canada                           | 61,354    | 61,559    | 61,214    | 61,174    | 57,280    |
| 22 | Central African Republic         |           |           |           |           |           |
| 23 | Chile                            | 122,348   | 117,969   | 116,276   | 122,240   | 121,147   |
| 24 | China                            | 3,698,250 | 3,842,709 | 3,838,959 | 3,829,279 | 3,793,372 |
| 25 | Colombia                         | 296,717   | 292,966   | 295,243   | 280,092   | 279,737   |
| 26 | Congo                            | 21,459    | 21,457    | 21,498    | 21,502    | 21,539    |
| 27 | Costa Rica                       | 23,973    | 21,300    | 19,137    | 18,669    | 19,553    |
| 28 | Côte d'Ivoire                    | 13,811    | 14,446    | 14,300    | 14,738    | 14,957    |
| 29 | Croatia*                         |           |           |           |           |           |
| 30 | Cuba                             |           |           |           |           |           |
| 31 | Cyprus                           | 360       | 322       | 316       | 371       | 253       |
| 32 | Czech Republic**                 |           |           |           |           |           |
| 33 | Democratic Republic of the Congo |           |           |           |           |           |
| 34 | Denmark                          | 5,885     | 5,754     | 5,587     | 5,714     | 5,543     |
| 35 | Dominican Republic               | 14,665    | 14,952    | 14,463    | 14,187    | 14,072    |
| 36 | Ecuador                          | 15,154    | 15,660    | 15,579    | 14,486    | 14,482    |

Agricultural land change percentage change with respect to the base year, 1990

Agricultural land per capita in millions of constant 2005 US\$

| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
|--------|--------|--------|--------|--------|------|------|------|------|
|        |        |        |        |        |      |      |      |      |
| 2,852  | 3,002  | 3,116  | 2,868  | 3,137  | 5    | 9    | 1    | 10   |
| 2,323  | 2,129  | 1,992  | 1,904  | 1,772  | -8   | -14  | -18  | -24  |
| 7,898  | 7,424  | 7,033  | 7,200  | 7,337  | -6   | -11  | -9   | -7   |
|        |        |        |        |        |      |      |      |      |
| 36,543 | 34,397 | 31,969 | 29,345 | 24,075 | -6   | -13  | -20  | -34  |
| 3,007  | 2,850  | 2,791  | 2,612  | 2,485  | -5   | -7   | -13  | -17  |
|        |        |        |        |        |      |      |      |      |
| 222    | 179    | 163    | 149    | 140    | -19  | -26  | -33  | -37  |
| 1,195  | 1,177  | 1,098  | 965    | 836    | -1   | -8   | -19  | -30  |
|        |        |        |        |        |      |      |      |      |
| 2,613  | 2,615  | 2,347  | 2,135  | 1,988  | 0    | -10  | -18  | -24  |
|        |        |        |        |        |      |      |      |      |
| 8,414  | 7,720  | 7,038  | 6,385  | 5,882  | -8   | -16  | -24  | -30  |
| 4,333  | 3,766  | 3,398  | 3,171  | 2,966  | -13  | -22  | -27  | -32  |
| 1,938  | 1,917  | 1,799  | 1,751  | 1,684  | -1   | -7   | -10  | -13  |
| 677    | 715    | 676    | 660    | 654    | 6    | 0    | -3   | -3   |
| 701    | 651    | 658    | 583    | 489    | -7   | -6   | -17  | -30  |
| 1,773  | 1,552  | 1,454  | 1,521  | 1,517  | -12  | -18  | -14  | -14  |
| 1,045  | 913    | 811    | 725    | 680    | -13  | -22  | -31  | -35  |
| 2,215  | 2,101  | 1,996  | 1,895  | 1,684  | -5   | -10  | -14  | -24  |
|        |        |        |        |        |      |      |      |      |
| 9,277  | 8,187  | 7,541  | 7,499  | 7,079  | -12  | -19  | -19  | -24  |
| 3,229  | 3,165  | 3,025  | 2,928  | 2,828  | -2   | -6   | -9   | -12  |
| 8,936  | 8,037  | 7,425  | 6,508  | 6,042  | -10  | -17  | -27  | -32  |
| 8,983  | 7,852  | 6,856  | 6,086  | 5,328  | -13  | -24  | -32  | -41  |
| 7,808  | 6,140  | 4,883  | 4,332  | 4,197  | -21  | -37  | -45  | -46  |
| 1,103  | 984    | 862    | 818    | 758    | -11  | -22  | -26  | -31  |
|        |        |        |        |        |      |      |      |      |
|        |        |        |        |        |      |      |      |      |
| 470    | 377    | 335    | 359    | 230    | -20  | -29  | -23  | -51  |
|        |        |        |        |        |      |      |      |      |
|        |        |        |        |        |      |      |      |      |
| 1,145  | 1,100  | 1,046  | 1,054  | 999    | -4   | -9   | -8   | -13  |
| 2,038  | 1,889  | 1,683  | 1,531  | 1,418  | -7   | -17  | -25  | -30  |
| 1,477  | 1,376  | 1,262  | 1,079  | 1,001  | -7   | -15  | -27  | -32  |

## Agricultural land

# Agricultural land level in millions of constant 2005 US\$

|    | Country/Year                     | 1990    | 1995    | 2000    | 2005    | 2010    |
|----|----------------------------------|---------|---------|---------|---------|---------|
| 37 | Egypt                            | 7,978   | 9,891   | 9,915   | 10,614  | 11,060  |
| 38 | El Salvador                      | 5,448   | 5,622   | 5,795   | 6,046   | 5,909   |
| 39 | Estonia                          |         |         |         |         |         |
| 40 | Fiji                             | 2,156   | 2,314   | 2,251   | 2,251   | 2,249   |
| 41 | Finland                          | 15,496  | 14,604  | 14,339  | 14,701  | 14,811  |
| 42 | France                           | 161,666 | 158,963 | 157,097 | 156,272 | 154,140 |
| 43 | Gabon                            |         |         |         |         |         |
| 44 | Gambia                           | 144     | 137     | 136     | 131     | 152     |
| 45 | Germany                          | 81,803  | 78,677  | 77,430  | 77,262  | 75,760  |
| 46 | Ghana                            | 19,770  | 20,546  | 22,632  | 23,683  | 24,781  |
| 47 | Greece                           | 138,347 | 137,477 | 127,950 | 59,572  | 122,295 |
| 48 | Guatemala                        |         |         |         |         |         |
| 49 | Guyana                           |         |         |         |         |         |
| 50 | Haiti                            |         |         |         |         |         |
| 51 | Honduras                         | 9,530   | 9,989   | 8,425   | 9,042   | 9,243   |
| 52 | Hungary                          | 23,184  | 22,128  | 20,964  | 20,996  | 19,134  |
| 53 | Iceland                          | 76,607  | 76,527  | 76,124  | 75,842  | 64,115  |
| 54 | India                            | 691,106 | 690,114 | 696,958 | 686,594 | 686,216 |
| 55 | Indonesia                        | 328,610 | 307,501 | 332,940 | 358,954 | 397,980 |
| 56 | Iran (Islamic Republic of)       | 88,278  | 92,165  | 90,264  | 68,370  | 69,956  |
| 57 | Iraq                             |         |         |         |         |         |
| 58 | Ireland                          | 37,242  | 28,935  | 29,087  | 28,362  | 30,115  |
| 59 | Israel                           | 4,049   | 4,007   | 3,958   | 3,598   | 3,618   |
| 60 | Italy                            | 183,719 | 167,278 | 170,595 | 160,765 | 156,257 |
| 61 | Jamaica                          | 23,602  | 24,643  | 23,751  | 23,156  | 22,263  |
| 62 | Japan                            | 237,571 | 227,138 | 219,418 | 195,799 | 191,667 |
| 63 | Jordan                           | 2,108   | 2,257   | 2,166   | 2,053   | 2,031   |
| 64 | Kazakhstan*                      |         |         |         |         |         |
| 65 | Kenya                            | 20,599  | 20,943  | 20,522  | 20,777  | 21,122  |
| 66 | Kuwait                           |         |         |         |         |         |
| 67 | Kyrgyzstan*                      |         |         |         |         |         |
| 68 | Lao People's Democratic Republic | 12,513  | 12,815  | 13,840  | 15,144  | 17,926  |
| 69 | Latvia                           |         |         |         |         |         |
| 70 | Lesotho                          |         |         |         |         |         |
| 71 | Liberia                          |         |         |         |         |         |
| 72 | Lithuania*                       |         |         |         |         |         |

Agricultural land change percentage change with respect to the base year, 1990

# Agricultural land per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
| 140     | 159     | 147     | 143     | 136     | 14   | 4    | 2    | -3   |
| 1,022   | 980     | 976     | 999     | 954     | -4   | -4   | -2   | -7   |
|         |         |         |         |         |      |      |      |      |
| 2,961   | 2,984   | 2,773   | 2,737   | 2,613   | 1    | -6   | -8   | -12  |
| 3,108   | 2,859   | 2,772   | 2,803   | 2,761   | -8   | -11  | -10  | -11  |
| 2,851   | 2,748   | 2,660   | 2,562   | 2,455   | -4   | -7   | -10  | -14  |
|         |         |         |         |         |      |      |      |      |
| 150     | 122     | 105     | 87      | 88      | -18  | -30  | -42  | -41  |
| 1,034   | 960     | 940     | 936     | 921     | -7   | -9   | -9   | -11  |
| 1,336   | 1,209   | 1,181   | 1,094   | 1,016   | -10  | -12  | -18  | -24  |
| 13,616  | 12,882  | 11,646  | 5,327   | 10,766  | -5   | -14  | -61  | -21  |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
| 1,949   | 1,792   | 1,355   | 1,314   | 1,216   | -8   | -30  | -33  | -38  |
| 2,234   | 2,142   | 2,053   | 2,082   | 1,917   | -4   | -8   | -7   | -14  |
| 300,665 | 286,104 | 270,700 | 255,580 | 200,273 | -5   | -10  | -15  | -33  |
| 791     | 716     | 661     | 602     | 560     | -10  | -16  | -24  | -29  |
| 1,783   | 1,542   | 1,560   | 1,579   | 1,659   | -13  | -12  | -11  | -7   |
| 1,609   | 1,542   | 1,381   | 980     | 946     | -4   | -14  | -39  | -41  |
|         |         |         |         |         |      |      |      |      |
| 10,547  | 8,013   | 7,647   | 6,821   | 6,737   | -24  | -27  | -35  | -36  |
| 900     | 751     | 658     | 545     | 488     | -16  | -27  | -39  | -46  |
| 3,233   | 2,936   | 2,994   | 2,740   | 2,581   | -9   | -7   | -15  | -20  |
| 9,980   | 10,012  | 9,200   | 8,635   | 8,122   | 0    | -8   | -13  | -19  |
| 1,943   | 1,825   | 1,745   | 1,549   | 1,515   | -6   | -10  | -20  | -22  |
| 617     | 515     | 449     | 384     | 328     | -17  | -27  | -38  | -47  |
|         |         |         |         |         |      |      |      |      |
| 879     | 764     | 657     | 583     | 521     | -13  | -25  | -34  | -41  |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
| 2,985   | 2,672   | 2,603   | 2,632   | 2,891   | -10  | -13  | -12  | -3   |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |

## Agricultural land

# Agricultural land level in millions of constant 2005 US\$

|     | Country/Year        | 1990    | 1995    | 2000    | 2005    | 2010    |
|-----|---------------------|---------|---------|---------|---------|---------|
| 73  | Luxembourg          |         |         |         |         |         |
| 74  | Malawi              | 5,678   | 5,761   | 6,354   | 6,960   | 7,511   |
| 75  | Malaysia            | 14,775  | 16,127  | 16,138  | 16,097  | 16,097  |
| 76  | Maldives            |         |         |         |         |         |
| 77  | Mali                | 15,221  | 16,783  | 18,325  | 19,115  | 19,437  |
| 78  | Malta               | 158     | 134     | 109     | 113     | 125     |
| 79  | Mauritania          |         |         |         |         |         |
| 80  | Mauritius           | 1,196   | 1,120   | 1,088   | 1,034   | 980     |
| 81  | Mexico              | 458,401 | 473,416 | 467,234 | 457,076 | 454,789 |
| 82  | Mongolia            | 84,489  | 79,657  | 87,726  | 76,282  | 76,375  |
| 83  | Morocco             | 31,943  | 32,370  | 32,268  | 31,570  | 31,570  |
| 84  | Mozambique          | 39,884  | 40,051  | 40,277  | 40,779  | 41,323  |
| 85  | Myanmar             |         |         |         |         |         |
| 86  | Namibia             | 25,986  | 26,092  | 26,092  | 26,092  | 26,085  |
| 87  | Nepal               | 9,702   | 9,709   | 9,871   | 9,950   | 9,950   |
| 88  | Netherlands         | 22,601  | 22,127  | 22,037  | 21,831  | 21,504  |
| 89  | New Zealand         | 113,443 | 104,981 | 108,052 | 82,106  | 79,975  |
| 90  | Nicaragua           | 7,975   | 9,021   | 10,093  | 10,501  | 10,196  |
| 91  | Niger               | 1,962   | 2,137   | 2,196   | 2,550   | 2,599   |
| 92  | Nigeria             | 200,126 | 202,226 | 199,504 | 212,416 | 211,583 |
| 93  | Norway              | 11,111  | 12,830  | 11,863  | 11,794  | 11,453  |
| 94  | Pakistan            | 73,494  | 75,222  | 76,356  | 76,667  | 74,797  |
| 95  | Panama              | 9,150   | 9,184   | 9,606   | 9,662   | 9,736   |
| 96  | Papua New Guinea    |         |         |         |         |         |
| 97  | Paraguay            | 26,616  | 25,528  | 31,527  | 30,929  | 32,558  |
| 98  | Peru                | 110,552 | 107,509 | 107,256 | 108,091 | 108,699 |
| 99  | Philippines         | 45,756  | 45,243  | 45,937  | 46,619  | 49,288  |
| 100 | Poland              | 39,712  | 39,350  | 38,909  | 33,611  | 30,860  |
| 101 | Portugal            | 16,345  | 16,184  | 15,797  | 15,759  | 15,166  |
| 102 | Qatar               | 487     | 519     | 527     | 508     | 527     |
| 103 | Republic of Korea   | 220,431 | 207,179 | 199,592 | 190,285 | 179,359 |
| 104 | Republic of Moldova |         |         |         |         |         |
| 105 | Romania             | 79,778  | 79,930  | 80,254  | 76,597  | 76,467  |
| 106 | Russian Federation* |         |         |         |         |         |
| 107 | Rwanda              | 1,744   | 1,378   | 1,550   | 1,690   | 1,782   |
| 108 | Saudi Arabia        | 601,962 | 847,191 | 847,191 | 846,859 | 845,095 |

Agricultural land change percentage change with respect to the base year, 1990

Agricultural land per capita in millions of constant 2005 US\$

| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
|--------|--------|--------|--------|--------|------|------|------|------|
|        |        |        |        |        |      |      |      |      |
| 605    | 583    | 566    | 543    | 504    | -4   | -7   | -10  | -17  |
| 811    | 778    | 689    | 617    | 567    | -4   | -15  | -24  | -30  |
|        |        |        |        |        |      |      |      |      |
| 1,755  | 1,708  | 1,622  | 1,451  | 1,265  | -3   | -8   | -17  | -28  |
| 430    | 346    | 275    | 276    | 301    | -20  | -36  | -36  | -30  |
|        |        |        |        |        |      |      |      |      |
| 1,129  | 985    | 910    | 823    | 755    | -13  | -19  | -27  | -33  |
| 5,437  | 5,131  | 4,674  | 4,292  | 4,010  | -6   | -14  | -21  | -26  |
| 38,535 | 34,541 | 36,380 | 29,946 | 27,712 | -10  | -6   | -22  | -28  |
| 1,289  | 1,202  | 1,121  | 1,039  | 988    | -7   | -13  | -19  | -23  |
| 2,944  | 2,514  | 2,213  | 1,963  | 1,767  | -15  | -25  | -33  | -40  |
|        |        |        |        |        |      |      |      |      |
| 18,367 | 15,800 | 13,763 | 12,545 | 11,424 | -14  | -25  | -32  | -38  |
| 508    | 450    | 405    | 365    | 332    | -12  | -20  | -28  | -35  |
| 1,518  | 1,435  | 1,389  | 1,339  | 1,294  | -5   | -8   | -12  | -15  |
| 33,385 | 28,566 | 28,007 | 19,861 | 18,309 | -14  | -16  | -41  | -45  |
| 1,935  | 1,945  | 1,989  | 1,936  | 1,761  | 1    | 3    | 0    | -9   |
| 252    | 233    | 201    | 196    | 168    | -8   | -20  | -22  | -33  |
| 2,051  | 1,838  | 1,613  | 1,519  | 1,336  | -10  | -21  | -26  | -35  |
| 2,620  | 2,943  | 2,642  | 2,551  | 2,345  | 12   | 1    | -3   | -10  |
| 657    | 591    | 528    | 483    | 431    | -10  | -20  | -26  | -34  |
| 3,787  | 3,431  | 3,250  | 2,984  | 2,768  | -9   | -14  | -21  | -27  |
|        |        |        |        |        |      |      |      |      |
| 6,272  | 5,324  | 5,900  | 5,244  | 5,044  | -15  | -6   | -16  | -20  |
| 5,098  | 4,512  | 4,147  | 3,922  | 3,738  | -11  | -19  | -23  | -27  |
| 742    | 653    | 594    | 545    | 529    | -12  | -20  | -27  | -29  |
| 1,044  | 1,025  | 1,016  | 881    | 806    | -2   | -3   | -16  | -23  |
| 1,647  | 1,598  | 1,528  | 1,495  | 1,421  | -3   | -7   | -9   | -14  |
| 1,029  | 1,036  | 893    | 619    | 300    | 1    | -13  | -40  | -71  |
| 5,129  | 4,639  | 4,340  | 4,045  | 3,722  | -10  | -15  | -21  | -27  |
|        |        |        |        |        |      |      |      |      |
| 3,438  | 3,524  | 3,616  | 3,518  | 3,559  | 3    | 5    | 2    | 4    |
|        |        |        |        |        |      |      |      |      |
| 245    | 247    | 191    | 184    | 168    | 1    | -22  | -25  | -32  |
| 37,298 | 45,814 | 42,264 | 35,225 | 30,789 | 23   | 13   | -6   | -17  |

## Agricultural land

# Agricultural land level in millions of constant 2005 US\$

|     | Country/Year                       | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|------------------------------------|-----------|-----------|-----------|-----------|-----------|
| 109 | Senegal                            | 4,617     | 4,518     | 4,557     | 4,594     | 4,948     |
| 110 | Serbia                             |           |           |           |           |           |
| 111 | Sierra Leone                       |           |           |           |           |           |
| 112 | Singapore                          |           |           |           |           |           |
| 113 | Slovakia**                         |           |           |           |           |           |
| 114 | Slovenia*                          |           |           |           |           |           |
| 115 | South Africa                       | 120,819   | 123,245   | 124,010   | 123,198   | 122,450   |
| 116 | Spain                              | 134,420   | 131,099   | 131,306   | 128,650   | 121,509   |
| 117 | Sri Lanka                          | 11,010    | 10,948    | 11,061    | 11,815    | 12,332    |
| 118 | Sudan (former)                     |           |           |           |           |           |
| 119 | Swaziland                          |           |           |           |           |           |
| 120 | Sweden                             | 10,037    | 9,607     | 9,272     | 9,457     | 9,072     |
| 121 | Switzerland                        | 65,870    | 65,870    | 63,828    | 63,524    | 63,474    |
| 122 | Syrian Arab Republic               |           |           |           |           |           |
| 123 | Tajikistan*                        |           |           |           |           |           |
| 124 | Thailand                           | 92,046    | 91,301    | 85,378    | 84,413    | 90,655    |
| 125 | Тодо                               | 1,758     | 1,818     | 2,000     | 1,791     | 2,039     |
| 126 | Trinidad and Tobago                | 684       | 675       | 595       | 480       | 480       |
| 127 | Tunisia                            | 5,374     | 5,812     | 5,938     | 6,108     | 6,241     |
| 128 | Turkey                             | 169,393   | 168,608   | 172,817   | 175,994   | 166,554   |
| 129 | Uganda                             |           |           |           |           |           |
| 130 | Ukraine*                           |           |           |           |           |           |
| 131 | United Arab Emirates               |           |           |           |           |           |
| 132 | United Kingdom                     |           |           |           |           |           |
| 133 | United Republic of Tanzania        |           |           |           |           |           |
| 134 | United States of America           | 1,351,532 | 1,329,978 | 1,311,807 | 1,303,529 | 1,300,835 |
| 135 | Uruguay                            | 33,453    | 33,521    | 33,754    | 33,262    | 32,409    |
| 136 | Venezuela (Bolivarian Republic of) | 216,967   | 214,585   | 212,352   | 211,905   | 210,913   |
| 137 | Vietnam                            | 33,382    | 35,134    | 43,576    | 49,899    | 53,443    |
| 138 | Yemen                              | 27,845    | 27,974    | 27,895    | 27,723    | 27,789    |
| 139 | Zambia                             | 20,233    | 20,862    | 21,873    | 22,129    | 23,075    |
| 140 | Zimbabwe                           |           |           |           |           |           |

| cultural land<br>percentage<br>o the base ye | with respe |      | per capita<br>2005 US\$ | ricultural land<br>ns of constant | Ag<br>in millior |       |       |
|--|------------|------|-------------------------|-----------------------------------|------------------|-------|-------|
| 2005   | 2000       | 1995 | 2010                    | 2005                              | 2000             | 1995  | 1990  |
| -34  | -25        | -15  | 398                     | 423                               | 479              | 540   | 638   |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |
| -21  | -16        | -9   | 2,443                   | 2,578                             | 2,771            | 2,977 | 3,284 |
| -14  | -6         | -4   | 2,637                   | 2,965                             | 3,259            | 3,325 | 3,457 |
| -6   | -7         | -5   | 591                     | 595                               | 590              | 601   | 635   |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |
| -11  | -11        | -7   | 967                     | 1,047                             | 1,046            | 1,088 | 1,173 |
| -13  | -10        | -5   | 8,282                   | 8,567                             | 8,905            | 9,385 | 9,870 |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |
| -22  | -16        | -5   | 1,312                   | 1,266                             | 1,352            | 1,531 | 1,613 |
| -31  | -13        | -7   | 338                     | 331                               | 417              | 445   | 479   |
| -35  | -18        | -5   | 357                     | 365                               | 460              | 535   | 563   |
| -6   | -4         | -1   | 595                     | 616                               | 628              | 650   | 654   |
| -17  | -13        | -8   | 2,289                   | 2,583                             | 2,716            | 2,864 | 3,129 |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |
|  |            |      |                         |                                   |                  |       |       |

| 5,335  | 4,994  | 4,644  | 4,392  | 4,191 | -6  | -13 | -18 | -21 |
|--------|--------|--------|--------|-------|-----|-----|-----|-----|
| 10,760 | 10,399 | 10,170 | 10,011 | 9,620 | -3  | -5  | -7  | -11 |
| 11,022 | 9,738  | 8,721  | 7,947  | 7,278 | -12 | -21 | -28 | -34 |
| 497    | 475    | 553    | 600    | 608   | -5  | 11  | 21  | 22  |
| 2,330  | 1,847  | 1,574  | 1,343  | 1,155 | -21 | -32 | -42 | -50 |
| 2,574  | 2,339  | 2,144  | 1,931  | 1,763 | -9  | -17 | -25 | -32 |
|        |        |        |        |       |     |     |     |     |

## Forest resources

#### Total forest level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 1  | Afghanistan                      | 5,482     | 5,482     | 5,482     | 5,482     | 5,482     |
| 2  | Albania                          | 5,097     | 5,086     | 5,075     | 5,079     | 5,102     |
| 3  | Algeria                          | 8,997     | 8,669     | 8,341     | 7,863     | 7,346     |
| 4  | Argentina                        | 320,856   | 313,500   | 306,143   | 299,496   | 295,301   |
| 5  | Armenia                          | 2,791     | 2,626     | 2,461     | 2,294     | 2,026     |
| 6  | Australia                        | 1,001,862 | 1,002,733 | 1,003,605 | 994,126   | 962,173   |
| 7  | Austria                          | 33,780    | 35,652    | 37,524    | 39,390    | 38,844    |
| 8  | Bahrain                          |           |           |           |           |           |
| 9  | Bangladesh                       | 15,267    | 14,984    | 14,702    | 14,345    | 14,759    |
| 10 | Barbados                         | 35        | 35        | 35        | 35        | 35        |
| 11 | Belgium                          | 3,369     | 3,907     | 4,445     | 4,874     | 5,021     |
| 12 | Belize                           | 13,308    | 12,905    | 12,503    | 12,096    | 11,688    |
| 13 | Benin                            | 36,726    | 34,478    | 32,231    | 30,626    | 28,999    |
| 14 | Bolivia (Plurinational State of) | 860,141   | 841,589   | 823,037   | 804,528   | 783,337   |
| 15 | Botswana                         | 60,163    | 57,657    | 55,150    | 52,639    | 50,129    |
| 16 | Brazil                           | 4,372,889 | 4,264,883 | 4,156,877 | 4,053,480 | 3,956,000 |
| 17 | Bulgaria                         | 14,831    | 16,086    | 17,342    | 20,034    | 22,752    |
| 18 | Burundi                          | 1,631     | 1,148     | 664       | 611       | 608       |
| 19 | Cambodia                         | 81,597    | 77,011    | 72,424    | 67,197    | 63,112    |
| 20 | Cameroon                         | 226,248   | 216,009   | 205,769   | 195,570   | 185,290   |
| 21 | Canada                           | 2,572,996 | 2,554,401 | 2,535,806 | 2,517,241 | 2,509,616 |
| 22 | Central African Republic         | 181,088   | 179,914   | 178,739   | 177,574   | 176,399   |
| 23 | Chile                            | 84,679    | 85,570    | 86,462    | 86,766    | 85,704    |
| 24 | China                            | 1,230,795 | 1,151,773 | 1,072,752 | 925,563   | 957,661   |
| 25 | Colombia                         | 679,688   | 673,542   | 667,396   | 661,075   | 654,755   |
| 26 | Congo                            | 199,385   | 198,616   | 197,847   | 197,110   | 196,413   |
| 27 | Costa Rica                       | 13,023    | 12,749    | 12,476    | 13,028    | 13,574    |
| 28 | Côte d'Ivoire                    | 56,181    | 56,187    | 56,193    | 56,201    | 56,160    |
| 29 | Croatia*                         | 17,727    | 18,491    | 19,446    | 20,405    | 21,376    |
| 30 | Cuba                             | 13,863    | 16,732    | 19,601    | 22,393    | 23,144    |
| 31 | Cyprus                           | 964       | 990       | 1,017     | 1,031     | 1,051     |
| 32 | Czech Republic**                 | 215       | 226       | 245       | 252       | 272       |
| 33 | Democratic Republic of the Congo | 985,906   | 976,326   | 966,747   | 957,169   | 947,585   |
| 34 | Denmark                          | 1,110     | 1,181     | 1,251     | 1,612     | 1,613     |
| 35 | Dominican Republic               | 14,421    | 14,421    | 14,421    | 14,421    | 14,421    |
| 36 | Ecuador                          | 109,395   | 101,402   | 93,410    | 85,476    | 77,559    |

Total forest change percentage change with respect to the base year, 1990

## Total forest per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
|---------|---------|--------|--------|--------|------|------|------|------|
| 421     | 277     | 240    | 199    | 175    | -34  | -43  | -53  | -59  |
| 1,550   | 1,619   | 1,652  | 1,617  | 1,592  | 4    | 7    | 4    | 3    |
| 356     | 306     | 273    | 239    | 207    | -14  | -23  | -33  | -42  |
| 9,829   | 8,994   | 8,290  | 7,743  | 7,307  | -8   | -16  | -21  | -26  |
| 787     | 815     | 800    | 748    | 655    | 3    | 2    | -5   | -17  |
| 58,602  | 55,343  | 52,368 | 48,723 | 43,208 | -6   | -11  | -17  | -26  |
| 4,404   | 4,493   | 4,688  | 4,785  | 4,628  | 2    | 6    | 9    | 5    |
|         |         |        |        |        |      |      |      |      |
| 145     | 128     | 113    | 102    | 99     | -12  | -22  | -30  | -32  |
| 134     | 132     | 130    | 128    | 127    | -1   | -3   | -4   | -6   |
| 339     | 388     | 437    | 468    | 469    | 14   | 29   | 38   | 38   |
| 69,951  | 58,577  | 49,908 | 43,054 | 37,508 | -16  | -29  | -38  | -46  |
| 7,694   | 6,101   | 4,945  | 4,012  | 3,277  | -21  | -36  | -48  | -57  |
| 129,180 | 112,649 | 99,075 | 87,959 | 78,887 | -13  | -23  | -32  | -39  |
| 43,522  | 36,362  | 31,372 | 28,064 | 24,978 | -16  | -28  | -36  | -43  |
| 29,221  | 26,351  | 23,832 | 21,794 | 20,293 | -10  | -18  | -25  | -31  |
| 1,682   | 1,925   | 2,166  | 2,589  | 3,036  | 14   | 29   | 54   | 81   |
| 291     | 189     | 104    | 84     | 73     | -35  | -64  | -71  | -75  |
| 8,560   | 6,895   | 5,819  | 5,031  | 4,464  | -19  | -32  | -41  | -48  |
| 18,574  | 15,495  | 13,124 | 11,141 | 9,454  | -17  | -29  | -40  | -49  |
| 92,885  | 87,175  | 82,687 | 77,973 | 73,776 | -6   | -11  | -16  | -21  |
| 61,704  | 54,065  | 48,287 | 44,196 | 40,081 | -12  | -22  | -28  | -35  |
| 6,421   | 5,938   | 5,607  | 5,323  | 5,008  | -8   | -13  | -17  | -22  |
| 1,075   | 949     | 845    | 708    | 714    | -12  | -21  | -34  | -34  |
| 20,470  | 18,477  | 16,784 | 15,359 | 14,143 | -10  | -18  | -25  | -31  |
| 83,463  | 72,681  | 63,093 | 55,788 | 48,582 | -13  | -24  | -33  | -42  |
| 4,242   | 3,675   | 3,183  | 3,023  | 2,914  | -13  | -25  | -29  | -31  |
| 4,488   | 3,828   | 3,389  | 3,119  | 2,845  | -15  | -24  | -31  | -37  |
| 3,898   | 3,960   | 4,316  | 4,594  | 4,854  | 2    | 11   | 18   | 25   |
| 1,312   | 1,535   | 1,765  | 1,990  | 2,056  | 17   | 35   | 52   | 57   |
| 1,257   | 1,158   | 1,078  | 998    | 952    | -8   | -14  | -21  | -24  |
| 21      | 22      | 24     | 25     | 26     | 5    | 15   | 19   | 25   |
| 27,081  | 22,155  | 19,481 | 16,669 | 14,365 | -18  | -28  | -38  | -47  |
| 216     | 226     | 234    | 297    | 291    | 4    | 8    | 38   | 35   |
| 2,004   | 1,822   | 1,678  | 1,557  | 1,453  | -9   | -16  | -22  | -28  |
| 10,662  | 8,907   | 7,567  | 6,366  | 5,362  | -16  | -29  | -40  | -50  |

## Forest resources

Total forest level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 37 | Egypt                            |           |           |           |           |           |
| 38 | El Salvador                      | 2,315     | 2,165     | 2,014     | 1,861     | 1,715     |
| 39 | Estonia                          | 17,255    | 17,949    | 18,643    | 18,612    | 18,349    |
| 40 | Fiji                             | 3,599     | 3,577     | 3,555     | 3,529     | 3,502     |
| 41 | Finland                          | 148,187   | 151,207   | 154,227   | 148,239   | 148,239   |
| 42 | France                           | 104,427   | 108,218   | 112,009   | 119,788   | 122,428   |
| 43 | Gabon                            | 118,073   | 118,073   | 118,073   | 118,073   | 118,073   |
| 44 | Gambia                           | 2,184     | 2,232     | 2,279     | 2,328     | 2,371     |
| 45 | Germany                          | 68,874    | 72,855    | 76,836    | 77,997    | 78,526    |
| 46 | Ghana                            | 40,986    | 37,277    | 33,569    | 29,857    | 26,164    |
| 47 | Greece                           | 21,631    | 22,613    | 23,595    | 24,580    | 25,612    |
| 48 | Guatemala                        | 34,702    | 32,549    | 30,395    | 28,375    | 25,748    |
| 49 | Guyana                           | 169,938   | 169,938   | 169,938   | 169,938   | 169,938   |
| 50 | Haiti                            | 653       | 606       | 559       | 510       | 459       |
| 51 | Honduras                         | 77,076    | 68,822    | 60,568    | 54,873    | 49,178    |
| 52 | Hungary                          | 2,597     | 2,817     | 3,037     | 3,196     | 3,238     |
| 53 | Iceland                          | 16        | 16        | 15        | 15        | 15        |
| 54 | India                            | 315,355   | 316,589   | 317,823   | 321,382   | 325,030   |
| 55 | Indonesia                        | 1,294,103 | 1,220,662 | 1,147,221 | 1,102,816 | 1,033,699 |
| 56 | Iran (Islamic Republic of)       | 52,641    | 52,651    | 52,660    | 52,851    | 53,023    |
| 57 | Iraq                             | 5,375     | 5,423     | 5,470     | 5,518     | 5,518     |
| 58 | Ireland                          | 715       | 683       | 651       | 636       | 626       |
| 59 | Israel                           | 285       | 282       | 279       | 288       | 284       |
| 60 | Italy                            | 123,877   | 137,250   | 150,624   | 164,032   | 177,442   |
| 61 | Jamaica                          | 2,049     | 2,030     | 2,012     | 2,003     | 1,961     |
| 62 | Japan                            | 148,918   | 158,842   | 168,765   | 180,561   | 181,080   |
| 63 | Jordan                           | 352       | 352       | 352       | 351       | 351       |
| 64 | Kazakhstan*                      | 12,041    | 11,890    | 11,701    | 12,325    | 12,242    |
| 65 | Kenya                            | 21,079    | 20,400    | 19,720    | 19,010    | 18,443    |
| 66 | Kuwait                           |           |           |           |           |           |
| 67 | Kyrgyzstan*                      | 5,506     | 5,713     | 5,973     | 6,227     | 8,107     |
| 68 | Lao People's Democratic Republic | 111,367   | 108,548   | 105,729   | 102,419   | 99,766    |
| 69 | Latvia                           | 18,098    | 19,157    | 20,216    | 21,049    | 23,011    |
| 70 | Lesotho                          | 150       | 149       | 149       | 149       | 148       |
| 71 | Liberia                          | 31,827    | 30,856    | 29,886    | 28,916    | 27,946    |
| 72 | Lithuania*                       | 14,141    | 14,382    | 14,684    | 15,219    | 15,243    |

Total forest change percentage change with respect to the base year, 1990

## Total forest per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
|         |         |         |         |         |      |      |      |      |
| 434     | 378     | 339     | 308     | 277     | -13  | -22  | -29  | -36  |
| 11,007  | 12,458  | 13,600  | 13,829  | 13,681  | 13   | 24   | 26   | 24   |
| 4,941   | 4,612   | 4,379   | 4,290   | 4,070   | -7   | -11  | -13  | -18  |
| 29,718  | 29,603  | 29,812  | 28,266  | 27,633  | 0    | 0    | -5   | -7   |
| 1,841   | 1,871   | 1,897   | 1,964   | 1,950   | 2    | 3    | 7    | 6    |
| 127,091 | 108,590 | 95,584  | 86,139  | 78,430  | -15  | -25  | -32  | -38  |
| 2,260   | 1,982   | 1,757   | 1,548   | 1,372   | -12  | -22  | -32  | -39  |
| 871     | 889     | 933     | 945     | 954     | 2    | 7    | 9    | 10   |
| 2,771   | 2,193   | 1,752   | 1,380   | 1,073   | -21  | -37  | -50  | -61  |
| 2,129   | 2,119   | 2,148   | 2,198   | 2,255   | 0    | 1    | 3    | 6    |
| 3,889   | 3,250   | 2,705   | 2,231   | 1,789   | -16  | -30  | -43  | -54  |
| 234,419 | 233,438 | 231,807 | 227,727 | 225,235 | 0    | -1   | -3   | -4   |
| 92      | 77      | 65      | 55      | 46      | -16  | -29  | -41  | -50  |
| 15,764  | 12,345  | 9,741   | 7,977   | 6,470   | -22  | -38  | -49  | -59  |
| 250     | 273     | 297     | 317     | 324     | 9    | 19   | 27   | 30   |
| 64      | 59      | 53      | 49      | 46      | -9   | -17  | -23  | -28  |
| 361     | 328     | 302     | 282     | 265     | -9   | -16  | -22  | -26  |
| 7,020   | 6,122   | 5,376   | 4,852   | 4,309   | -13  | -23  | -31  | -39  |
| 959     | 881     | 806     | 758     | 717     | -8   | -16  | -21  | -25  |
| 309     | 267     | 229     | 202     | 174     | -14  | -26  | -35  | -44  |
| 202     | 189     | 171     | 153     | 140     | -7   | -16  | -24  | -31  |
| 63      | 53      | 46      | 44      | 38      | -16  | -27  | -31  | -40  |
| 2,180   | 2,409   | 2,643   | 2,796   | 2,930   | 11   | 21   | 28   | 34   |
| 866     | 825     | 779     | 747     | 715     | -5   | -10  | -14  | -17  |
| 1,218   | 1,276   | 1,342   | 1,429   | 1,431   | 5    | 10   | 17   | 17   |
| 103     | 80      | 73      | 66      | 57      | -22  | -29  | -36  | -45  |
| 729     | 747     | 782     | 812     | 764     | 2    | 7    | 11   | 5    |
| 899     | 744     | 631     | 534     | 455     | -17  | -30  | -41  | -49  |
|         |         |         |         |         |      |      |      |      |
| 1,240   | 1,244   | 1,205   | 1,235   | 1,520   | 0    | -3   | 0    | 23   |
| 26,564  | 22,637  | 19,885  | 17,802  | 16,089  | -15  | -25  | -33  | -39  |
| 6,794   | 7,687   | 8,476   | 9,130   | 10,218  | 13   | 25   | 34   | 50   |
| 91      | 83      | 76      | 72      | 68      | -9   | -17  | -21  | -25  |
| 14,962  | 14,731  | 10,496  | 9,086   | 6,997   | -2   | -30  | -39  | -53  |
| 3,824   | 3,963   | 4,195   | 4,456   | 4,586   | 4    | 10   | 17   | 20   |

## Forest resources

Total forest level in millions of constant 2005 US\$

|     | Country/Year        | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|---------------------|-----------|-----------|-----------|-----------|-----------|
| 73  | Luxembourg          | 1,015     | 1,125     | 1,235     | 1,235     | 1,235     |
| 74  | Malawi              | 17,769    | 16,839    | 15,909    | 14,715    | 13,555    |
| 75  | Malaysia            | 195,547   | 191,534   | 187,522   | 180,111   | 172,320   |
| 76  | Maldives            | 3         | 3         | 3         | 3         | 3         |
| 77  | Mali                | 63,715    | 61,811    | 59,906    | 57,434    | 54,172    |
| 78  | Malta               |           |           |           |           |           |
| 79  | Mauritania          | 1,860     | 1,619     | 1,379     | 1,134     | 1,002     |
| 80  | Mauritius           | 164       | 164       | 165       | 146       | 145       |
| 81  | Mexico              | 321,479   | 310,988   | 300,498   | 289,046   | 281,817   |
| 82  | Mongolia            | 84,924    | 81,964    | 79,003    | 75,966    | 72,980    |
| 83  | Morocco             | 22,173    | 22,169    | 22,164    | 22,461    | 22,298    |
| 84  | Mozambique          | 190,951   | 186,131   | 181,311   | 176,429   | 171,841   |
| 85  | Myanmar             | 268,976   | 252,897   | 236,818   | 225,097   | 213,313   |
| 86  | Namibia             | 37,353    | 35,797    | 34,240    | 32,658    | 31,075    |
| 87  | Nepal               | 96,978    | 87,692    | 78,406    | 73,018    | 73,018    |
| 88  | Netherlands         |           |           |           |           |           |
| 89  | New Zealand         | 39,260    | 39,257    | 39,253    | 38,841    | 42,335    |
| 90  | Nicaragua           | 28,530    | 26,318    | 24,105    | 21,863    | 19,606    |
| 91  | Niger               | 9,756     | 8,120     | 6,485     | 5,943     | 5,400     |
| 92  | Nigeria             | 76,234    | 67,588    | 58,942    | 50,268    | 41,574    |
| 93  | Norway              | 48,776    | 49,578    | 50,379    | 53,489    | 56,593    |
| 94  | Pakistan            | 31,417    | 27,877    | 24,338    | 20,824    | 17,407    |
| 95  | Panama              | 30,663    | 28,823    | 26,984    | 26,360    | 25,743    |
| 96  | Papua New Guinea    | 284,410   | 277,987   | 271,564   | 265,133   | 258,681   |
| 97  | Paraguay            | 106,362   | 101,827   | 97,293    | 92,763    | 88,244    |
| 98  | Peru                | 1,002,476 | 992,478   | 982,480   | 975,129   | 960,944   |
| 99  | Philippines         | 53,882    | 55,445    | 57,008    | 58,489    | 59,928    |
| 100 | Poland              | 3,841     | 4,257     | 4,673     | 5,142     | 5,519     |
| 101 | Portugal            | 14,027    | 14,656    | 15,284    | 14,872    | 14,799    |
| 102 | Qatar               |           |           |           |           |           |
| 103 | Republic of Korea   | 106,657   | 134,891   | 163,125   | 195,600   | 220,182   |
| 104 | Republic of Moldova | 3,583     | 3,787     | 3,991     | 4,296     | 4,522     |
| 105 | Romania             | 71,301    | 71,271    | 71,242    | 71,529    | 73,567    |
| 106 | Russian Federation* | 5,349,101 | 5,344,681 | 5,339,156 | 5,331,594 | 5,358,877 |
| 107 | Rwanda              | 589       | 551       | 513       | 530       | 532       |
| 108 | Saudi Arabia        | 6,177     | 6,177     | 6,177     | 6,177     | 6,177     |

Total forest change percentage change with respect to the base year, 1990

## Total forest per capita in millions of constant 2005 US\$

|       |           |        |        |        |      |      | •    |      |
|-------|-----------|--------|--------|--------|------|------|------|------|
| 199   | 0 1995    | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
| 2,60  | 61 2,761  | 2,837  | 2,702  | 2,435  | 4    | 7    | 2    | -9   |
| 1,89  | 1,704     | 1,417  | 1,148  | 910    | -10  | -25  | -39  | -52  |
| 10,73 | 9,244     | 8,009  | 6,901  | 6,067  | -14  | -25  | -36  | -44  |
|       | 15 13     | 12     | 11     | 11     | -12  | -20  | -26  | -31  |
| 7,34  | 47 6,292  | 5,304  | 4,359  | 3,525  | -14  | -28  | -41  | -52  |
|       |           |        |        |        |      |      |      |      |
| 93    | 32 706    | 522    | 372    | 290    | -24  | -44  | -60  | -69  |
| 15    | 5 145     | 138    | 116    | 112    | -7   | -11  | -25  | -28  |
| 3,8   | 3,370     | 3,006  | 2,714  | 2,485  | -12  | -21  | -29  | -35  |
| 38,73 | 33 35,542 | 32,763 | 29,822 | 26,480 | -8   | -15  | -23  | -32  |
| 89    | 95 823    | 770    | 739    | 698    | -8   | -14  | -17  | -22  |
| 14,09 | 95 11,682 | 9,962  | 8,494  | 7,347  | -17  | -29  | -40  | -48  |
| 6,85  | 6,002     | 5,268  | 4,859  | 4,447  | -12  | -23  | -29  | -35  |
| 26,40 | 01 21,677 | 18,061 | 15,701 | 13,610 | -18  | -32  | -41  | -48  |
| 5,08  | 4,061     | 3,213  | 2,676  | 2,437  | -20  | -37  | -47  | -52  |
|       |           |        |        |        |      |      |      |      |
| 11,55 | 10,682    | 10,174 | 9,395  | 9,692  | -8   | -12  | -19  | -16  |
| 6,92  | .3 5,676  | 4,751  | 4,031  | 3,387  | -18  | -31  | -42  | -51  |
| 1,25  | 63 885    | 594    | 457    | 348    | -29  | -53  | -63  | -72  |
| 78    | 614       | 477    | 360    | 262    | -21  | -39  | -54  | -66  |
| 11,5C | 0 11,373  | 11,218 | 11,569 | 11,590 | -1   | -2   | 1    | 1    |
| 28    | 31 219    | 168    | 131    | 100    | -22  | -40  | -53  | -64  |
| 12,69 | 10,767    | 9,128  | 8,140  | 7,320  | -15  | -28  | -36  | -42  |
| 68,40 | 6 58,950  | 50,488 | 43,497 | 37,718 | -14  | -26  | -36  | -45  |
| 25,06 | 21,235    | 18,208 | 15,728 | 13,672 | -15  | -27  | -37  | -45  |
| 46,22 | 41,653    | 37,989 | 35,384 | 33,049 | -10  | -18  | -23  | -29  |
| 87    | 74 801    | 737    | 684    | 643    | -8   | -16  | -22  | -27  |
| 10    | 01 111    | 122    | 135    | 144    | 10   | 21   | 33   | 43   |
| 1,47  | 1,447     | 1,479  | 1,410  | 1,386  | 2    | 5    | 0    | -2   |
|       |           |        |        |        |      |      |      |      |
| 2,48  | 3,020     | 3,547  | 4,158  | 4,570  | 22   | 43   | 68   | 84   |
| 82    | 21 873    | 972    | 1,140  | 1,266  | 6    | 18   | 39   | 54   |
| 3,07  | 72 3,142  | 3,210  | 3,285  | 3,424  | 2    | 4    | 7    | 11   |
| 35,97 | 9 35,943  | 36,381 | 37,065 | 37,486 | 0    | 1    | 3    | 4    |
| 8     | 3 99      | 63     | 58     | 50     | 19   | -23  | -31  | -40  |
| 38    | 3 334     | 308    | 257    | 225    | -13  | -19  | -33  | -41  |

## Forest resources

Total forest level in millions of constant 2005 US\$

|     | Country/Year                       | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|------------------------------------|-----------|-----------|-----------|-----------|-----------|
| 109 | Senegal                            | 56,810    | 55,118    | 53,426    | 51,420    | 49,794    |
| 110 | Serbia                             | 21,306    | 21,996    | 22,687    | 25,208    | 30,456    |
| 111 | Sierra Leone                       | 16,514    | 15,862    | 15,211    | 14,584    | 13,984    |
| 112 | Singapore                          | 34        | 34        | 34        | 34        | 34        |
| 113 | Slovakia**                         | 9,202     | 9,437     | 9,829     | 10,229    | 10,545    |
| 114 | Slovenia*                          | 13,512    | 14,323    | 15,337    | 16,644    | 18,056    |
| 115 | South Africa                       | 36,258    | 36,024    | 35,791    | 35,667    | 35,605    |
| 116 | Spain                              | 62,950    | 71,669    | 80,387    | 81,641    | 85,932    |
| 117 | Sri Lanka                          | 10,140    | 9,522     | 8,905     | 8,291     | 7,967     |
| 118 | Sudan (former)                     | 350,775   | 335,682   | 320,590   | 318,175   | 315,766   |
| 119 | Swaziland                          | 1,496     | 1,609     | 1,723     | 1,834     | 1,941     |
| 120 | Sweden                             | 187,755   | 186,826   | 185,897   | 194,950   | 198,482   |
| 121 | Switzerland                        | 16,640    | 17,099    | 17,559    | 17,836    | 18,096    |
| 122 | Syrian Arab Republic               | 1,084     | 1,085     | 1,087     | 1,083     | 1,084     |
| 123 | Tajikistan*                        | 1,471     | 1,465     | 1,458     | 1,452     | 1,455     |
| 124 | Thailand                           | 112,850   | 109,537   | 106,224   | 103,305   | 100,174   |
| 125 | Тодо                               | 6,371     | 5,364     | 4,357     | 3,354     | 2,362     |
| 126 | Trinidad and Tobago                | 4,760     | 4,677     | 4,595     | 4,500     | 4,355     |
| 127 | Tunisia                            | 1,469     | 1,402     | 1,334     | 1,335     | 1,326     |
| 128 | Turkey                             | 105,957   | 105,122   | 104,287   | 100,693   | 92,016    |
| 129 | Uganda                             | 19,822    | 17,978    | 16,135    | 14,296    | 12,423    |
| 130 | Ukraine*                           | 34,794    | 36,967    | 39,683    | 41,085    | 42,618    |
| 131 | United Arab Emirates               |           |           |           |           |           |
| 132 | United Kingdom                     | 7,804     | 7,879     | 7,953     | 8,477     | 9,138     |
| 133 | United Republic of Tanzania        | 188,852   | 179,528   | 170,204   | 160,850   | 151,597   |
| 134 | United States of America           | 3,239,701 | 3,299,888 | 3,360,075 | 3,508,145 | 3,666,612 |
| 135 | Uruguay                            | 3,126     | 3,178     | 3,231     | 3,285     | 3,353     |
| 136 | Venezuela (Bolivarian Republic of) | 314,233   | 305,551   | 296,868   | 288,183   | 279,497   |
| 137 | Vietnam                            | 50,180    | 53,720    | 57,261    | 60,079    | 59,914    |
| 138 | Yemen                              | 2,757     | 2,757     | 2,757     | 2,757     | 2,757     |
| 139 | Zambia                             | 332,811   | 327,554   | 322,297   | 317,042   | 311,771   |
| 140 | Zimbabwe                           | 92.697    | 85.859    | 79.021    | 72.182    | 65.345    |

Total forest change percentage change with respect to the base year, 1990

## Total forest per capita in millions of constant 2005 US\$

|   | 1990   |   | 1995   |   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
|---|--------|---|--------|---|--------|--------|--------|------|------|------|------|
|   | 7,845  |   | 6,586  |   | 5,620  | 4,730  | 4,005  | -16  | -28  | -40  | -49  |
|   | 2,227  |   | 2,156  |   | 2,239  | 2,558  | 3,090  | -3   | 1    | 15   | 39   |
|   | 4,148  |   | 4,069  |   | 3,671  | 2,830  | 2,383  | -2   | -11  | -32  | -43  |
|   | 11     |   | 10     |   | 9      | 8      | 7      | -13  | -23  | -29  | -41  |
|   | 1,731  |   | 1,758  |   | 1,819  | 1,889  | 1,931  | 2    | 5    | 9    | 12   |
| ( | 6,982  |   | 7,285  |   | 7,725  | 8,313  | 8,896  | 4    | 11   | 19   | 27   |
|   | 985    |   | 870    |   | 800    | 746    | 710    | -12  | -19  | -24  | -28  |
|   | 1,619  |   | 1,818  |   | 1,995  | 1,881  | 1,865  | 12   | 23   | 16   | 15   |
|   | 585    |   | 522    |   | 475    | 418    | 382    | -11  | -19  | -29  | -35  |
| 1 | 3,240  |   | 11,137 |   | 9,377  | 8,284  | 7,250  | -16  | -29  | -37  | -45  |
|   | 1,733  |   | 1,670  |   | 1,619  | 1,660  | 1,636  | -4   | -7   | -4   | -6   |
| 2 | 21,937 |   | 21,165 | 2 | 20,981 | 21,591 | 21,161 | -4   | -4   | -2   | -4   |
|   | 2,493  |   | 2,436  |   | 2,450  | 2,405  | 2,361  | -2   | -2   | -4   | -5   |
|   | 88     |   | 77     |   | 68     | 59     | 53     | -13  | -23  | -33  | -40  |
|   | 271    |   | 254    |   | 236    | 225    | 212    | -7   | -13  | -17  | -22  |
|   | 1,977  |   | 1,836  |   | 1,682  | 1,549  | 1,449  | -7   | -15  | -22  | -27  |
|   | 1,738  |   | 1,313  |   | 909    | 620    | 392    | -24  | -48  | -64  | -77  |
|   | 3,916  |   | 3,708  |   | 3,556  | 3,421  | 3,246  | -5   | -9   | -13  | -17  |
|   | 179    |   | 157    |   | 141    | 135    | 127    | -12  | -21  | -25  | -29  |
|   | 1,957  |   | 1,786  |   | 1,639  | 1,478  | 1,265  | -9   | -16  | -25  | -35  |
|   | 1,120  |   | 863    |   | 666    | 503    | 372    | -23  | -40  | -55  | -67  |
|   | 673    |   | 723    |   | 812    | 876    | 938    | 7    | 21   | 30   | 39   |
|   |        |   |        |   |        |        |        |      |      |      |      |
|   | 136    |   | 136    |   | 135    | 141    | 147    | 0    | -1   | 3    | 8    |
|   | 7,412  |   | 5,996  |   | 5,000  | 4,142  | 3,381  | -19  | -33  | -44  | -54  |
| 1 | 2,788  |   | 12,391 |   | 11,894 | 11,819 | 11,813 | -3   | -7   | -8   | -8   |
|   | 1,006  |   | 986    |   | 973    | 989    | 995    | -2   | -3   | -2   | -1   |
| 1 | 5,963  |   | 13,867 |   | 12,193 | 10,808 | 9,645  | -13  | -24  | -32  | -40  |
|   | 748    |   | 726    |   | 727    | 722    | 682    | -3   | -3   | -3   | -9   |
|   | 231    |   | 182    |   | 156    | 134    | 115    | -21  | -33  | -42  | -50  |
| 4 | 2,342  | 3 | 36,724 | : | 31,593 | 27,659 | 23,820 | -13  | -25  | -35  | -44  |
|   | 8,854  |   | 7,348  |   | 6,317  | 5,742  | 5,198  | -17  | -29  | -35  | -41  |

## Fossil fuels

#### Fossil fuels level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 1  | Afghanistan                      | 2,990     | 2,934     | 2,886     | 2,877     | 2,864     |
| 2  | Albania                          | 16,169    | 15,914    | 15,782    | 15,661    | 15,531    |
| 3  | Algeria                          | 464,491   | 430,399   | 391,003   | 343,990   | 292,908   |
| 4  | Argentina                        | 117,483   | 103,540   | 84,567    | 64,220    | 44,094    |
| 5  | Armenia                          |           |           |           |           |           |
| 6  | Australia                        | 1,244,704 | 1,217,886 | 1,185,823 | 1,149,089 | 1,108,598 |
| 7  | Austria                          | 3,816     | 3,186     | 2,546     | 1,834     | 1,120     |
| 8  | Bahrain                          | 4,849     | 3,931     | 3,025     | 2,140     | 1,283     |
| 9  | Bangladesh                       | 2,790     | 2,590     | 2,317     | 1,903     | 1,313     |
| 10 | Barbados                         | 122       | 96        | 66        | 43        | 21        |
| 11 | Belgium                          |           |           |           |           |           |
| 12 | Belize                           |           |           |           |           |           |
| 13 | Benin                            |           |           |           |           |           |
| 14 | Bolivia (Plurinational State of) | 20,802    | 20,143    | 19,349    | 17,982    | 16,104    |
| 15 | Botswana                         | 989       | 912       | 835       | 754       | 676       |
| 16 | Brazil                           | 278,607   | 265,103   | 245,249   | 217,711   | 180,818   |
| 17 | Bulgaria                         | 50,114    | 47,798    | 45,338    | 43,028    | 40,564    |
| 18 | Burundi                          |           |           |           |           |           |
| 19 | Cambodia                         |           |           |           |           |           |
| 20 | Cameroon                         | 8,719     | 6,470     | 4,580     | 3,308     | 1,927     |
| 21 | Canada                           | 2,080,365 | 2,013,411 | 1,933,582 | 1,845,581 | 1,754,582 |
| 22 | Central African Republic         |           |           |           |           |           |
| 23 | Chile                            | 6,103     | 5,675     | 5,346     | 5,075     | 4,757     |
| 24 | China                            | 2,723,608 | 2,586,137 | 2,425,505 | 2,225,843 | 1,937,952 |
| 25 | Colombia                         | 191,437   | 180,549   | 164,483   | 149,413   | 130,693   |
| 26 | Congo                            | 30,516    | 27,578    | 23,175    | 18,957    | 14,649    |
| 27 | Costa Rica                       |           |           |           |           |           |
| 28 | Côte d'Ivoire                    | 2,266     | 2,210     | 1,966     | 1,645     | 821       |
| 29 | Croatia*                         | 3,175     | 2,608     | 2,026     | 1,539     | 1,103     |
| 30 | Cuba                             | 7,006     | 6,627     | 5,922     | 4,948     | 3,907     |
| 31 | Cyprus                           |           |           |           |           |           |
| 32 | Czech Republic**                 | 31,274    | 27,912    | 23,010    | 18,403    | 14,102    |
| 33 | Democratic Republic of the Congo | 3,294     | 2,839     | 2,390     | 2,021     | 1,667     |
| 34 | Denmark                          | 37,343    | 33,530    | 27,470    | 19,272    | 12,537    |
| 35 | Dominican Republic               |           |           |           |           |           |
| 36 | Ecuador                          | 99,332    | 92,789    | 85,456    | 76,825    | 67,252    |

## Fossil fuels change percentage change with respect to the base year, 1990

## Fossil fuels per capita in millions of constant 2005 US\$

| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
|--------|--------|--------|--------|--------|------|------|------|------|
| 229    | 148    | 126    | 104    | 91     | -35  | -45  | -55  | -60  |
| 4,915  | 5,066  | 5,138  | 4,985  | 4,847  | 3    | 5    | 1    | -1   |
| 18,360 | 15,213 | 12,806 | 10,459 | 8,258  | -17  | -30  | -43  | -55  |
| 3,599  | 2,971  | 2,290  | 1,660  | 1,091  | -17  | -36  | -54  | -70  |
|        |        |        |        |        |      |      |      |      |
| 72,806 | 67,218 | 61,876 | 56,318 | 49,784 | -8   | -15  | -23  | -32  |
| 498    | 401    | 318    | 223    | 133    | -19  | -36  | -55  | -73  |
| 9,837  | 7,031  | 4,739  | 2,953  | 1,017  | -29  | -52  | -70  | -90  |
| 27     | 22     | 18     | 14     | 9      | -17  | -33  | -49  | -67  |
| 470    | 363    | 248    | 159    | 77     | -23  | -47  | -66  | -84  |
|        |        |        |        |        |      |      |      |      |
|        |        |        |        |        |      |      |      |      |
|        |        |        |        |        |      |      |      |      |
| 3,124  | 2,696  | 2,329  | 1,966  | 1,622  | -14  | -25  | -37  | -48  |
| 716    | 575    | 475    | 402    | 337    | -20  | -34  | -44  | -53  |
| 1,862  | 1,638  | 1,406  | 1,171  | 928    | -12  | -24  | -37  | -50  |
| 5,683  | 5,720  | 5,663  | 5,560  | 5,413  | 1    | 0    | -2   | -5   |
|        |        |        |        |        |      |      |      |      |
|        |        |        |        |        |      |      |      |      |
| 716    | 464    | 292    | 188    | 98     | -35  | -59  | -74  | -86  |
| 75,101 | 68,712 | 63,050 | 57,168 | 51,580 | -9   | -16  | -24  | -31  |
|        |        |        |        |        |      |      |      |      |
| 463    | 394    | 347    | 311    | 278    | -15  | -25  | -33  | -40  |
| 2,378  | 2,130  | 1,911  | 1,702  | 1,445  | -10  | -20  | -28  | -39  |
| 5,766  | 4,953  | 4,136  | 3,471  | 2,823  | -14  | -28  | -40  | -51  |
| 12,774 | 10,092 | 7,391  | 5,365  | 3,623  | -21  | -42  | -58  | -72  |
|        |        |        |        |        |      |      |      |      |
| 181    | 151    | 119    | 91     | 42     | -17  | -35  | -50  | -77  |
| 698    | 559    | 450    | 346    | 250    | -20  | -36  | -50  | -64  |
| 663    | 608    | 533    | 440    | 347    | -8   | -20  | -34  | -48  |
|        |        |        |        |        |      |      |      |      |
| 3,032  | 2,705  | 2,246  | 1,801  | 1,344  | -11  | -26  | -41  | -56  |
| 90     | 64     | 48     | 35     | 25     | -29  | -47  | -61  | -72  |
| 7,264  | 6,407  | 5,145  | 3,556  | 2,259  | -12  | -29  | -51  | -69  |
|        |        |        |        |        |      |      |      |      |
| 9.681  | 8,150  | 6.922  | 5,722  | 4.649  | -16  | -28  | -41  | -52  |

## Fossil fuels

#### Fossil fuels level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 37 | Egypt                            | 136,744   | 118,476   | 100,687   | 84,225    | 66,244    |
| 38 | El Salvador                      |           |           |           |           |           |
| 39 | Estonia                          |           |           |           |           |           |
| 40 | Fiji                             |           |           |           |           |           |
| 41 | Finland                          |           |           |           |           |           |
| 42 | France                           | 2,589     | 1,820     | 1,300     | 876       | 566       |
| 43 | Gabon                            | 38,409    | 33,059    | 27,244    | 22,997    | 18,919    |
| 44 | Gambia                           |           |           |           |           |           |
| 45 | Germany                          | 1,442,897 | 1,389,371 | 1,346,621 | 1,305,655 | 1,268,885 |
| 46 | Ghana                            | 1         | 1         | 1         | 1         |           |
| 47 | Greece                           | 68,678    | 63,906    | 58,773    | 53,042    | 47,780    |
| 48 | Guatemala                        | 1,610     | 1,510     | 1,226     | 937       | 671       |
| 49 | Guyana                           |           |           |           |           |           |
| 50 | Haiti                            |           |           |           |           |           |
| 51 | Honduras                         |           |           |           |           |           |
| 52 | Hungary                          | 24,297    | 22,313    | 20,458    | 18,982    | 17,792    |
| 53 | Iceland                          |           |           |           |           |           |
| 54 | India                            | 1,354,496 | 1,316,992 | 1,272,097 | 1,218,856 | 1,151,561 |
| 55 | Indonesia                        | 471,571   | 425,922   | 375,237   | 326,894   | 266,096   |
| 56 | Iran (Islamic Republic of)       | 3,052,570 | 2,977,450 | 2,894,785 | 2,802,213 | 2,695,922 |
| 57 | Iraq                             | 1,047,065 | 1,040,149 | 1,015,468 | 989,299   | 958,875   |
| 58 | Ireland                          | 1,037     | 735       | 515       | 425       | 370       |
| 59 | Israel                           | 1,353     | 1,347     | 1,341     | 1,286     | 1,067     |
| 60 | Italy                            | 17,548    | 14,330    | 10,960    | 8,012     | 5,227     |
| 61 | Jamaica                          |           |           |           |           |           |
| 62 | Japan                            | 14,258    | 12,035    | 10,206    | 8,751     | 7,389     |
| 63 | Jordan                           | 433       | 393       | 339       | 282       | 235       |
| 64 | Kazakhstan*                      | 843,641   | 832,261   | 818,386   | 795,775   | 766,123   |
| 65 | Kenya                            |           |           |           |           |           |
| 66 | Kuwait                           | 1,273,616 | 1,244,676 | 1,203,211 | 1,158,284 | 1,107,973 |
| 67 | Kyrgyzstan*                      | 8,224     | 8,155     | 8,115     | 8,076     | 8,040     |
| 68 | Lao People's Democratic Republic |           |           |           |           |           |
| 69 | Latvia                           |           |           |           |           |           |
| 70 | Lesotho                          |           |           |           |           |           |
| 71 | Liberia                          |           |           |           |           |           |
| 72 | Lithuania*                       | 292       | 274       | 219       | 130       | 79        |

Fossil fuels change percentage change with respect to the base year, 1990

## Fossil fuels per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010    | 2005    | 2000     | 1995    | 1990    |
|------|------|------|------|---------|---------|----------|---------|---------|
| -66  | -53  | -38  | -21  | 817     | 1,135   | 1,488    | 1,909   | 2,406   |
|      |      |      |      |         |         |          |         |         |
|      |      |      |      |         |         |          |         |         |
|      |      |      |      |         |         |          |         |         |
|      |      |      |      |         |         |          |         |         |
| -80  | -69  | -52  | -31  | 9       | 14      | 22       | 31      | 46      |
| -70  | -59  | -47  | -26  | 12,567  | 16,777  | 22,055   | 30,404  | 41,342  |
|      |      |      |      |         |         |          |         |         |
| -15  | -13  | -10  | -7   | 15,417  | 15,818  | 16,353   | 16,958  | 18,242  |
| -83  | -66  | -43  | -18  |         |         |          |         |         |
| -38  | -30  | -21  | -11  | 4,206   | 4,743   | 5,349    | 5,988   | 6,759   |
| -74  | -59  | -40  | -16  | 47      | 74      | 109      | 151     | 180     |
|      |      |      |      |         |         |          |         |         |
|      |      |      |      |         |         |          |         |         |
|      |      |      |      |         |         |          |         |         |
| -24  | -20  | -14  | -8   | 1,782   | 1,882   | 2,004    | 2,160   | 2,342   |
|      |      |      |      |         |         |          |         |         |
| -39  | -31  | -22  | -12  | 940     | 1,069   | 1,207    | 1,365   | 1,550   |
| -57  | -44  | -31  | -16  | 1,109   | 1,438   | 1,758    | 2,136   | 2,558   |
| -34  | -28  | -20  | -10  | 36,444  | 40,185  | 44,302   | 49,826  | 55,632  |
| -50  | -40  | -29  | -15  | 30,276  | 36,159  | 42,564   | 51,269  | 60,267  |
| -72  | -65  | -54  | -31  | 83      | 102     | 135      | 204     | 294     |
| -52  | -35  | -26  | -16  | 144     | 195     | 223      | 253     | 301     |
| -72  | -56  | -38  | -19  | 86      | 137     | 192      | 252     | 309     |
| 50   | 14   | 22   | 47   | 50      |         | 01       | 07      | 447     |
| -50  | -41  | -30  | -1/  | 58      | 69      | 81       | 97      | 117     |
| -/0  | -58  | -45  | -29  | 38      | 53      | 70       | 90      | 127     |
| -6   | 3    | 1    | 2    | 47,804  | 52,451  | 54,/1/   | 52,259  | 51,088  |
| 24   | 10   | 2    | 05   | 404.052 | F11 000 | C10 0 C1 | 704 505 | C10 0C1 |
| -34  | -16  | 2    | 25   | 404,853 | 511,606 | 619,961  | /64,585 | 610,061 |
| -19  | -13  | -12  | -4   | 1,507   | 1,602   | 1,638    | 1,776   | 1,852   |
|      |      |      |      |         |         |          |         |         |
|      |      |      |      |         |         |          |         |         |
|      |      |      |      |         |         |          |         |         |
| _70  | -52  | _21  | _1   | 24      | 28      | 63       | 75      | 70      |
| 10   | 52   | 21   | 4    | 24      | 50      | 00       | 15      | 13      |

## Fossil fuels

#### Fossil fuels level in millions of constant 2005 US\$

|     | Country/Year        | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|---------------------|-----------|-----------|-----------|-----------|-----------|
| 73  | Luxembourg          |           |           |           |           |           |
| 74  | Malawi              |           |           |           |           |           |
| 75  | Malaysia            | 234,671   | 217,196   | 194,689   | 169,184   | 143,478   |
| 76  | Maldives            |           |           |           |           |           |
| 77  | Mali                |           |           |           |           |           |
| 78  | Malta               |           |           |           |           |           |
| 79  | Mauritania          |           |           |           |           |           |
| 80  | Mauritius           |           |           |           |           |           |
| 81  | Mexico              | 198,001   | 170,880   | 140,005   | 106,447   | 73,634    |
| 82  | Mongolia            | 39,552    | 39,115    | 38,735    | 38,300    | 37,289    |
| 83  | Могоссо             | 87        | 82        | 74        | 65        | 55        |
| 84  | Mozambique          | 3,229     | 3,226     | 3,225     | 3,223     | 3,220     |
| 85  | Myanmar             | 19,718    | 19,199    | 18,566    | 16,107    | 12,944    |
| 86  | Namibia             |           |           |           |           |           |
| 87  | Nepal               | 15        | 15        | 14        | 13        | 12        |
| 88  | Netherlands         | 101,482   | 86,547    | 72,474    | 59,156    | 45,555    |
| 89  | New Zealand         | 18,909    | 16,794    | 14,593    | 12,728    | 10,650    |
| 90  | Nicaragua           |           |           |           |           |           |
| 91  | Niger               | 596       | 589       | 582       | 574       | 566       |
| 92  | Nigeria             | 549,763   | 513,915   | 474,860   | 431,889   | 388,732   |
| 93  | Norway              | 377,049   | 328,812   | 263,403   | 193,767   | 134,791   |
| 94  | Pakistan            | 97,073    | 92,410    | 86,890    | 80,294    | 71,312    |
| 95  | Panama              |           |           |           |           |           |
| 96  | Papua New Guinea    | 11,759    | 10,562    | 9,235     | 8,425     | 7,827     |
| 97  | Paraguay            |           |           |           |           |           |
| 98  | Peru                | 29,477    | 27,061    | 24,922    | 22,976    | 19,773    |
| 99  | Philippines         | 5,018     | 4,912     | 4,832     | 4,686     | 4,388     |
| 100 | Poland              | 159,577   | 141,879   | 125,590   | 110,796   | 97,519    |
| 101 | Portugal            |           |           |           |           |           |
| 102 | Qatar               | 350,254   | 342,692   | 331,075   | 315,654   | 295,457   |
| 103 | Republic of Korea   | 6,064     | 4,713     | 4,109     | 3,661     | 3,305     |
| 104 | Republic of Moldova |           |           |           |           |           |
| 105 | Romania             | 34,325    | 26,653    | 20,430    | 14,740    | 9,377     |
| 106 | Russian Federation* | 5,150,441 | 4,964,039 | 4,754,122 | 4,501,983 | 4,219,197 |
| 107 | Rwanda              |           |           |           |           |           |
| 108 | Saudi Arabia        | 3,351,183 | 3,185,652 | 3,014,983 | 2,832,766 | 2,641,310 |

| Fossil fuels change                |
|------------------------------------|
| percentage change                  |
| with respect to the base year 1990 |

Fossil fuels per capita in millions of constant 2005 US\$

|         |         |         |         |         |      | with tespe |      | e year, 1990 |
|---------|---------|---------|---------|---------|------|------------|------|--------------|
| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000       | 2005 | 2010         |
|         |         |         |         |         |      |            |      |              |
|         |         |         |         |         |      |            |      |              |
| 12,888  | 10,482  | 8,315   | 6,482   | 5,052   | -19  | -35        | -50  | -61          |
|         |         |         |         |         |      |            |      |              |
|         |         |         |         |         |      |            |      |              |
|         |         |         |         |         |      |            |      |              |
|         |         |         |         |         |      |            |      |              |
|         |         |         |         |         |      |            |      |              |
| 2,349   | 1,852   | 1,401   | 1,000   | 649     | -21  | -40        | -57  | -72          |
| 18,039  | 16,961  | 16,064  | 15,035  | 13,530  | -6   | -11        | -17  | -25          |
| 4       | 3       | 3       | 2       | 2       | -13  | -27        | -39  | -51          |
| 238     | 202     | 177     | 155     | 138     | -15  | -26        | -35  | -42          |
| 502     | 456     | 413     | 348     | 270     | -9   | -18        | -31  | -46          |
|         |         |         |         |         |      |            |      |              |
| 0,79    | 0,69    | 0,57    | 0,49    | 0,41    | -13  | -28        | -38  | -48          |
| 6,815   | 5,612   | 4,569   | 3,628   | 2,742   | -18  | -33        | -47  | -60          |
| 5,565   | 4,570   | 3,782   | 3,079   | 2,438   | -18  | -32        | -45  | -56          |
|         |         |         |         |         |      |            |      |              |
| 77      | 64      | 53      | 44      | 37      | -16  | -30        | -42  | -52          |
| 5,636   | 4,671   | 3,839   | 3,089   | 2,454   | -17  | -32        | -45  | -56          |
| 88,895  | 75,431  | 58,653  | 41,911  | 27,603  | -15  | -34        | -53  | -69          |
| 868     | 726     | 601     | 506     | 411     | -16  | -31        | -42  | -53          |
|         |         |         |         |         |      |            |      |              |
| 2,828   | 2,240   | 1,717   | 1,382   | 1,141   | -21  | -39        | -51  | -60          |
|         |         |         |         |         |      |            |      |              |
| 1,359   | 1,136   | 964     | 834     | 680     | -16  | -29        | -39  | -50          |
| 81      | 71      | 63      | 55      | 47      | -13  | -23        | -33  | -42          |
| 4,193   | 3,696   | 3,279   | 2,903   | 2,548   | -12  | -22        | -31  | -39          |
|         |         |         |         |         |      |            |      |              |
| 739,366 | 683,509 | 560,235 | 384,482 | 167,988 | -8   | -24        | -48  | -77          |
| 141     | 106     | 89      | 78      | 69      | -25  | -37        | -45  | -51          |
|         |         |         |         |         |      |            |      |              |
| 1,479   | 1,175   | 921     | 677     | 436     | -21  | -38        | -54  | -70          |
| 34,643  | 33,383  | 32,394  | 31,298  | 29,514  | -4   | -6         | -10  | -15          |
|         |         |         |         |         |      |            |      |              |
| 207,644 | 172,273 | 150,409 | 117,830 | 96,229  | -17  | -28        | -43  | -54          |

## Fossil fuels

#### Fossil fuels level in millions of constant 2005 US\$

|     | Country/Year                       | 1990      | 1995      | 2000      | 2005      | 2010      |
|-----|------------------------------------|-----------|-----------|-----------|-----------|-----------|
| 109 | Senegal                            |           |           |           |           |           |
| 110 | Serbia                             |           |           |           |           |           |
| 111 | Sierra Leone                       |           |           |           |           |           |
| 112 | Singapore                          |           |           |           |           |           |
| 113 | Slovakia**                         | 6,469     | 6,231     | 5,830     | 5,436     | 5,120     |
| 114 | Slovenia*                          | 5,604     | 5,228     | 4,788     | 4,364     | 3,950     |
| 115 | South Africa                       | 504,431   | 488,230   | 469,134   | 448,662   | 427,642   |
| 116 | Spain                              | 21,730    | 18,058    | 15,305    | 12,940    | 11,570    |
| 117 | Sri Lanka                          |           |           |           |           |           |
| 118 | Sudan (former)                     | 50,170    | 50,170    | 49,432    | 45,582    | 39,266    |
| 119 | Swaziland                          | 891       | 887       | 878       | 864       | 850       |
| 120 | Sweden                             |           |           |           |           |           |
| 121 | Switzerland                        |           |           |           |           |           |
| 122 | Syrian Arab Republic               | 58,592    | 50,752    | 42,085    | 34,160    | 27,340    |
| 123 | Tajikistan*                        | 4,640     | 4,623     | 4,611     | 4,599     | 4,582     |
| 124 | Thailand                           | 76,131    | 71,666    | 64,217    | 54,064    | 39,771    |
| 125 | Тодо                               |           |           |           |           |           |
| 126 | Trinidad and Tobago                | 49,238    | 45,575    | 41,212    | 34,122    | 23,738    |
| 127 | Tunisia                            | 14,030    | 12,140    | 10,330    | 8,497     | 6,671     |
| 128 | Turkey                             | 62,975    | 57,348    | 50,992    | 45,659    | 38,607    |
| 129 | Uganda                             |           |           |           |           |           |
| 130 | Ukraine*                           | 579,631   | 570,306   | 561,267   | 551,939   | 542,700   |
| 131 | United Arab Emirates               | 1,054,588 | 1,015,103 | 974,480   | 931,210   | 883,415   |
| 132 | United Kingdom                     | 259,344   | 204,324   | 137,674   | 80,518    | 41,729    |
| 133 | United Republic of Tanzania        | 1,582     | 1,581     | 1,578     | 1,576     | 1,572     |
| 134 | United States of America           | 4,776,408 | 4,489,046 | 4,200,976 | 3,921,848 | 3,635,680 |
| 135 | Uruguay                            |           |           |           |           |           |
| 136 | Venezuela (Bolivarian Republic of) | 1,431,829 | 1,377,597 | 1,310,444 | 1,251,207 | 1,198,785 |
| 137 | Vietnam                            | 36,030    | 33,738    | 29,275    | 21,910    | 12,615    |
| 138 | Yemen                              | 40,228    | 36,823    | 31,664    | 26,034    | 21,888    |
| 139 | Zambia                             | 191       | 170       | 157       | 144       | 143       |
| 140 | Zimbabwe                           | 8,471     | 8,090     | 7,776     | 7,501     | 7,274     |

Fossil fuels change percentage change with respect to the base year, 1990

## Fossil fuels per capita in millions of constant 2005 US\$

| 1990    | 1995    | 2000    | 2005    | 2010    | 1995 | 2000 | 2005 | 2010 |
|---------|---------|---------|---------|---------|------|------|------|------|
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
| 1,217   | 1,161   | 1,079   | 1,004   | 937     | -5   | -11  | -18  | -23  |
| 2,896   | 2,659   | 2,412   | 2,180   | 1,946   | -8   | -17  | -25  | -33  |
| 13,710  | 11,792  | 10,481  | 9,388   | 8,530   | -14  | -24  | -32  | -38  |
| 559     | 458     | 380     | 298     | 251     | -18  | -32  | -47  | -55  |
|         |         |         |         |         |      |      |      |      |
| 1,894   | 1,664   | 1,446   | 1,187   | 902     | -12  | -24  | -37  | -52  |
| 1,033   | 921     | 825     | 782     | 717     | -11  | -20  | -24  | -31  |
|         |         |         |         |         |      |      |      |      |
|         |         |         |         |         |      |      |      |      |
| 4,754   | 3,581   | 2,632   | 1,848   | 1,340   | -25  | -45  | -61  | -72  |
| 856     | 800     | 747     | 713     | 666     | -7   | -13  | -17  | -22  |
| 1,334   | 1,201   | 1,017   | 811     | 575     | -10  | -24  | -39  | -57  |
|         |         |         |         |         |      |      |      |      |
| 40,508  | 36,129  | 31,896  | 25,941  | 17,696  | -11  | -21  | -36  | -56  |
| 1,708   | 1,359   | 1,092   | 857     | 636     | -20  | -36  | -50  | -63  |
| 1,163   | 974     | 801     | 670     | 531     | -16  | -31  | -42  | -54  |
|         |         |         |         |         |      |      |      |      |
| 11,219  | 11,156  | 11,480  | 11,762  | 11,941  | -1   | 2    | 5    | 6    |
| 583,083 | 432,228 | 321,240 | 228,835 | 117,605 | -26  | -45  | -61  | -80  |
| 4,533   | 3,523   | 2,338   | 1,337   | 673     | -22  | -48  | -70  | -85  |
| 62      | 53      | 46      | 41      | 35      | -15  | -25  | -35  | -44  |
| 18,854  | 16,856  | 14,871  | 13,213  | 11,713  | -11  | -21  | -30  | -38  |
|         |         |         |         |         |      |      |      |      |
| 72,736  | 62,519  | 53,821  | 46,925  | 41,366  | -14  | -26  | -35  | -43  |
| 537     | 456     | 372     | 263     | 144     | -15  | -31  | -51  | -73  |
| 3,367   | 2,431   | 1,787   | 1,261   | 910     | -28  | -47  | -63  | -73  |
| 24      | 19      | 15      | 13      | 11      | -21  | -36  | -48  | -55  |
| 809     | 692     | 622     | 597     | 579     | -14  | -23  | -26  | -28  |

## Minerals

#### Minerals level in millions of constant 2005 US\$

|    | Country/Year                     | 1990    | 1995    | 2000    | 2005    | 2010    |
|----|----------------------------------|---------|---------|---------|---------|---------|
| 1  | Afghanistan                      |         |         |         |         |         |
| 2  | Albania                          |         |         |         |         |         |
| 3  | Algeria                          | 8,646   | 8,629   | 8,608   | 8,592   | 8,562   |
| 4  | Argentina                        |         |         |         |         |         |
| 5  | Armenia                          |         |         |         |         |         |
| 6  | Australia                        | 134,612 | 130,511 | 125,156 | 118,553 | 110,480 |
| 7  | Austria                          |         |         |         |         |         |
| 8  | Bahrain                          |         |         |         |         |         |
| 9  | Bangladesh                       |         |         |         |         |         |
| 10 | Barbados                         |         |         |         |         |         |
| 11 | Belgium                          |         |         |         |         |         |
| 12 | Belize                           |         |         |         |         |         |
| 13 | Benin                            |         |         |         |         |         |
| 14 | Bolivia (Plurinational State of) | 1,974   | 1,833   | 1,690   | 1,540   | 1,244   |
| 15 | Botswana                         | 810     | 728     | 635     | 529     | 416     |
| 16 | Brazil                           | 119,869 | 116,661 | 113,162 | 108,884 | 102,723 |
| 17 | Bulgaria                         |         |         |         |         |         |
| 18 | Burundi                          |         |         |         |         |         |
| 19 | Cambodia                         |         |         |         |         |         |
| 20 | Cameroon                         |         |         |         |         |         |
| 21 | Canada                           | 35,587  | 32,285  | 28,985  | 26,103  | 23,429  |
| 22 | Central African Republic         |         |         |         |         |         |
| 23 | Chile                            | 58,347  | 55,683  | 51,021  | 45,156  | 38,805  |
| 24 | China                            | 139,845 | 134,018 | 127,107 | 118,562 | 100,003 |
| 25 | Colombia                         | 2,191   | 2,081   | 1,921   | 1,623   | 1,377   |
| 26 | Congo                            |         |         |         |         |         |
| 27 | Costa Rica                       |         |         |         |         |         |
| 28 | Côte d'Ivoire                    |         |         |         |         |         |
| 29 | Croatia*                         |         |         |         |         |         |
| 30 | Cuba                             | 5,324   | 5,197   | 4,952   | 4,661   | 4,390   |
| 31 | Cyprus                           |         |         |         |         |         |
| 32 | Czech Republic**                 |         |         |         |         |         |
| 33 | Democratic Republic of the Congo |         |         |         |         |         |
| 34 | Denmark                          |         |         |         |         |         |
| 35 | Dominican Republic               | 1,312   | 1,136   | 967     | 828     | 766     |
| 36 | Ecuador                          |         |         |         |         |         |
Minerals change percentage change with respect to the base year, 1990

Minerals per capita in millions of constant 2005 US\$

| 1990   | 1995  | 2000  | 2005  | 2010  | 1995 | 2000 | 2005 | 2010 |
|--------|-------|-------|-------|-------|------|------|------|------|
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
| 342    | 305   | 282   | 261   | 241   | -11  | -18  | -24  | -29  |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
| 7,874  | 7,203 | 6,531 | 5,810 | 4,961 | -9   | -17  | -26  | -37  |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
| 296    | 245   | 203   | 168   | 125   | -17  | -31  | -43  | -58  |
| 586    | 459   | 361   | 282   | 207   | -22  | -38  | -52  | -65  |
| 801    | 721   | 649   | 585   | 527   | -10  | -19  | -27  | -34  |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
| 1,285  | 1,102 | 945   | 809   | 689   | -14  | -26  | -37  | -46  |
| 1 40 4 | 0.004 | 0.000 | 0.770 | 0.007 | 10   | 05   | 07   | 40   |
| 4,424  | 3,864 | 3,309 | 2,770 | 2,267 | -13  | -25  | -37  | -49  |
| 66     | 57    | 100   | 38    | 30    | -10  | -10  | -20  | -59  |
| 00     | 57    | 40    | 50    | 30    | -15  | -21  | -43  | -55  |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
| 504    | 477   | 446   | 414   | 390   | -5   | -11  | -18  | -23  |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
|        |       |       |       |       |      |      |      |      |
| 182    | 143   | 113   | 89    | 77    | -21  | -38  | -51  | -58  |
|        |       |       |       |       |      |      |      |      |

### Minerals

#### Minerals level in millions of constant 2005 US\$

|    | Country/Year                     | 1990    | 1995    | 2000    | 2005    | 2010    |
|----|----------------------------------|---------|---------|---------|---------|---------|
| 37 | Egypt                            | 524     | 518     | 498     | 462     | 398     |
| 38 | El Salvador                      |         |         |         |         |         |
| 39 | Estonia                          |         |         |         |         |         |
| 40 | Fiji                             |         |         |         |         |         |
| 41 | Finland                          |         |         |         |         |         |
| 42 | France                           |         |         |         |         |         |
| 43 | Gabon                            |         |         |         |         |         |
| 44 | Gambia                           |         |         |         |         |         |
| 45 | Germany                          |         |         |         |         |         |
| 46 | Ghana                            | 2,814   | 2,614   | 2,266   | 1,909   | 1,469   |
| 47 | Greece                           | 1,284   | 1,262   | 1,243   | 1,220   | 1,199   |
| 48 | Guatemala                        |         |         |         |         |         |
| 49 | Guyana                           | 1,748   | 1,727   | 1,703   | 1,686   | 1,670   |
| 50 | Haiti                            |         |         |         |         |         |
| 51 | Honduras                         |         |         |         |         |         |
| 52 | Hungary                          |         |         |         |         |         |
| 53 | Iceland                          |         |         |         |         |         |
| 54 | India                            | 123,148 | 122,004 | 120,632 | 118,390 | 114,646 |
| 55 | Indonesia                        | 22,124  | 21,028  | 19,097  | 16,442  | 13,800  |
| 56 | Iran (Islamic Republic of)       | 8,692   | 8,602   | 8,462   | 8,172   | 7,669   |
| 57 | Iraq                             |         |         |         |         |         |
| 58 | Ireland                          | 807     | 706     | 601     | 413     | 219     |
| 59 | Israel                           | 994     | 942     | 875     | 803     | 742     |
| 60 | Italy                            |         |         |         |         |         |
| 61 | Jamaica                          | 4,430   | 4,318   | 4,201   | 4,070   | 3,950   |
| 62 | Japan                            | 307     | 265     | 224     | 186     | 149     |
| 63 | Jordan                           | 4,882   | 4,812   | 4,725   | 4,627   | 4,538   |
| 64 | Kazakhstan*                      | 30,688  | 30,262  | 29,643  | 28,761  | 27,688  |
| 65 | Kenya                            |         |         |         |         |         |
| 66 | Kuwait                           |         |         |         |         |         |
| 67 | Kyrgyzstan*                      |         |         |         |         |         |
| 68 | Lao People's Democratic Republic |         |         |         |         |         |
| 69 | Latvia                           |         |         |         |         |         |
| 70 | Lesotho                          |         |         |         |         |         |
| 71 | Liberia                          |         |         |         |         |         |
| 72 | Lithuania*                       |         |         |         |         |         |

\* Data for this country are available from 1991, thus all data refer to 1991 instead of 1990.
\*\* Data for this country are available from 1992, thus all data refer to 1992 instead of 1990.
Empty cells represent either missing data or unavailability of the resource in the related countries.

| change<br>change<br>ear, 1990 | Minera<br>percenta<br>ct to the base | with respec |      | rals per capita<br>ant 2005 US\$ | Minerals per of in millions of constant 2009 |       |       |       |  |  |  |  |  |
|-------------------------------|--------------------------------------|-------------|------|----------------------------------|--|-------|-------|-------|--|--|--|--|--|
| 2010                          | 2005                                 | 2000        | 1995 | 2010                             | 2005   | 2000  | 1995  | 1990  |  |  |  |  |  |
| -47                           | -33                                  | -20         | -10  | 5                                | 6  | 7     | 8     | 9     |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
| -68                           | -54                                  | -38         | -19  | 60                               | 88   | 118   | 154   | 190   |  |  |  |  |  |
| -16                           | -14                                  | -10         | -6   | 106                              | 109  | 113   | 118   | 126   |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
| -8                            | -6                                   | -4          | -2   | 2,213                            | 2,260  | 2,323 | 2,373 | 2,411 |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |
| 24                            | 26                                   | 10          | 10   | 04                               | 10.4   | 11 /  | 100   | 1.41  |  |  |  |  |  |
| -54                           | -20                                  | -19         | -10  | 594                              | 104  | 90    | 120   | 141   |  |  |  |  |  |
| -52                           | -40                                  | -25         | -12  | 10.4                             | 12   | 120   | 14.4  | 120   |  |  |  |  |  |
| -30                           | -20                                  | -10         | -9   | 104                              | 117  | 150   | 144   | 100   |  |  |  |  |  |
| _79                           | -57                                  | -31         | -14  | 49                               | 99   | 158   | 196   | 229   |  |  |  |  |  |
| -55                           | -45                                  | -34         | -20  | 100                              | 122  | 145   | 177   | 220   |  |  |  |  |  |
| 00                            | 10                                   | 07          | 20   | 100                              | 122  | 110   |       | 221   |  |  |  |  |  |
| -23                           | -19                                  | -13         | -6   | 1,441                            | 1.518  | 1,627 | 1.754 | 1.873 |  |  |  |  |  |
| -53                           | -41                                  | -29         | -15  | ., 1                             | .,   | 2     | 2     | 3     |  |  |  |  |  |
| -49                           | -39                                  | -32         | -23  | 733                              | 866  | 979   | 1,098 | 1,429 |  |  |  |  |  |
| -7                            | 2                                    | 7           | 2    | 1,728                            | 1,896  | 1,982 | 1,900 | 1,858 |  |  |  |  |  |
|                               |                                      |             |      |                                  |  |       |       |       |  |  |  |  |  |

### Minerals

#### Minerals level in millions of constant 2005 US\$

|     | Country/Year        | 1990    | 1995    | 2000    | 2005   | 2010   |
|-----|---------------------|---------|---------|---------|--------|--------|
| 73  | Luxembourg          |         |         |         |        |        |
| 74  | Malawi              |         |         |         |        |        |
| 75  | Malaysia            | 174     | 146     | 132     | 124    | 118    |
| 76  | Maldives            |         |         |         |        |        |
| 77  | Mali                |         |         |         |        |        |
| 78  | Malta               |         |         |         |        |        |
| 79  | Mauritania          | 4,109   | 3,943   | 3,767   | 3,606  | 3,433  |
| 80  | Mauritius           |         |         |         |        |        |
| 81  | Mexico              | 18,782  | 17,863  | 16,723  | 15,555 | 14,155 |
| 82  | Mongolia            |         |         |         |        |        |
| 83  | Могоссо             |         |         |         |        |        |
| 84  | Mozambique          |         |         |         |        |        |
| 85  | Myanmar             |         |         |         |        |        |
| 86  | Namibia             |         |         |         |        |        |
| 87  | Nepal               |         |         |         |        |        |
| 88  | Netherlands         |         |         |         |        |        |
| 89  | New Zealand         |         |         |         |        |        |
| 90  | Nicaragua           |         |         |         |        |        |
| 91  | Niger               |         |         |         |        |        |
| 92  | Nigeria             |         |         |         |        |        |
| 93  | Norway              |         |         |         |        |        |
| 94  | Pakistan            |         |         |         |        |        |
| 95  | Panama              |         |         |         |        |        |
| 96  | Papua New Guinea    | 2,186   | 1,917   | 1,649   | 1,348  | 1,066  |
| 97  | Paraguay            |         |         |         |        |        |
| 98  | Peru                | 35,004  | 33,880  | 32,107  | 29,316 | 25,977 |
| 99  | Philippines         | 1,588   | 1,536   | 1,464   | 1,373  | 911    |
| 100 | Poland              | 8,440   | 7,987   | 7,451   | 6,832  | 6,279  |
| 101 | Portugal            | 58      | 47      | 39      | 38     | 38     |
| 102 | Qatar               |         |         |         |        |        |
| 103 | Republic of Korea   |         |         |         |        |        |
| 104 | Republic of Moldova |         |         |         |        |        |
| 105 | Romania             |         |         |         |        |        |
| 106 | Russian Federation* | 107,666 | 104,869 | 101,414 | 97,287 | 92,949 |
| 107 | Rwanda              |         |         |         |        |        |
| 108 | Saudi Arabia        |         |         |         |        |        |

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\*\* Data for this country are available from 1992, thus all data refer to 1992 instead of 1990.
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Minerals change percentage change with respect to the base year, 1990

Minerals per capita in millions of constant 2005 US\$

| 1990  | 1995  | 2000    | 2005  | 2010 | 1995 | 2000 | 2005 | 2010 |
|-------|-------|---------|-------|------|------|------|------|------|
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
| 10    | 7     | 6       | 5     | 4    | -26  | -41  | -50  | -56  |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
| 2.050 | 1720  | 1.405   | 1192  | 002  | 16   | 21   | 12   | 50   |
| 2,039 | 1,720 | 1,423   | 1,103 | 992  | -10  | -31  | -43  | -52  |
| 223   | 194   | 167     | 146   | 125  | -13  | -25  | -34  | -44  |
| 220   |       | 107     |       | 120  |      | 20   | 0.   |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
| FOG   | 106   | 206     | 201   | 166  | 00   | 40   | FO   | 70   |
| 520   | 406   | 306     | 221   | 155  | -23  | -42  | -20  | -70  |
| 1 614 | 1422  | 1 7 4 1 | 1064  | 893  | _12  | -23  | -34  | -45  |
| 26    | 22    | 19      | 1,004 | 10   | -14  | -26  | -38  | -62  |
| 222   | 208   | 195     | 179   | 164  | -6   | -12  | -19  | -26  |
| 6     | 5     | 4       | 4     | 4    | -21  | -35  | -38  | -39  |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |
| 724   | 705   | 691     | 676   | 650  | -3   | -5   | -7   | -10  |
|       |       |         |       |      |      |      |      |      |
|       |       |         |       |      |      |      |      |      |

### Minerals

#### Minerals level in millions of constant 2005 US\$

|     | Country/Year                       | 1990   | 1995   | 2000   | 2005   | 2010   |
|-----|------------------------------------|--------|--------|--------|--------|--------|
| 109 | Senegal                            | 911    | 873    | 837    | 802    | 785    |
| 110 | Serbia                             |        |        |        |        |        |
| 111 | Sierra Leone                       |        |        |        |        |        |
| 112 | Singapore                          |        |        |        |        |        |
| 113 | Slovakia**                         |        |        |        |        |        |
| 114 | Slovenia*                          |        |        |        |        |        |
| 115 | South Africa                       | 30,614 | 27,011 | 23,960 | 21,378 | 19,323 |
| 116 | Spain                              |        |        |        |        |        |
| 117 | Sri Lanka                          |        |        |        |        |        |
| 118 | Sudan (former)                     |        |        |        |        |        |
| 119 | Swaziland                          |        |        |        |        |        |
| 120 | Sweden                             | 10,603 | 10,311 | 10,005 | 9,704  | 9,380  |
| 121 | Switzerland                        |        |        |        |        |        |
| 122 | Syrian Arab Republic               | 5,590  | 5,571  | 5,537  | 5,497  | 5,446  |
| 123 | Tajikistan*                        |        |        |        |        |        |
| 124 | Thailand                           | 119    | 103    | 99     | 96     | 95     |
| 125 | Тодо                               | 414    | 362    | 314    | 287    | 268    |
| 126 | Trinidad and Tobago                |        |        |        |        |        |
| 127 | Tunisia                            | 943    | 828    | 681    | 530    | 381    |
| 128 | Turkey                             |        |        |        |        |        |
| 129 | Uganda                             |        |        |        |        |        |
| 130 | Ukraine*                           | 71,873 | 71,312 | 70,724 | 70,005 | 69,153 |
| 131 | United Arab Emirates               |        |        |        |        |        |
| 132 | United Kingdom                     |        |        |        |        |        |
| 133 | United Republic of Tanzania        |        |        |        |        |        |
| 134 | United States of America           | 56,878 | 51,334 | 45,491 | 40,906 | 36,760 |
| 135 | Uruguay                            |        |        |        |        |        |
| 136 | Venezuela (Bolivarian Republic of) | 14,157 | 13,841 | 13,534 | 13,198 | 12,879 |
| 137 | Vietnam                            |        |        |        |        |        |
| 138 | Yemen                              |        |        |        |        |        |
| 139 | Zambia                             | 6,060  | 5,629  | 5,294  | 4,901  | 4,276  |
| 140 | Zimbabwe                           |        |        |        |        |        |

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Minerals change percentage change with respect to the base year, 1990

Minerals per capita in millions of constant 2005 US\$

|       |       |       |       |       |      | •    |      | · ·  |
|-------|-------|-------|-------|-------|------|------|------|------|
| 1990  | 1995  | 2000  | 2005  | 2010  | 1995 | 2000 | 2005 | 2010 |
| 126   | 104   | 88    | 74    | 63    | -17  | -30  | -41  | -50  |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
| 832   | 652   | 535   | 447   | 385   | -22  | -36  | -46  | -54  |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
| 1,239 | 1,168 | 1,129 | 1,075 | 1,000 | -6   | -9   | -13  | -19  |
|       |       |       |       |       |      |      |      |      |
| 454   | 393   | 346   | 297   | 267   | -13  | -24  | -34  | -41  |
|       |       |       |       |       |      |      |      |      |
| 2     | 2     | 2     | 1     | 1     | -17  | -25  | -31  | -34  |
| 113   | 89    | 65    | 53    | 45    | -22  | -42  | -53  | -61  |
|       |       |       |       |       |      |      |      |      |
| 115   | 93    | 72    | 53    | 36    | -19  | -37  | -53  | -68  |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
| 1,391 | 1,395 | 1,447 | 1,492 | 1,522 | 0    | 4    | 7    | 9    |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
| 225   | 193   | 161   | 138   | 118   | -14  | -28  | -39  | -47  |
|       |       |       |       |       |      |      |      |      |
| 719   | 628   | 556   | 495   | 444   | -13  | -23  | -31  | -38  |
|       |       |       |       |       |      |      |      |      |
|       |       |       |       |       |      |      |      |      |
| 771   | 631   | 519   | 428   | 327   | -18  | -33  | -45  | -58  |
|       |       |       |       |       |      |      |      |      |

|    |                                  | IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj | IWI   | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP     | IWI<br>adj |
|----|----------------------------------|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|-------|-----------------|----------------------|---------------------------|----------------------------|---------|------------|
|    | Country/Year                     | aver | age gro         | wth rat              | e (in pe                  | rcentage                   | e) 1991- | -1995      | avera | age gro         | wth rat              | e (in pe                  | rcentage                   | ) 1996- | -2000      |
| 1  | Afghanistan                      | 5.3  | 8.7             | -3.5                 | -0.1                      | -0.1                       | 15.4     | 11.7       | 1.9   | 2.9             | -1.1                 | -0.1                      | -0.1                       | 6.1     | 4.9        |
| 2  | Albania                          | -1.0 | -0.9            | -0.1                 | -0.1                      | -0.1                       | 2.4      | 2.1        | -0.2  | -0.4            | 0.2                  | -0.1                      | -0.1                       | -4.6    | -4.5       |
| 3  | Algeria                          | 0.9  | 2.3             | -1.3                 | -0.1                      | 0.9                        | 2.7      | 2.1        | 0.7   | 1.5             | -0.8                 | -0.1                      | 1.0                        | -0.8    | -0.7       |
| 4  | Argentina                        | 1.0  | 1.3             | -0.4                 | -0.1                      | -0.1                       | -4.0     | -4.5       | 1.4   | 1.2             | 0.2                  | -0.1                      | -0.1                       | -0.5    | -0.4       |
| 5  | Armenia                          | -1.3 | -1.9            | 0.6                  | -0.1                      | -0.4                       | 15.2     | 15.3       | -0.3  | -0.9            | 0.6                  | -0.1                      | -0.7                       | -5.6    | -5.7       |
| 6  | Australia                        | 1.1  | 1.2             | -0.1                 | 0.0                       | -0.1                       | -1.3     | -1.4       | 1.3   | 1.1             | 0.1                  | 0.0                       | -0.1                       | -1.4    | -1.4       |
| 7  | Austria                          | 2.3  | 0.7             | 1.6                  | -0.2                      | -0.1                       | 0.1      | 1.4        | 1.1   | 0.2             | 0.9                  | -0.2                      | -0.1                       | -2.3    | -1.7       |
| 8  | Bahrain                          | 4.0  | 2.6             | 1.4                  | -0.2                      | 0.0                        | -0.7     | 0.5        | 3.1   | 2.7             | 0.4                  | -0.2                      | -0.1                       | -0.8    | -0.6       |
| 9  | Bangladesh                       | 2.7  | 2.2             | 0.5                  | -0.2                      | -0.1                       | -0.3     | -0.1       | 3.8   | 2.0             | 1.8                  | -0.1                      | -0.1                       | 0.4     | 2.0        |
| 10 | Barbados                         | 0.5  | 0.3             | 0.2                  | -0.1                      | -0.2                       | 1.2      | 1.2        | 0.8   | 0.3             | 0.4                  | -0.1                      | -0.2                       | -1.4    | -1.3       |
| 11 | Belgium                          | 1.5  | 0.3             | 1.3                  | -0.2                      | -0.2                       | 0.6      | 1.6        | 1.3   | 0.2             | 1.1                  | -0.2                      | -0.2                       | -0.6    | 0.1        |
| 12 | Belize                           | 1.0  | 3.0             | -2.0                 | 0.0                       | -0.1                       | -0.7     | -2.9       | 1.1   | 2.6             | -1.5                 | 0.0                       | -0.2                       | -2.9    | -4.7       |
| 13 | Benin                            | 1.1  | 3.4             | -2.3                 | -0.1                      | -0.1                       | -1.6     | -4.1       | 1.6   | 2.9             | -1.3                 | -0.1                      | -0.1                       | -2.3    | -3.9       |
| 14 | Bolivia (Plurinational State of) | -0.1 | 2.3             | -2.5                 | 0.0                       | 0.0                        | -1.2     | -3.7       | 0.0   | 2.1             | -2.2                 | 0.0                       | 0.0                        | 1.0     | -1.2       |
| 15 | Botswana                         | 1.7  | 2.8             | -1.1                 | -0.2                      | -0.1                       | 0.2      | -1.1       | 1.5   | 2.1             | -0.6                 | -0.2                      | -0.1                       | -4.8    | -5.7       |
| 16 | Brazil                           | 1.8  | 1.6             | 0.2                  | -0.1                      | -0.1                       | -0.5     | -0.4       | 1.6   | 1.5             | 0.1                  | -0.1                      | -0.1                       | 0.2     | 0.1        |
| 17 | Bulgaria                         | -0.7 | -1.1            | 0.4                  | 0.0                       | -0.4                       | 1.9      | 1.8        | -0.5  | -0.9            | 0.3                  | 0.0                       | -0.4                       | 0.5     | 0.4        |
| 18 | Burundi                          | 1.7  | 1.7             | 0.0                  | -0.1                      | -0.1                       | 5.1      | 4.8        | 0.4   | 0.9             | -0.6                 | -0.1                      | -0.1                       | 1.6     | 0.8        |
| 19 | Cambodia                         | 0.3  | 3.2             | -3.0                 | -0.1                      | 0.0                        | -3.5     | -6.6       | 0.5   | 2.2             | -1.7                 | -0.1                      | -0.2                       | -3.5    | -5.5       |
| 20 | Cameroon                         | 0.3  | 2.7             | -2.4                 | -0.1                      | 0.0                        | 2.8      | 0.4        | 0.6   | 2.4             | -1.7                 | -0.1                      | 0.0                        | -3.0    | -4.8       |
| 21 | Canada                           | 0.7  | 1.1             | -0.5                 | 0.0                       | 0.6                        | -0.3     | -0.2       | 1.2   | 0.9             | 0.3                  | 0.0                       | 0.7                        | -2.0    | -1.0       |
| 22 | Central African Republic         | 0.0  | 2.5             | -2.6                 | 0.0                       | 0.0                        | 1.3      | -1.3       | -0.1  | 2.2             | -2.2                 | 0.0                       | 0.0                        | -0.2    | -2.4       |
| 23 | Chile                            | 1.9  | 1.8             | 0.1                  | -0.1                      | -0.2                       | -4.1     | -4.3       | 2.3   | 1.4             | 0.9                  | -0.1                      | -0.3                       | 0.1     | 0.6        |
| 24 | China                            | 2.1  | 1.2             | 0.9                  | 0.0                       | -0.1                       | -7.1     | -6.2       | 2.3   | 0.9             | 1.4                  | 0.0                       | -0.2                       | -3.2    | -2.0       |
| 25 | Colombia                         | 1.7  | 1.9             | -0.2                 | -0.1                      | 0.0                        | -1.2     | -1.4       | 1.5   | 1.8             | -0.2                 | -0.1                      | 0.0                        | 1.3     | 1.0        |
| 26 | Congo                            | 0.1  | 2.7             | -2.7                 | 0.0                       | 0.6                        | 1.9      | -0.2       | -0.3  | 2.8             | -3.0                 | 0.0                       | 0.7                        | -0.6    | -3.0       |
| 27 | Costa Rica                       | 2.4  | 2.5             | -0.1                 | -0.1                      | -0.2                       | -2.0     | -2.4       | 3.5   | 2.5             | 1.0                  | -0.1                      | -0.2                       | -0.7    | 0.0        |
| 28 | Côte d'Ivoire                    | 2.5  | 3.2             | -0.7                 | -0.2                      | -0.1                       | 0.9      | -0.1       | 1.7   | 2.5             | -0.7                 | -0.1                      | -0.1                       | -1.5    | -2.5       |
| 29 | Croatia*                         | 0.9  | 0.7             | 0.2                  | 0.0                       | -0.1                       | 2.3      | 2.4        | 0.0   | -0.7            | 0.7                  | 0.0                       | -0.2                       | -1.2    | -0.7       |
| 30 | Cuba                             | 0.1  | 0.6             | -0.5                 | -0.1                      | -0.5                       | 7.7      | 6.5        | 0.2   | 0.4             | -0.2                 | -0.1                      | -0.6                       | -4.5    | -5.4       |
| 31 | Cyprus                           | 2.8  | 2.2             | 0.6                  | -0.1                      | -0.3                       | -2.2     | -2.0       | 3.1   | 2.0             | 1.1                  | -0.1                      | -0.3                       | -1.2    | -0.5       |
| 32 | Czech Republic**                 | 1.9  | 0.0             | 1.9                  | 0.0                       | -0.2                       | -0.8     | 0.9        | 0.9   | -0.1            | 1.0                  | 0.0                       | -0.2                       | -1.6    | -0.7       |
| 33 | Democratic Republic of the Congo | 0.2  | 3.9             | -3.7                 | 0.0                       | 0.0                        | 9.2      | 5.6        | 0.1   | 2.4             | -2.3                 | 0.0                       | 0.0                        | 4.6     | 2.3        |
| 34 | Denmark                          | 0.1  | 0.4             | -0.3                 | -0.2                      | 0.0                        | -1.8     | -2.3       | 0.8   | 0.4             | 0.4                  | -0.2                      | 0.0                        | -0.6    | -0.5       |
| 35 | Dominican Republic               | 2.7  | 1.9             | 0.8                  | -0.1                      | -0.3                       | -1.7     | -1.3       | 2.4   | 1.7             | 0.8                  | -0.1                      | -0.4                       | -2.1    | -1.8       |
| 36 | Ecuador                          | 0.8  | 2.1             | -1.3                 | -0.1                      | 0.9                        | -1.0     | -1.5       | 0.4   | 1.6             | -1.3                 | -0.1                      | 1.0                        | 0.2     | -0.1       |

|   | IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj | IW  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj |
|---|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|-----|-----------------|----------------------|---------------------------|----------------------------|----------|------------|
|   | av   | erage gi        | rowth rate           | e (in perc                | entage) 2                  | 001-2005 | 5          |     | average gi      | owth rate            | e (in perc                | entage) 2                  | 006-2010 | )          |
| 7 | 4.2  | 3.9             | 0.4                  | -0.1                      | -0.1                       | -11.3    | -11.1      | 3.  | 5 2.6           | 0.8                  | -0.1                      | -0.2                       | -4.2     | -3.7       |
|   | 1.7  | 0.5             | 1.2                  | -0.1                      | -0.2                       | -3.9     | -3.0       | 2.  | 0.4             | 1.6                  | -0.1                      | -0.2                       | -3.8     | -2.5       |
|   | 0.8  | 1.5             | -0.7                 | -0.1                      | 1.0                        | -2.9     | -2.7       | 1.  | 3 1.5           | -0.2                 | -0.1                      | 0.8                        | -1.1     | -0.5       |
|   | 1.6  | 0.9             | 0.6                  | -0.1                      | -0.1                       | -0.4     | 0.1        | 1.  | 7 0.9           | 0.8                  | -0.1                      | -0.2                       | -3.3     | -2.8       |
|   | 0.9  | -0.1            | 1.0                  | -0.1                      | -0.9                       | -9.1     | -9.2       | 2.  | 0.2             | 1.9                  | -0.1                      | -1.1                       | 2.0      | 2.6        |
|   | 1.8  | 1.3             | 0.6                  | 0.0                       | -0.1                       | -0.2     | 0.3        | 2.  | 5 1.8           | 0.7                  | 0.0                       | -0.1                       | 0.5      | 1.1        |
|   | 1.8  | 0.6             | 1.2                  | -0.2                      | -0.1                       | 0.4      | 1.3        | 1.  | 2 0.4           | 0.8                  | -0.2                      | -0.1                       | 0.2      | 0.8        |
|   | 3.6  | 2.6             | 1.0                  | -0.2                      | -0.2                       | -1.4     | -0.8       | 12. | 3 11.7          | 0.6                  | -0.1                      | -0.4                       | 2.2      | 2.2        |
|   | 3.2  | 1.6             | 1.6                  | -0.1                      | -0.2                       | 0.2      | 1.5        | 3.  | ) 1.1           | 1.8                  | -0.1                      | -0.2                       | -0.6     | 0.9        |
|   | 1.6  | 0.2             | 1.4                  | -0.1                      | -0.2                       | -0.6     | 0.4        | 0.  | 6 0.2           | 0.4                  | -0.1                      | -0.2                       | 1.0      | 1.1        |
|   | 1.7  | 0.5             | 1.2                  | -0.2                      | -0.2                       | 0.6      | 1.5        | 1   | 1 0.6           | 0.5                  | -0.2                      | -0.2                       | 1.3      | 1.4        |
|   | 1.3  | 2.3             | -1.0                 | 0.0                       | -0.4                       | -2.6     | -4.1       | 1.0 | ) 2.1           | -1.1                 | -0.1                      | -0.4                       | 0.0      | -1.5       |
|   | 2.0  | 3.2             | -1.2                 | -0.1                      | -0.2                       | -1.0     | -2.6       | 2.  | 7 3.0           | -0.3                 | -0.1                      | -0.4                       | -0.3     | -1.1       |
|   | 0.0  | 1.9             | -2.0                 | 0.0                       | 0.0                        | -0.4     | -2.4       | -0  | 1 1.7           | -1.8                 | 0.0                       | 0.0                        | -1.2     | -3.1       |
|   | 1.9  | 1.3             | 0.6                  | -0.2                      | -0.1                       | -2.7     | -2.4       | 1.  | 7 1.4           | 0.3                  | -0.2                      | -0.2                       | -1.4     | -1.5       |
|   | 1.3  | 1.3             | 0.0                  | -0.1                      | -0.1                       | -1.2     | -1.4       | 1   | 1 0.9           | 0.1                  | -0.1                      | -0.2                       | -2.0     | -2.1       |
|   | 0.4  | -0.7            | 1.0                  | 0.0                       | -0.4                       | -4.2     | -3.6       | 1.  | 5 –0.6          | 2.1                  | 0.0                       | -0.5                       | -0.7     | 0.9        |
|   | 2.8  | 2.6             | 0.2                  | -0.1                      | -0.2                       | -0.6     | -0.7       | 2.  | 9 2.9           | -0.1                 | -0.1                      | -0.1                       | -3.2     | -3.5       |
|   | 0.8  | 1.4             | -0.7                 | -0.1                      | -0.3                       | -4.1     | -5.1       | 1   | 1 1.1           | -0.1                 | -0.1                      | -0.4                       | -0.2     | -0.7       |
|   | 0.6  | 2.3             | -1.7                 | -0.1                      | 0.0                        | -1.9     | -3.8       | 1.0 | ) 2.2           | -1.3                 | -0.1                      | 0.0                        | -0.6     | -2.1       |
|   | 2.1  | 1.0             | 1.0                  | 0.0                       | 0.8                        | 0.4      | 2.3        | 1.  | 3 1.1           | 0.3                  | 0.0                       | 1.0                        | 1.7      | 3.0        |
|   | -0.1 | 1.7             | -1.7                 | 0.0                       | 0.0                        | 0.9      | -0.9       | 0.  | ) 1.8           | -1.9                 | 0.0                       | 0.0                        | -1.9     | -3.8       |
|   | 2.5  | 1.1             | 1.3                  | -0.1                      | -0.3                       | -0.2     | 0.8        | 3   | 1 1.0           | 2.1                  | -0.1                      | -0.3                       | 2.4      | 4.1        |
|   | 2.6  | 0.6             | 2.0                  | 0.0                       | -0.2                       | -3.4     | -1.7       | 4   | 1 0.5           | 3.6                  | 0.0                       | -0.4                       | -8.5     | -5.2       |
|   | 1.2  | 1.6             | -0.4                 | -0.1                      | 0.0                        | -1.9     | -2.4       | 1.9 | 9 1.5           | 0.4                  | -0.1                      | -0.1                       | -0.8     | -0.6       |
|   | -0.2 | 2.4             | -2.7                 | -0.1                      | 0.7                        | -2.4     | -4.4       | 0.  | 3 2.7           | -2.5                 | -0.1                      | 0.7                        | -2.1     | -3.9       |
|   | 2.9  | 1.9             | 0.9                  | -0.1                      | -0.2                       | -0.3     | 0.3        | 2.  | 9 1.6           | 1.3                  | -0.1                      | -0.2                       | 0.3      | 1.3        |
|   | 1.6  | 1.7             | -0.1                 | -0.1                      | -0.1                       | 1.3      | 1.0        | 1.  | 2 1.8           | -0.6                 | -0.2                      | -0.1                       | -1.3     | -2.1       |
|   | 1.0  | -0.3            | 1.3                  | 0.0                       | -0.2                       | -1.2     | -0.1       | 1   | 1 -0.2          | 1.3                  | 0.0                       | -0.2                       | 3.0      | 4.1        |
|   | 0.6  | 0.3             | 0.3                  | -0.1                      | -0.7                       | -5.0     | -5.5       | 0.  | 3 0.0           | 0.8                  | -0.1                      | -0.6                       | -3.8     | -3.7       |
|   | 2.3  | 1.8             | 0.4                  | -0.1                      | -0.4                       | -0.4     | -0.4       | 3   | 1 1.3           | 1.8                  | -0.1                      | -0.4                       | 1.2      | 2.5        |
|   | 1.8  | 0.0             | 1.8                  | 0.0                       | -0.2                       | -2.1     | -0.5       | 1   | 1 0.5           | 0.6                  | 0.0                       | -0.2                       | -1.2     | -0.9       |
|   | 0.3  | 3.0             | -2.6                 | 0.0                       | 0.0                        | -3.1     | -5.7       | 0   | 4 2.8           | -2.4                 | 0.0                       | 0.0                        | -2.0     | -4.4       |
|   | 1.2  | 0.3             | 0.9                  | -0.2                      | 0.0                        | 1.3      | 2.0        | 1.  | 4 0.5           | 0.9                  | -0.2                      | 0.0                        | 2.4      | 3.1        |
|   | 2.7  | 1.5             | 1.2                  | -0.1                      | -0.5                       | 0.4      | 1.0        | 2.  | 7 1.4           | 1.3                  | -0.1                      | -0.5                       | -2.3     | -1.5       |
|   | 0.7  | 1.7             | -1.0                 | -0.1                      | 1.2                        | -3.4     | -3.2       | 1.  | 2 1.5           | -0.3                 | -0.1                      | 1.1                        | -2.0     | -1.2       |

|    |                                  | IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama–<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj | IWI   | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP     | IWI<br>adj |
|----|----------------------------------|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|-------|-----------------|----------------------|---------------------------|----------------------------|---------|------------|
|    | Country/Year                     | aver | age gro         | wth rat              | te (in pe                 | rcentag                    | e) 1991- | -1995      | avera | age gro         | wth rat              | e (in pe                  | rcentage                   | ) 1996- | -2000      |
| 37 | Egypt                            | 1.8  | 1.8             | 0.0                  | -0.2                      | -0.1                       | -1.0     | -1.2       | 2.4   | 1.7             | 0.7                  | -0.2                      | -0.2                       | -1.2    | -0.9       |
| 38 | El Salvador                      | 2.6  | 1.5             | 1.2                  | -0.1                      | -0.2                       | -2.8     | -2.0       | 2.2   | 0.7             | 1.5                  | -0.1                      | -0.3                       | 0.0     | 1.1        |
| 39 | Estonia                          | -0.5 | -1.7            | 1.2                  | 0.0                       | -0.2                       | 7.3      | 8.2        | 1.0   | -1.0            | 1.9                  | 0.0                       | -0.2                       | -4.8    | -3.1       |
| 40 | Fiji                             | 2.7  | 1.3             | 1.4                  | -0.1                      | -0.2                       | 0.5      | 1.6        | 0.8   | 0.9             | -0.1                 | -0.1                      | -0.3                       | -1.8    | -2.3       |
| 41 | Finland                          | 0.8  | 0.5             | 0.4                  | -0.1                      | -0.1                       | 1.4      | 1.5        | 0.1   | 0.3             | -0.2                 | -0.1                      | -0.1                       | -3.8    | -4.3       |
| 42 | France                           | 1.5  | 0.4             | 1.1                  | -0.2                      | -0.1                       | 0.3      | 1.2        | 1.7   | 0.4             | 1.3                  | -0.2                      | -0.1                       | -0.7    | 0.3        |
| 43 | Gabon                            | 0.5  | 3.2             | -2.7                 | -0.1                      | 0.8                        | -0.6     | -2.7       | 0.6   | 2.6             | -2.0                 | -0.1                      | 0.8                        | 2.5     | 1.3        |
| 44 | Gambia                           | 2.8  | 3.1             | -0.3                 | -0.2                      | -0.3                       | -0.1     | -1.0       | 1.8   | 2.9             | -1.1                 | -0.2                      | -0.3                       | -2.5    | -4.0       |
| 45 | Germany                          | 2.3  | 0.7             | 1.6                  | -0.2                      | -0.1                       | 0.5      | 1.9        | 1.4   | 0.1             | 1.3                  | -0.2                      | -0.1                       | 0.1     | 1.1        |
| 46 | Ghana                            | 1.4  | 2.8             | -1.4                 | -0.2                      | -0.2                       | -1.4     | -3.2       | 1.4   | 2.4             | -1.1                 | -0.2                      | -0.2                       | -1.9    | -3.4       |
| 47 | Greece                           | 1.4  | 1.0             | 0.4                  | -0.2                      | -0.2                       | 0.6      | 0.6        | 1.2   | 0.6             | 0.6                  | -0.2                      | -0.3                       | -2.3    | -2.1       |
| 48 | Guatemala                        | 2.4  | 2.3             | 0.0                  | -0.1                      | -0.1                       | -1.7     | -1.9       | 2.7   | 2.3             | 0.3                  | -0.1                      | -0.2                       | -0.6    | -0.6       |
| 49 | Guyana                           | 0.1  | 0.1             | 0.0                  | 0.0                       | 0.0                        | -4.1     | -4.1       | 0.1   | 0.1             | 0.0                  | 0.0                       | -0.1                       | 0.0     | -0.1       |
| 50 | Haiti                            | 2.9  | 2.0             | 0.9                  | -0.2                      | -0.2                       | 5.5      | 6.0        | 2.9   | 1.9             | 1.1                  | -0.2                      | -0.3                       | -0.1    | 0.5        |
| 51 | Honduras                         | 0.9  | 2.7             | -1.8                 | -0.1                      | -0.1                       | -0.4     | -2.3       | 1.6   | 2.2             | -0.6                 | -0.1                      | -0.2                       | 0.5     | -0.4       |
| 52 | Hungary                          | 1.5  | -0.1            | 1.6                  | 0.0                       | -0.2                       | 3.9      | 5.2        | 1.0   | -0.2            | 1.2                  | 0.0                       | -0.2                       | -1.6    | -0.6       |
| 53 | Iceland                          | 1.0  | 1.0             | 0.0                  | -0.1                      | -0.1                       | 1.2      | 1.0        | 1.1   | 1.0             | 0.1                  | -0.1                      | -0.1                       | -2.6    | -2.7       |
| 54 | India                            | 1.9  | 2.0             | -0.1                 | -0.2                      | -0.1                       | -1.4     | -1.8       | 2.4   | 1.8             | 0.7                  | -0.2                      | -0.2                       | -1.8    | -1.5       |
| 55 | Indonesia                        | 1.1  | 1.6             | -0.5                 | -0.1                      | 0.0                        | -3.7     | -4.3       | 1.5   | 1.4             | 0.1                  | -0.1                      | -0.1                       | 2.8     | 2.7        |
| 56 | Iran (Islamic Republic of)       | 0.6  | 1.7             | -1.1                 | 0.0                       | 1.9                        | -1.0     | -0.3       | 0.9   | 1.8             | -0.9                 | 0.0                       | 2.2                        | -2.0    | -0.7       |
| 57 | Iraq                             | 0.1  | 3.2             | -3.0                 | 0.0                       | 4.1                        | 33.7     | 34.8       | 0.0   | 3.3             | -3.3                 | 0.0                       | 5.1                        | -18.5   | -16.7      |
| 58 | Ireland                          | 1.6  | 0.4             | 1.1                  | -0.2                      | -0.1                       | -2.7     | -1.9       | 3.3   | 1.0             | 2.3                  | -0.2                      | -0.2                       | -4.2    | -2.3       |
| 59 | Israel                           | 4.8  | 3.5             | 1.3                  | 0.0                       | -0.2                       | -1.7     | -0.6       | 3.5   | 2.4             | 1.1                  | 0.0                       | -0.3                       | -0.7    | 0.2        |
| 60 | Italy                            | 0.8  | 0.0             | 0.8                  | -0.2                      | -0.1                       | -1.0     | -0.5       | 0.9   | 0.0             | 0.9                  | -0.2                      | -0.2                       | -1.6    | -1.0       |
| 61 | Jamaica                          | 2.1  | 0.8             | 1.3                  | -0.1                      | -0.4                       | -0.1     | 0.7        | 1.4   | 1.0             | 0.4                  | -0.1                      | -0.6                       | 1.7     | 1.4        |
| 62 | Japan                            | 2.0  | 0.4             | 1.6                  | 0.0                       | -0.1                       | -0.3     | 1.2        | 1.3   | 0.2             | 1.1                  | 0.0                       | -0.2                       | -0.3    | 0.6        |
| 63 | Jordan                           | 6.7  | 5.1             | 1.6                  | -0.1                      | -0.8                       | 0.1      | 0.8        | 2.3   | 2.0             | 0.3                  | -0.1                      | -0.9                       | -0.8    | -1.4       |
| 64 | Kazakhstan*                      | 0.1  | -0.9            | 1.0                  | 0.0                       | 0.8                        | 10.5     | 12.3       | -0.4  | -1.2            | 0.9                  | 0.0                       | 1.2                        | -2.0    | 0.0        |
| 65 | Kenya                            | 3.4  | 3.2             | 0.2                  | -0.2                      | -0.2                       | 1.4      | 1.3        | 2.3   | 2.6             | -0.4                 | -0.2                      | -0.3                       | 0.4     | -0.4       |
| 66 | Kuwait                           | -0.5 | -4.8            | 4.4                  | 0.0                       | 5.0                        | 0.7      | 10.0       | -0.1  | 3.6             | -3.7                 | -0.1                      | 6.2                        | 1.6     | 4.1        |
| 67 | Kyrgyzstan*                      | 0.7  | 0.8             | -0.1                 | -0.1                      | -0.6                       | 18.0     | 17.1       | 0.8   | 1.5             | -0.7                 | -0.1                      | -0.4                       | -3.4    | -4.6       |
| 68 | Lao People's Democratic Republic | 0.0  | 2.7             | -2.7                 | 0.0                       | 0.0                        | -1.2     | -4.0       | 0.3   | 2.1             | -1.8                 | 0.0                       | 0.0                        | 1.9     | 0.0        |
| 69 | Latvia                           | -0.9 | -1.3            | 0.4                  | 0.0                       | -0.3                       | 13.2     | 13.4       | 0.4   | -0.9            | 1.3                  | 0.0                       | -0.3                       | -3.9    | -3.0       |
| 70 | Lesotho                          | 3.4  | 1.8             | 1.6                  | -0.1                      | -0.1                       | 1.8      | 3.2        | 3.0   | 1.8             | 1.1                  | -0.1                      | -0.1                       | 0.3     | 1.3        |
| 71 | Liberia                          | -0.7 | -0.3            | -0.4                 | 0.0                       | -0.1                       | 29.9     | 29.4       | 0.0   | 6.3             | -6.3                 | 0.0                       | -0.1                       | -21.6   | -28.0      |
| 72 | Lithuania*                       | -0.2 | -0.5            | 0.3                  | 0.0                       | -0.4                       | 13.4     | 13.3       | 0.2   | -0.7            | 0.9                  | 0.0                       | -0.3                       | -3.7    | -3.1       |

| IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj | I | wı  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj |
|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|---|-----|-----------------|----------------------|---------------------------|----------------------------|----------|------------|
| av   | erage gi        | rowth rate           | e (in perc                | entage) 2                  | 001–2005 | 5          |   | av  | erage gro       | owth rate            | e (in perc                | entage) 2                  | 006-2010 | )          |
| 2.3  | 1.9             | 0.4                  | -0.1                      | -0.3                       | 0.0      | -0.1       |   | 3.3 | 1.8             | 1.5                  | -0.2                      | -0.6                       | -0.3     | 0.5        |
| 2.5  | 0.4             | 2.1                  | -0.1                      | -0.4                       | 0.6      | 2.3        |   | 2.0 | 0.5             | 1.5                  | -0.1                      | -0.4                       | 0.8      | 1.8        |
| 1.6  | -0.4            | 1.9                  | 0.0                       | -0.2                       | -3.3     | -1.6       |   | 1.8 | -0.1            | 1.8                  | 0.0                       | -0.2                       | 3.5      | 5.0        |
| 1.3  | 0.3             | 1.0                  | -0.1                      | -0.6                       | 0.2      | 0.5        |   | 1.5 | 0.9             | 0.6                  | -0.1                      | -0.6                       | 2.2      | 2.1        |
| 2.0  | 0.3             | 1.7                  | -0.1                      | -0.1                       | -0.4     | 1.0        |   | 1.3 | 0.5             | 0.9                  | -0.2                      | -0.1                       | 0.6      | 1.1        |
| 1.7  | 0.7             | 1.0                  | -0.2                      | -0.1                       | 0.7      | 1.4        |   | 1.6 | 0.6             | 1.0                  | -0.2                      | -0.1                       | 1.5      | 2.2        |
| 0.7  | 2.1             | -1.4                 | -0.1                      | 0.8                        | -0.3     | -0.9       |   | 1.1 | 1.9             | -0.8                 | -0.1                      | 0.8                        | -0.8     | -0.8       |
| 3.6  | 3.0             | 0.6                  | -0.2                      | -0.4                       | 5.6      | 5.7        |   | 3.7 | 2.8             | 0.9                  | -0.2                      | -0.5                       | 5.3      | 5.6        |
| 2.4  | 0.0             | 2.4                  | -0.2                      | -0.1                       | 1.0      | 3.1        |   | 0.5 | -0.1            | 0.6                  | -0.2                      | -0.1                       | 0.0      | 0.3        |
| 1.7  | 2.5             | -0.8                 | -0.2                      | -0.3                       | -2.8     | -4.0       |   | 2.6 | 2.4             | 0.2                  | -0.2                      | -0.4                       | -2.6     | -3.0       |
| 1.5  | 0.4             | 1.1                  | -0.2                      | -0.3                       | -2.2     | -1.6       |   | 2.2 | 0.3             | 1.9                  | -0.2                      | -0.3                       | 1.2      | 2.6        |
| 2.5  | 2.5             | 0.0                  | -0.1                      | -0.2                       | -0.1     | -0.4       |   | 3.2 | 2.5             | 0.7                  | -0.1                      | -0.2                       | -0.9     | -0.6       |
| 0.1  | 0.4             | -0.3                 | 0.0                       | -0.1                       | 0.7      | 0.3        |   | 0.2 | 0.2             | 0.0                  | 0.0                       | -0.1                       | -1.4     | -1.5       |
| 2.3  | 1.6             | 0.7                  | -0.2                      | -0.4                       | 3.2      | 3.3        |   | 2.5 | 1.3             | 1.1                  | -0.2                      | -0.4                       | 1.4      | 1.9        |
| 1.4  | 2.0             | -0.6                 | -0.1                      | -0.3                       | -2.0     | -3.0       |   | 2.5 | 2.0             | 0.5                  | -0.1                      | -0.3                       | 0.1      | 0.1        |
| 1.5  | -0.2            | 1.7                  | 0.0                       | -0.2                       | -1.8     | -0.3       |   | 0.5 | -0.2            | 0.7                  | 0.0                       | -0.2                       | 1.9      | 2.4        |
| 1.7  | 1.1             | 0.6                  | -0.1                      | -0.1                       | -1.3     | -0.9       |   | 0.9 | 1.5             | -0.6                 | -0.1                      | -0.1                       | 1.9      | 1.0        |
| 2.5  | 1.6             | 0.9                  | -0.2                      | -0.2                       | -1.9     | -1.4       |   | 3.0 | 1.4             | 1.6                  | -0.2                      | -0.3                       | -0.7     | 0.5        |
| 1.1  | 1.3             | -0.2                 | -0.1                      | -0.2                       | -2.1     | -2.6       |   | 1.5 | 1.1             | 0.4                  | -0.1                      | -0.3                       | -1.7     | -1.7       |
| 1.2  | 1.3             | -0.1                 | 0.0                       | 2.6                        | -3.1     | -0.7       |   | 0.4 | 1.2             | -0.8                 | 0.0                       | 2.9                        | -0.5     | 1.6        |
| 0.0  | 2.8             | -2.8                 | 0.0                       | 6.4                        | 6.2      | 9.8        |   | 0.2 | 3.0             | -2.8                 | 0.0                       | 7.8                        | 0.6      | 5.5        |
| 3.9  | 1.8             | 2.1                  | -0.2                      | -0.2                       | 0.3      | 2.1        |   | 2.0 | 1.5             | 0.6                  | -0.2                      | -0.2                       | 3.6      | 3.9        |
| 2.3  | 1.9             | 0.4                  | 0.0                       | -0.2                       | 0.4      | 0.6        |   | 3.1 | 2.4             | 0.7                  | 0.0                       | -0.2                       | -0.7     | -0.2       |
| 1.7  | 0.6             | 1.1                  | -0.2                      | -0.1                       | 1.0      | 1.7        |   | 1.1 | 0.6             | 0.5                  | -0.2                      | -0.1                       | 1.9      | 2.0        |
| 1.2  | 0.8             | 0.4                  | -0.1                      | -0.6                       | 0.5      | 0.2        |   | 1.0 | 0.4             | 0.5                  | -0.1                      | -0.6                       | 1.4      | 1.3        |
| 0.7  | 0.1             | 0.6                  | 0.0                       | -0.1                       | -1.0     | -0.6       |   | 0.4 | 0.0             | 0.3                  | 0.0                       | -0.1                       | 0.0      | 0.1        |
| 2.3  | 2.0             | 0.2                  | -0.1                      | -0.9                       | -3.6     | -4.4       |   | 3.7 | 3.0             | 0.7                  | -0.1                      | -0.8                       | -2.2     | -2.5       |
| 0.1  | 0.3             | -0.2                 | 0.0                       | 1.5                        | -8.6     | -7.4       |   | 0.8 | 1.1             | -0.3                 | 0.0                       | 1.7                        | -4.8     | -3.4       |
| 3.1  | 2.6             | 0.4                  | -0.1                      | -0.3                       | -1.0     | -1.0       |   | 3.2 | 2.6             | 0.5                  | -0.1                      | -0.3                       | 0.4      | 0.5        |
| 0.1  | 3.1             | -3.1                 | -0.1                      | 7.7                        | -3.6     | 0.9        |   | 0.4 | 3.9             | -3.5                 | -0.1                      | 9.2                        | 2.2      | 7.9        |
| 0.5  | 0.4             | 0.1                  | -0.1                      | -0.4                       | -2.8     | -3.2       |   | 2.4 | 1.1             | 1.3                  | -0.1                      | -0.8                       | -2.2     | -1.9       |
| 0.4  | 1.6             | -1.2                 | 0.0                       | 0.0                        | 3.3      | 2.1        |   | 1.1 | 1.5             | -0.4                 | 0.0                       | 0.0                        | 0.3      | -0.1       |
| 2.5  | -0.7            | 3.1                  | 0.0                       | -0.3                       | -3.6     | -0.7       |   | 1.5 | -0.5            | 2.0                  | 0.0                       | -0.3                       | 4.1      | 5.7        |
| 1.3  | 1.0             | 0.2                  | -0.1                      | -0.1                       | -1.5     | -1.5       |   | 1.6 | 1.0             | 0.6                  | -0.1                      | -0.1                       | -1.9     | -1.5       |
| -0.3 | 2.3             | -2.6                 | 0.0                       | -0.1                       | 6.5      | 3.8        |   | 0.5 | 4.6             | -4.2                 | 0.0                       | -0.2                       | -3.5     | -7.9       |
| 1.5  | -0.5            | 1.9                  | 0.0                       | -0.3                       | -5.2     | -3.6       |   | 1.3 | -0.5            | 1.8                  | 0.0                       | -0.3                       | 1.9      | 3.4        |

|     |                     | IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj | IWI   | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP     | IWI<br>adj |
|-----|---------------------|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|-------|-----------------|----------------------|---------------------------|----------------------------|---------|------------|
|     | Country/Year        | aver | age gro         | wth rat              | te (in pe                 | rcentage                   | e) 1991- | -1995      | avera | ige grov        | wth rat              | e (in pe                  | rcentage                   | ) 1996- | -2000      |
| 73  | Luxembourg          | 2.1  | 1.3             | 0.7                  | -0.3                      | -0.3                       | -1.5     | -1.3       | 2.9   | 1.3             | 1.6                  | -0.3                      | -0.3                       | -2.8    | -1.7       |
| 74  | Malawi              | 0.0  | 1.1             | -1.0                 | -0.1                      | -0.1                       | 0.0      | -1.3       | 0.7   | 2.6             | -1.9                 | -0.1                      | -0.1                       | -1.4    | -3.6       |
| 75  | Malaysia            | 3.7  | 2.6             | 1.1                  | -0.1                      | 0.0                        | -2.4     | -1.4       | 2.9   | 2.5             | 0.4                  | -0.1                      | -0.1                       | 0.2     | 0.4        |
| 76  | Maldives            | 4.6  | 2.5             | 2.0                  | -0.2                      | -0.3                       | 1.5      | 3.1        | 5.3   | 1.9             | 3.4                  | -0.1                      | -0.4                       | 0.0     | 2.8        |
| 77  | Mali                | 1.0  | 2.5             | -1.5                 | -0.1                      | 0.0                        | -1.8     | -3.4       | 1.2   | 2.8             | -1.6                 | -0.1                      | 0.0                        | -1.3    | -3.1       |
| 78  | Malta               | 2.2  | 1.0             | 1.2                  | -0.1                      | -0.3                       | -0.3     | 0.4        | 2.8   | 0.6             | 2.2                  | -0.1                      | -0.4                       | 0.1     | 1.7        |
| 79  | Mauritania          | 2.6  | 2.8             | -0.2                 | -0.1                      | -0.6                       | -2.9     | -3.9       | 2.3   | 2.9             | -0.6                 | -0.1                      | -0.7                       | -2.9    | -4.3       |
| 80  | Mauritius           | 2.0  | 1.4             | 0.6                  | -0.1                      | -0.3                       | 0.8      | 0.9        | 1.9   | 1.0             | 0.9                  | -0.1                      | -0.4                       | -1.6    | -1.2       |
| 81  | Mexico              | 2.3  | 1.8             | 0.5                  | -0.1                      | -0.3                       | 1.3      | 1.4        | 2.2   | 1.6             | 0.6                  | -0.1                      | -0.3                       | -2.4    | -2.2       |
| 82  | Mongolia            | -0.5 | 1.0             | -1.5                 | 0.0                       | -0.1                       | 4.3      | 2.7        | 0.6   | 0.9             | -0.3                 | 0.0                       | -0.1                       | -0.9    | -1.3       |
| 83  | Morocco             | 3.0  | 1.7             | 1.3                  | -0.1                      | -0.2                       | 2.2      | 3.2        | 2.0   | 1.3             | 0.7                  | ′ –0.1                    | -0.3                       | -0.8    | -0.5       |
| 84  | Mozambique          | 0.1  | 3.3             | -3.2                 | 0.0                       | 0.0                        | 0.0      | -3.3       | 0.2   | 2.7             | -2.5                 | 0.0                       | 0.0                        | -4.1    | -6.6       |
| 85  | Myanmar             | -0.6 | 1.4             | -2.0                 | 0.0                       | 0.0                        | -0.9     | -2.9       | -0.6  | 1.3             | -1.9                 | 0.0                       | -0.1                       | -2.3    | -4.3       |
| 86  | Namibia             | 1.4  | 3.1             | -1.7                 | 0.0                       | -0.1                       | -2.3     | -4.1       | 1.2   | 2.8             | -1.6                 | 0.0                       | -0.1                       | -1.2    | -2.8       |
| 87  | Nepal               | -0.3 | 2.5             | -2.8                 | -0.1                      | -0.1                       | -2.7     | -5.6       | 0.3   | 2.5             | -2.2                 | -0.1                      | -0.1                       | -2.2    | -4.6       |
| 88  | Netherlands         | 1.4  | 0.7             | 0.7                  | -0.2                      | -0.2                       | -0.8     | -0.4       | 1.5   | 0.6             | 1.0                  | -0.2                      | -0.2                       | -1.4    | -0.8       |
| 89  | New Zealand         | 1.5  | 1.6             | -0.1                 | 0.0                       | -0.1                       | -1.1     | -1.3       | 1.6   | 1.0             | 0.7                  | 0.0                       | -0.2                       | -1.3    | -0.8       |
| 90  | Nicaragua           | 0.7  | 2.4             | -1.7                 | -0.1                      | -0.2                       | 0.3      | -1.7       | 1.6   | 1.8             | -0.2                 | -0.1                      | -0.3                       | -1.9    | -2.5       |
| 91  | Niger               | 1.6  | 3.3             | -1.8                 | -0.1                      | -0.1                       | 0.1      | -2.0       | 2.1   | 3.5             | -1.4                 | -0.1                      | -0.1                       | -2.1    | -3.8       |
| 92  | Nigeria             | 0.2  | 2.4             | -2.2                 | -0.1                      | 1.6                        | 0.2      | -0.5       | 0.3   | 2.4             | -2.1                 | -0.1                      | 1.9                        | -2.8    | -3.1       |
| 93  | Norway              | 0.5  | 0.5             | -0.1                 | -0.2                      | 0.5                        | -2.7     | -2.5       | 0.8   | 0.6             | 0.2                  | -0.2                      | 0.4                        | -2.0    | -1.6       |
| 94  | Pakistan            | 3.1  | 2.6             | 0.5                  | -0.1                      | -0.3                       | -0.7     | -0.6       | 2.8   | 2.6             | 0.3                  | -0.1                      | -0.3                       | -0.2    | -0.4       |
| 95  | Panama              | 2.8  | 2.1             | 0.7                  | -0.1                      | -0.7                       | -1.4     | -1.5       | 2.1   | 2.0             | 0.1                  | -0.1                      | -0.7                       | -0.9    | -1.6       |
| 96  | Papua New Guinea    | -0.3 | 2.6             | -2.9                 | 0.0                       | 0.0                        | -5.6     | -8.5       | -0.3  | 2.7             | -3.0                 | 0.0                       | 0.0                        | 2.1     | -0.9       |
| 97  | Paraguay            | 0.8  | 2.5             | -1.7                 | 0.0                       | -0.1                       | -0.5     | -2.3       | 1.1   | 2.2             | -1.1                 | 0.0                       | -0.1                       | 2.0     | 0.7        |
| 98  | Peru                | 0.6  | 1.9             | -1.3                 | 0.0                       | -0.1                       | -3.6     | -5.0       | 1.1   | 1.7             | -0.6                 | 0.0                       | -0.1                       | -0.2    | -0.9       |
| 99  | Philippines         | 2.6  | 2.4             | 0.2                  | -0.2                      | -0.4                       | 0.7      | 0.4        | 3.0   | 2.2             | 0.7                  | ′                         | -0.5                       | -0.1    | 0.0        |
| 100 | Poland              | 1.1  | 0.2             | 0.9                  | 0.0                       | -0.1                       | -0.3     | 0.4        | 1.3   | 0.0             | 1.4                  | 0.0                       | -0.1                       | -1.2    | -0.1       |
| 101 | Portugal            | 1.5  | 0.4             | 1.1                  | -0.1                      | -0.1                       | 0.6      | 1.4        | 1.9   | 0.4             | 1.5                  | -0.1                      | -0.2                       | -1.0    | 0.2        |
| 102 | Qatar               | 0.2  | 1.1             | -1.0                 | -0.1                      | 3.4                        | -1.0     | 1.4        | 0.6   | 3.3             | -2.8                 | -0.1                      | 4.1                        | -7.2    | -5.9       |
| 103 | Republic of Korea   | 4.0  | 0.8             | 3.2                  | -0.2                      | -0.4                       | -0.1     | 2.6        | 3.0   | 0.6             | 2.4                  | -0.1                      | -0.4                       | 0.3     | 2.2        |
| 104 | Republic of Moldova | -0.6 | -0.1            | -0.5                 | 0.0                       | -0.5                       | 22.4     | 21.3       | -1.2  | -1.1            | -0.1                 | 0.0                       | -0.3                       | 2.5     | 2.2        |
| 105 | Romania             | 0.6  | -0.5            | 1.1                  | 0.0                       | -0.2                       | 3.1      | 3.9        | 0.7   | -0.4            | 1.2                  | 0.0                       | -0.2                       | 1.3     | 2.2        |
| 106 | Russian Federation* | -0.2 | 0.0             | -0.2                 | 0.0                       | 0.1                        | 8.2      | 8.1        | 0.0   | -0.3            | 0.3                  | 0.0                       | 0.2                        | -0.3    | 0.1        |
| 107 | Rwanda              | -3.5 | -4.7            | 1.2                  | -0.2                      | -0.3                       | 14.7     | 15.4       | 7.4   | 7.8             | -0.4                 | -0.2                      | -0.3                       | -6.5    | -7.4       |
| 108 | Saudi Arabia        | 1.6  | 2.8             | -1.2                 | -0.1                      | 2.9                        | 0.2      | 1.9        | 0.4   | 1.6             | -1.2                 | -0.1                      | 3.3                        | -1.0    | 1,1        |

|   | IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj |   | IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj |
|---|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|---|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|
|   | av   | erage gr        | owth rate            | e (in perc                | entage) 2                  | 001-2005 |            | ĺ | a    | /erage gr       | owth rate            | (in perc                  | entage) 2                  | 006-2010 | )          |
| 1 | 3.0  | 1.0             | 2.0                  | -0.2                      | -0.3                       | -0.4     | 1.0        | ľ | 3.9  | 2.1             | 1.8                  | -0.2                      | -0.3                       | 1.9      | 3.1        |
|   | 1.2  | 2.7             | -1.5                 | -0.1                      | -0.2                       | -1.0     | -2.8       |   | 2.2  | 3.0             | -0.8                 | -0.1                      | -0.2                       | -2.9     | -4.0       |
|   | 1.9  | 2.2             | -0.3                 | -0.1                      | -0.1                       | -1.6     | -2.0       |   | 2.3  | 1.7             | 0.6                  | -0.1                      | -0.2                       | -1.0     | -0.7       |
|   | 4.9  | 1.6             | 3.3                  | -0.1                      | -0.8                       | 4.4      | 6.8        |   | 5.7  | 1.4             | 4.3                  | -0.1                      | -0.8                       | 3.6      | 7.0        |
|   | 1.1  | 3.1             | -2.0                 | -0.1                      | 0.0                        | -3.7     | -5.9       |   | 1.7  | 3.1             | -1.4                 | -0.1                      | 0.0                        | -1.6     | -3.1       |
|   | 2.2  | 0.6             | 1.6                  | -0.1                      | -0.4                       | 2.7      | 3.8        |   | 2.0  | 0.4             | 1.6                  | -0.1                      | -0.6                       | 0.4      | 1.2        |
|   | 4.1  | 2.9             | 1.2                  | -0.1                      | -0.6                       | 1.2      | 1.6        |   | 3.4  | 2.6             | 0.8                  | -0.1                      | -0.4                       | 2.0      | 2.3        |
|   | 2.7  | 1.0             | 1.7                  | -0.1                      | -0.5                       | 0.9      | 2.0        |   | 2.7  | 0.7             | 2.0                  | -0.1                      | -0.5                       | 0.0      | 1.4        |
|   | 2.7  | 1.3             | 1.4                  | -0.1                      | -0.3                       | 1.3      | 2.3        |   | 2.5  | 1.3             | 1.2                  | -0.1                      | -0.3                       | 1.4      | 2.1        |
|   | -1.1 | 1.1             | -2.2                 | 0.0                       | -0.1                       | -4.1     | -6.4       |   | 0.7  | 1.6             | -0.9                 | 0.0                       | -0.1                       | 0.3      | -0.7       |
|   | 2.4  | 1.1             | 1.3                  | -0.1                      | -0.3                       | -1.3     | -0.3       |   | 2.2  | 1.0             | 1.2                  | -0.1                      | -0.4                       | -0.3     | 0.5        |
|   | 0.4  | 2.7             | -2.3                 | 0.0                       | -0.1                       | -3.5     | -5.9       |   | 0.5  | 2.4             | -1.9                 | -0.1                      | -0.1                       | -2.0     | -4.0       |
|   | -0.3 | 0.6             | -0.9                 | 0.0                       | -0.1                       | -2.4     | -3.5       |   | 0.2  | 0.7             | -0.5                 | -0.1                      | -0.2                       | 1.8      | 1.2        |
|   | 1.8  | 1.9             | -0.1                 | 0.0                       | -0.1                       | -2.5     | -2.8       |   | 2.1  | 1.9             | 0.2                  | 0.0                       | -0.2                       | -0.3     | -0.3       |
|   | 1.0  | 2.3             | -1.2                 | -0.1                      | -0.1                       | -0.7     | -2.2       |   | 1.7  | 1.9             | -0.2                 | -0.1                      | -0.2                       | -0.9     | -1.4       |
|   | 1.5  | 0.6             | 0.9                  | -0.2                      | -0.2                       | 1.0      | 1.5        |   | 1.5  | 0.4             | 1.1                  | -0.2                      | -0.2                       | 0.7      | 1.4        |
|   | 2.1  | 1.4             | 0.7                  | 0.0                       | -0.2                       | -0.7     | -0.2       |   | 1.8  | 1.1             | 0.7                  | 0.0                       | -0.2                       | 1.0      | 1.5        |
|   | 1.1  | 1.3             | -0.3                 | -0.1                      | -0.4                       | -1.2     | -2.0       |   | 1.0  | 1.3             | -0.4                 | -0.1                      | -0.5                       | -1.4     | -2.2       |
|   | 2.4  | 3.5             | -1.1                 | -0.1                      | -0.1                       | -3.0     | -4.3       |   | 3.6  | 3.6             | 0.0                  | -0.1                      | -0.1                       | -1.1     | -1.3       |
|   | 0.8  | 2.5             | -1.7                 | -0.1                      | 2.2                        | -8.7     | -8.3       |   | 1.2  | 2.5             | -1.4                 | -0.1                      | 2.5                        | -0.3     | 0.7        |
|   | 1.3  | 0.6             | 0.7                  | -0.2                      | 0.3                        | -0.4     | 0.5        |   | 1.5  | 1.1             | 0.4                  | -0.2                      | 0.2                        | 1.2      | 1.7        |
|   | 3.0  | 1.9             | 1.1                  | -0.1                      | -0.3                       | -2.4     | -1.7       |   | 2.6  | 1.8             | 0.8                  | -0.1                      | -0.3                       | -0.7     | -0.3       |
|   | 2.4  | 1.8             | 0.6                  | -0.1                      | -0.7                       | -1.4     | -1.6       |   | 2.8  | 1.7             | 1.1                  | -0.1                      | -0.8                       | -1.7     | -1.5       |
|   | -0.2 | 2.5             | -2.7                 | 0.0                       | -0.1                       | 1.5      | -1.3       |   | -0.1 | 2.4             | -2.5                 | 0.0                       | -0.1                       | -2.5     | -5.2       |
|   | 0.8  | 2.0             | -1.2                 | 0.0                       | -0.2                       | -1.5     | -2.9       |   | 1.2  | 1.8             | -0.6                 | 0.0                       | -0.2                       | -3.5     | -4.3       |
|   | 0.5  | 1.3             | -0.8                 | 0.0                       | -0.1                       | -3.0     | -3.9       |   | 1.4  | 1.1             | 0.3                  | 0.0                       | -0.1                       | -2.8     | -2.7       |
|   | 2.2  | 2.0             | 0.1                  | -0.1                      | -0.5                       | -1.8     | -2.3       |   | 2.7  | 1.7             | 0.9                  | -0.1                      | -0.4                       | -1.9     | -1.5       |
|   | 0.7  | -0.1            | 0.7                  | 0.0                       | -0.2                       | -0.4     | 0.1        |   | 1.5  | 0.1             | 1.5                  | 0.0                       | -0.2                       | -0.8     | 0.4        |
|   | 1.1  | 0.4             | 0.7                  | -0.1                      | -0.2                       | 2.0      | 2.4        |   | 1.0  | 0.2             | 0.8                  | -0.1                      | -0.2                       | 1.2      | 1.7        |
|   | 2.6  | 6.9             | -4.3                 | -0.1                      | 4.7                        | -0.9     | -0.6       |   | 8.2  | 16.5            | -8.3                 | -0.1                      | 4.2                        | -4.0     | -8.3       |
|   | 2.6  | 0.5             | 2.1                  | -0.1                      | -0.4                       | -0.7     | 0.9        |   | 2.2  | 0.5             | 1.7                  | -0.1                      | -0.4                       | -2.5     | -1.3       |
|   | -1.4 | -1.7            | 0.3                  | 0.0                       | -0.3                       | -7.8     | -7.8       |   | -1.5 | -1.1            | -0.4                 | 0.0                       | -0.4                       | -5.8     | -6.7       |
|   | -0.2 | -0.4            | 0.1                  | 0.0                       | -0.2                       | -4.4     | -4.5       |   | 1.5  | -0.3            | 1.7                  | 0.0                       | -0.2                       | 0.4      | 1.8        |
|   | -0.1 | -0.4            | 0.4                  | 0.0                       | 0.2                        | -5.6     | -5.1       |   | 0.0  | -0.1            | 0.1                  | 0.0                       | 0.2                        | -2.3     | -2.0       |
|   | 3.0  | 2.6             | 0.4                  | -0.2                      | -0.2                       | -3.4     | -3.4       |   | 4.3  | 2.9             | 1.4                  | -0.2                      | -0.2                       | -0.4     | 0.6        |
|   | 1.4  | 3.7             | -2.3                 | -0.1                      | 3.8                        | -0.9     | 0.5        |   | 1,83 | 2,69            | -0,86                | -0,06                     | 4,12                       | 2,01     | 5,22       |

|     |                                    | IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP     | IWI<br>adj | IWI   | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP     | IWI<br>adj |
|-----|------------------------------------|------|-----------------|----------------------|---------------------------|----------------------------|---------|------------|-------|-----------------|----------------------|---------------------------|----------------------------|---------|------------|
|     | Country/Year                       | aver | age gro         | wth ra               | te (in pe                 | ercentage                  | e) 1991 | -1995      | avera | age grov        | vth rat              | e (in pe                  | rcentage                   | ) 1996- | 2000       |
| 109 | Senegal                            | 1.1  | 2.9             | -1.9                 | -0.1                      | -0.2                       | 0.9     | -1.3       | 1.6   | 2.6             | -1.0                 | -0.1                      | -0.3                       | -1.6    | -3.0       |
| 110 | Serbia                             | 1.9  | 1.3             | 0.7                  | 0.0                       | 0.0                        | 17.3    | 18.0       | 0.2   | -0.1            | 0.3                  | 0.0                       | 0.0                        | -1.2    | -0.9       |
| 111 | Sierra Leone                       | -0.6 | -0.4            | -0.2                 | -0.1                      | -0.2                       | 5.2     | 4.7        | 0.4   | 1.2             | -0.8                 | -0.1                      | -0.2                       | 10.7    | 9.5        |
| 112 | Singapore                          | 4.5  | 2.9             | 1.6                  | 0.0                       | -0.9                       | -4.4    | -3.7       | 4.9   | 2.4             | 2.5                  | 0.0                       | -1.1                       | -1.6    | -0.2       |
| 113 | Slovakia**                         | 1.1  | 0.3             | 0.8                  | 0.0                       | -0.2                       | -3.4    | -2.8       | 1.2   | 0.1             | 1.0                  | 0.0                       | -0.2                       | -2.7    | -1.9       |
| 114 | Slovenia*                          | 2.3  | 0.4             | 1.9                  | 0.0                       | -0.1                       | 0.4     | 2.1        | 1.5   | 0.2             | 1.3                  | 0.0                       | -0.2                       | -1.7    | -0.6       |
| 115 | South Africa                       | 2.9  | 2.4             | 0.5                  | -0.1                      | -0.2                       | 2.2     | 2.4        | 0.7   | 1.6             | -0.8                 | -0.1                      | -0.2                       | -1.5    | -2.6       |
| 116 | Spain                              | 2.9  | 0.3             | 2.6                  | -0.2                      | -0.1                       | 1.3     | 3.6        | 2.7   | 0.4             | 2.2                  | -0.2                      | -0.2                       | -0.9    | 1.0        |
| 117 | Sri Lanka                          | 2.1  | 1.0             | 1.0                  | -0.1                      | -0.2                       | -2.1    | -1.3       | 2.7   | 0.6             | 2.2                  | -0.1                      | -0.3                       | -1.1    | 0.7        |
| 118 | Sudan (former)                     | 0.0  | 2.6             | -2.6                 | -0.1                      | 0.5                        | -1.8    | -4.1       | 0.5   | 2.6             | -2.1                 | -0.1                      | 0.6                        | -0.6    | -2.2       |
| 119 | Swaziland                          | 3.4  | 2.2             | 1.2                  | -0.1                      | -0.1                       | 0.7     | 1.6        | 2.1   | 2.0             | 0.1                  | -0.1                      | -0.1                       | -1.3    | -1.3       |
| 120 | Sweden                             | 0.6  | 0.6             | 0.0                  | -0.2                      | -0.1                       | 0.2     | -0.1       | 0.7   | 0.1             | 0.7                  | -0.2                      | -0.1                       | -2.0    | -1.7       |
| 121 | Switzerland                        | 0.9  | 1.0             | -0.1                 | -0.2                      | -0.1                       | 1.0     | 0.6        | 0.7   | 0.4             | 0.3                  | -0.2                      | -0.1                       | -0.5    | -0.5       |
| 122 | Syrian Arab Republic               | 2.5  | 2.8             | -0.3                 | -0.1                      | -0.1                       | -4.4    | -4.9       | 2.2   | 2.4             | -0.2                 | -0.1                      | -0.2                       | -1.0    | -1.6       |
| 123 | Tajikistan*                        | 0.1  | 1.6             | -1.5                 | -0.1                      | -0.6                       | 26.5    | 24.2       | -0.2  | 1.3             | -1.6                 | -0.1                      | -0.8                       | 2.1     | -0.5       |
| 124 | Thailand                           | 3.2  | 0.9             | 2.3                  | -0.1                      | -0.4                       | -5.4    | -3.6       | 2.0   | 1.1             | 0.8                  | -0.1                      | -0.5                       | 1.2     | 1.4        |
| 125 | Тодо                               | 0.8  | 2.2             | -1.4                 | -0.1                      | -0.1                       | 1.6     | -0.1       | 2.5   | 3.2             | -0.7                 | -0.1                      | -0.2                       | -0.8    | -1.8       |
| 126 | Trinidad and Tobago                | 0.1  | 0.7             | -0.6                 | -0.1                      | 0.3                        | -0.1    | -0.5       | 0.8   | 0.5             | 0.3                  | -0.1                      | 0.4                        | -4.9    | -4.3       |
| 127 | Tunisia                            | 2.6  | 1.7             | 0.9                  | -0.1                      | -0.1                       | -1.1    | -0.4       | 2.4   | 1.1             | 1.2                  | -0.1                      | -0.1                       | -2.7    | -1.7       |
| 128 | Turkey                             | 2.9  | 1.7             | 1.2                  | -0.1                      | -0.2                       | 1.3     | 2.2        | 2.1   | 1.6             | 0.5                  | -0.1                      | -0.2                       | -0.4    | -0.2       |
| 129 | Uganda                             | 1.8  | 3.3             | -1.6                 | -0.3                      | -0.1                       | -4.2    | -6.1       | 2.6   | 3.1             | -0.5                 | -0.2                      | -0.1                       | -2.3    | -3.2       |
| 130 | Ukraine*                           | 0.5  | -0.3            | 0.7                  | 0.0                       | -0.3                       | 19.3    | 19.6       | -0.9  | -0.9            | 0.0                  | 0.0                       | -0.2                       | 1.6     | 1.3        |
| 131 | United Arab Emirates               | 2.3  | 5.4             | -3.1                 | -0.1                      | 3.0                        | 1.6     | 1.5        | 2.6   | 5.3             | -2.6                 | -0.1                      | 3.3                        | -1.6    | -1.0       |
| 132 | United Kingdom                     | 0.3  | 0.3             | 0.1                  | -0.2                      | -0.1                       | -1.1    | -1.2       | 1.4   | 0.3             | 1.1                  | -0.2                      | -0.1                       | -1.9    | -1.1       |
| 133 | United Republic of Tanzania        | 0.3  | 3.3             | -3.0                 | -0.1                      | -0.1                       | -0.4    | -3.6       | 0.0   | 2.6             | -2.6                 | -0.1                      | -0.1                       | -3.0    | -5.8       |
| 134 | United States of America           | 1.2  | 1.0             | 0.2                  | 0.0                       | -0.2                       | -0.5    | -0.5       | 1.9   | 1.2             | 0.7                  | 0.0                       | -0.2                       | -1.1    | -0.6       |
| 135 | Uruguay                            | 1.0  | 0.7             | 0.3                  | -0.1                      | -0.2                       | -2.4    | -2.3       | 1.7   | 0.6             | 1.1                  | -0.1                      | -0.2                       | 0.4     | 1.2        |
| 136 | Venezuela (Bolivarian Republic of) | 0.6  | 2.3             | -1.7                 | 0.0                       | 1.8                        | -1.3    | -1.2       | 0.8   | 2.0             | -1.2                 | 0.0                       | 2.2                        | 1.2     | 2.1        |
| 137 | Vietnam                            | 2.5  | 2.0             | 0.6                  | -0.1                      | 0.0                        | -0.9    | -0.5       | 3.7   | 1.3             | 2.5                  | -0.1                      | -0.1                       | 1.0     | 3.2        |
| 138 | Yemen                              | 3.2  | 4.9             | -1.7                 | -0.1                      | 0.6                        | -1.6    | -2.8       | 3.5   | 3.2             | 0.3                  | -0.1                      | 0.5                        | -3.3    | -2.5       |
| 139 | Zambia                             | -0.1 | 2.6             | -2.6                 | 0.0                       | 0.0                        | 1.6     | -1.1       | 0.1   | 2.7             | -2.6                 | 0.0                       | 0.0                        | 0.4     | -2.2       |
| 140 | Zimbabwe                           | 0.2  | 2.2             | -2.0                 | -0.1                      | -0.2                       | 0.9     | -1.4       | -0.1  | 1.4             | -1.5                 | -0.1                      | -0.2                       | 1.4     | -0.4       |

| IWI  | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama-<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj | I | WI   | Popu-<br>lation | IWI<br>per<br>capita | Carbon<br>dama–<br>ges*** | Oil<br>capital<br>gains*** | TFP      | IWI<br>adj |
|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|---|------|-----------------|----------------------|---------------------------|----------------------------|----------|------------|
| av   | erage gr        | owth rate            | e (in perc                | entage) 20                 | 001–2005 |            |   | av   | erage gr        | owth rate            | e (in perce               | entage) 20                 | 006-2010 |            |
| 1.6  | 2.7             | -1.1                 | -0.1                      | -0.4                       | -1.4     | -3.0       |   | 2.5  | 2.7             | -0.2                 | -0.1                      | -0.4                       | 0.7      | -0.1       |
| 0.3  | -0.6            | 0.8                  | 0.0                       | 0.0                        | -3.1     | -2.3       |   | 1.1  | 0.0             | 1.1                  | 0.0                       | -0.2                       | 1.1      | 1.9        |
| 2.9  | 4.5             | -1.6                 | -0.1                      | -0.3                       | -9.3     | -11.3      |   | 1.6  | 2.6             | -1.0                 | -0.1                      | -0.3                       | -1.9     | -3.3       |
| 3.5  | 1.7             | 1.8                  | 0.0                       | -1.1                       | -2.4     | -1.7       |   | 5.4  | 3.6             | 1.8                  | 0.0                       | -1.4                       | -3.2     | -2.7       |
| 1.5  | 0.0             | 1.5                  | 0.0                       | -0.2                       | -3.0     | -1.8       |   | 1.3  | 0.2             | 1.2                  | 0.0                       | -0.2                       | -2.3     | -1.4       |
| 1.7  | 0.2             | 1.5                  | 0.0                       | -0.2                       | -0.2     | 1.1        |   | 1.1  | 0.3             | 0.8                  | 0.0                       | -0.2                       | 1.5      | 2.1        |
| 1.9  | 1.3             | 0.6                  | -0.1                      | -0.2                       | -1.6     | -1.3       |   | 1.1  | 1.0             | 0.2                  | -0.1                      | -0.2                       | 0.5      | 0.3        |
| 2.7  | 1.5             | 1.2                  | -0.1                      | -0.2                       | 0.6      | 1.5        |   | 2.8  | 1.2             | 1.6                  | -0.1                      | -0.2                       | 2.4      | 3.7        |
| 1.3  | 1.1             | 0.2                  | -0.1                      | -0.3                       | -0.4     | -0.6       |   | 2.4  | 1.0             | 1.4                  | -0.1                      | -0.3                       | -1.6     | -0.6       |
| 1.3  | 2.4             | -1.0                 | -0.1                      | 0.6                        | 2.9      | 2.3        |   | 1.8  | 2.5             | -0.8                 | -0.1                      | 0.5                        | 0.0      | -0.3       |
| 1.2  | 0.8             | 0.4                  | -0.1                      | -0.1                       | -3.0     | -2.8       |   | 1.0  | 1.4             | -0.4                 | -0.1                      | -0.1                       | -2.8     | -3.4       |
| 1.3  | 0.4             | 0.9                  | -0.2                      | -0.1                       | -0.6     | 0.1        |   | 1.7  | 0.8             | 0.9                  | -0.2                      | -0.1                       | 1.1      | 1.7        |
| 1.3  | 0.7             | 0.6                  | -0.2                      | -0.1                       | 0.6      | 0.9        |   | 1.6  | 0.7             | 1.0                  | -0.2                      | -0.1                       | 0.0      | 0.7        |
| 2.0  | 2.9             | -1.0                 | -0.1                      | -0.3                       | -1.9     | -3.3       |   | 2.3  | 2.0             | 0.3                  | -0.1                      | -0.5                       | -1.0     | -1.3       |
| 0.3  | 0.9             | -0.6                 | -0.1                      | -1.3                       | -6.9     | -8.9       |   | 0.6  | 1.3             | -0.7                 | -0.1                      | -1.3                       | -4.6     | -6.7       |
| 1.7  | 1.1             | 0.6                  | -0.1                      | -0.6                       | -3.0     | -3.0       |   | 2.0  | 0.7             | 1.3                  | -0.1                      | -0.7                       | -1.7     | -1.3       |
| 2.3  | 2.4             | -0.2                 | -0.1                      | -0.4                       | -0.6     | -1.2       |   | 2.0  | 2.2             | -0.2                 | -0.1                      | -0.4                       | -1.6     | -2.3       |
| 0.9  | 0.4             | 0.5                  | -0.1                      | 0.3                        | -5.7     | -5.0       | - | -0.7 | 0.4             | -1.1                 | -0.1                      | 0.2                        | -3.7     | -4.7       |
| 2.0  | 0.9             | 1.0                  | -0.1                      | -0.2                       | -1.5     | -0.8       |   | 2.6  | 1.1             | 1.4                  | -0.1                      | -0.2                       | -1.7     | -0.5       |
| 1.9  | 1.4             | 0.5                  | -0.1                      | -0.2                       | -1.1     | -0.9       |   | 2.8  | 1.3             | 1.5                  | -0.1                      | -0.2                       | 0.8      | 2.0        |
| 2.9  | 3.3             | -0.4                 | -0.2                      | -0.2                       | -1.9     | -2.8       |   | 4.4  | 3.3             | 1.1                  | -0.2                      | -0.3                       | 0.4      | 1.1        |
| -0.4 | -0.8            | 0.5                  | 0.0                       | -0.2                       | -7.2     | -7.0       | - | -0.1 | -0.6            | 0.5                  | 0.0                       | -0.2                       | -0.8     | -0.6       |
| 3.7  | 6.1             | -2.4                 | -0.1                      | 3.6                        | -1.5     | -0.4       |   | 8.4  | 13.1            | -4.7                 | -0.1                      | 3.2                        | 2.8      | 1.2        |
| 1.9  | 0.4             | 1.4                  | -0.2                      | -0.1                       | -0.6     | 0.6        |   | 1.5  | 0.6             | 0.9                  | -0.2                      | -0.1                       | 1.7      | 2.3        |
| 0.5  | 2.7             | -2.1                 | -0.1                      | -0.1                       | -4.5     | -6.9       |   | 1.3  | 2.9             | -1.7                 | -0.1                      | -0.2                       | -5.0     | -7.0       |
| 2.0  | 1.0             | 1.0                  | 0.0                       | -0.2                       | 1.1      | 1.8        |   | 1.5  | 0.9             | 0.6                  | 0.0                       | -0.2                       | 1.5      | 1.8        |
| 0.2  | 0.0             | 0.2                  | -0.1                      | -0.2                       | 0.7      | 0.6        |   | 1.6  | 0.3             | 1.3                  | -0.1                      | -0.3                       | -2.8     | -1.9       |
| 0.8  | 1.8             | -1.0                 | 0.0                       | 2.5                        | -1.2     | 0.2        |   | 1.1  | 1.7             | -0.6                 | 0.0                       | 2.9                        | -0.9     | 1.4        |
| 3.3  | 1.1             | 2.2                  | -0.1                      | -0.3                       | -0.2     | 1.6        |   | 3.6  | 1.1             | 2.5                  | -0.1                      | -0.5                       | 1.2      | 3.1        |
| 3.0  | 3.1             | -0.1                 | -0.1                      | 0.3                        | -1.2     | -1.1       |   | 2.9  | 3.1             | -0.2                 | -0.1                      | 0.2                        | -2.0     | -2.1       |
| 0.3  | 2.4             | -2.1                 | 0.0                       | 0.0                        | 0.1      | -2.1       |   | 0.5  | 2.7             | -2.2                 | 0.0                       | -0.1                       | -0.3     | -2.6       |
| 0.1  | 0.1             | 0.0                  | -0.1                      | -0.2                       | 6.3      | 6.0        | - | -1.0 | 0.0             | -1.0                 | -0.1                      | -0.1                       | -2.0     | -3.3       |

## Population

|    |                                  |           |           |           | Pc<br>in th | opulation<br>lousands | with resp | Pc<br>per<br>ect to the | pulation<br>centage<br>base yea | change<br>change<br>ar, 1990 |
|----|----------------------------------|-----------|-----------|-----------|-------------|-----------------------|-----------|-------------------------|---------------------------------|------------------------------|
|    | Country/Year                     | 1990      | 1995      | 2000      | 2005        | 2010                  | 1995      | 2000                    | 2005                            | 2010                         |
| 1  | Afghanistan                      | 13,032    | 19,790    | 22,856    | 27,615      | 31,412                | 52        | 75                      | 112                             | 141                          |
| 2  | Albania                          | 3,289     | 3,141     | 3,072     | 3,142       | 3,204                 | -5        | -7                      | -4                              | -3                           |
| 3  | Algeria                          | 25,299    | 28,292    | 30,534    | 32,888      | 35,468                | 12        | 21                      | 30                              | 40                           |
| 4  | Argentina                        | 32,642    | 34,855    | 36,931    | 38,681      | 40,412                | 7         | 13                      | 18                              | 24                           |
| 5  | Armenia                          | 3,545     | 3,223     | 3,076     | 3,066       | 3,092                 | -9        | -13                     | -14                             | -13                          |
| 6  | Australia                        | 17,096    | 18,118    | 19,164    | 20,404      | 22,268                | 6         | 12                      | 19                              | 30                           |
| 7  | Austria                          | 7,671     | 7,936     | 8,005     | 8,232       | 8,394                 | 3         | 4                       | 7                               | 9                            |
| 8  | Bahrain                          | 493       | 559       | 638       | 725         | 1,262                 | 13        | 29                      | 47                              | 156                          |
| 9  | Bangladesh                       | 105,256   | 117,487   | 129,592   | 140,588     | 148,692               | 12        | 23                      | 34                              | 41                           |
| 10 | Barbados                         | 260       | 263       | 268       | 271         | 273                   | 2         | 3                       | 4                               | 5                            |
| 11 | Belgium                          | 9,949     | 10,080    | 10,176    | 10,414      | 10,712                | 1         | 2                       | 5                               | 8                            |
| 12 | Belize                           | 190       | 220       | 251       | 281         | 312                   | 16        | 32                      | 48                              | 64                           |
| 13 | Benin                            | 4,773     | 5,651     | 6,518     | 7,634       | 8,850                 | 18        | 37                      | 60                              | 85                           |
| 14 | Bolivia (Plurinational State of) | 6,658     | 7,471     | 8,307     | 9,147       | 9,930                 | 12        | 25                      | 37                              | 49                           |
| 15 | Botswana                         | 1,382     | 1,586     | 1,758     | 1,876       | 2,007                 | 15        | 27                      | 36                              | 45                           |
| 16 | Brazil                           | 149,650   | 161,848   | 174,425   | 185,987     | 194,946               | 8         | 17                      | 24                              | 30                           |
| 17 | Bulgaria                         | 8,819     | 8,357     | 8,006     | 7,739       | 7,494                 | -5        | -9                      | -12                             | -15                          |
| 18 | Burundi                          | 5,602     | 6,087     | 6,374     | 7,251       | 8,383                 | 9         | 14                      | 29                              | 50                           |
| 19 | Cambodia                         | 9,532     | 11,169    | 12,447    | 13,358      | 14,138                | 17        | 31                      | 40                              | 48                           |
| 20 | Cameroon                         | 12,181    | 13,940    | 15,678    | 17,554      | 19,599                | 14        | 29                      | 44                              | 61                           |
| 21 | Canada                           | 27,701    | 29,302    | 30,667    | 32,283      | 34,017                | 6         | 11                      | 17                              | 23                           |
| 22 | Central African Republic         | 2,935     | 3,328     | 3,702     | 4,018       | 4,401                 | 13        | 26                      | 37                              | 50                           |
| 23 | Chile                            | 13,188    | 14,409    | 15,420    | 16,302      | 17,114                | 9         | 17                      | 24                              | 30                           |
| 24 | China                            | 1,145,195 | 1,213,987 | 1,269,117 | 1,307,593   | 1,341,335             | 6         | 11                      | 14                              | 17                           |
| 25 | Colombia                         | 33,203    | 36,453    | 39,764    | 43,041      | 46,295                | 10        | 20                      | 30                              | 39                           |
| 26 | Congo                            | 2,389     | 2,733     | 3,136     | 3,533       | 4,043                 | 14        | 31                      | 48                              | 69                           |
| 27 | Costa Rica                       | 3,070     | 3,469     | 3,919     | 4,309       | 4,659                 | 13        | 28                      | 40                              | 52                           |
| 28 | Côte d'Ivoire                    | 12,518    | 14,677    | 16,582    | 18,021      | 19,738                | 17        | 32                      | 44                              | 58                           |
| 29 | Croatia*                         | 4,548     | 4,669     | 4,506     | 4,442       | 4,403                 | 3         | -1                      | -2                              | -3                           |
| 30 | Cuba                             | 10,570    | 10,901    | 11,104    | 11,254      | 11,258                | 3         | 5                       | 6                               | 7                            |
| 31 | Cyprus                           | 767       | 855       | 943       | 1,033       | 1,104                 | 12        | 23                      | 35                              | 44                           |
| 32 | Czech Republic**                 | 10,314    | 10,319    | 10,243    | 10,221      | 10,493                | 0         | -1                      | -1                              | 2                            |
| 33 | Democratic Republic of the Congo | 36,406    | 44,067    | 49,626    | 57,421      | 65,966                | 21        | 36                      | 58                              | 81                           |
| 34 | Denmark                          | 5,141     | 5,233     | 5,340     | 5,419       | 5,550                 | 2         | 4                       | 5                               | 8                            |
| 35 | Dominican Republic               | 7,195     | 7,916     | 8,592     | 9,264       | 9,927                 | 10        | 19                      | 29                              | 38                           |
| 36 | Ecuador                          | 10,261    | 11,385    | 12,345    | 13,426      | 14,465                | 11        | 20                      | 31                              | 41                           |

|    |                                  | Pop<br>in the |         |           |           |           |      | Population of percentage of with respect to the base year |      |      |
|----|----------------------------------|---------------|---------|-----------|-----------|-----------|------|---|------|------|
|    | Country/Year                     | 1990          | 1995    | 2000      | 2005      | 2010      | 1995 | 2000  | 2005 | 2010 |
| 37 | Egypt                            | 56,843        | 62,064  | 67,648    | 74,203    | 81,121    | 9    | 19  | 31   | 43   |
| 38 | El Salvador                      | 5,333         | 5,733   | 5,940     | 6,051     | 6,193     | 8    | 11  | 13   | 16   |
| 39 | Estonia                          | 1,568         | 1,441   | 1,371     | 1,346     | 1,341     | -8   | -13   | -14  | -14  |
| 40 | Fiji                             | 728           | 776     | 812       | 823       | 861       | 6    | 11  | 13   | 18   |
| 41 | Finland                          | 4,986         | 5,108   | 5,173     | 5,244     | 5,365     | 2    | 4   | 5    | 8    |
| 42 | France                           | 56,708        | 57,845  | 59,048    | 60,997    | 62,787    | 2    | 4   | 8    | 11   |
| 43 | Gabon                            | 929           | 1,087   | 1,235     | 1,371     | 1,505     | 17   | 33  | 48   | 62   |
| 44 | Gambia                           | 966           | 1,126   | 1,297     | 1,504     | 1,728     | 17   | 34  | 56   | 79   |
| 45 | Germany                          | 79,098        | 81,929  | 82,349    | 82,541    | 82,302    | 4    | 4   | 4    | 4    |
| 46 | Ghana                            | 14,793        | 16,997  | 19,165    | 21,640    | 24,392    | 15   | 30  | 46   | 65   |
| 47 | Greece                           | 10,161        | 10,672  | 10,987    | 11,183    | 11,359    | 5    | 8   | 10   | 12   |
| 48 | Guatemala                        | 8,923         | 10,016  | 11,237    | 12,717    | 14,389    | 12   | 26  | 43   | 61   |
| 49 | Guyana                           | 725           | 728     | 733       | 746       | 754       | 0    | 1   | 3    | 4    |
| 50 | Haiti                            | 7,125         | 7,878   | 8,645     | 9,347     | 9,993     | 11   | 21  | 31   | 40   |
| 51 | Honduras                         | 4,889         | 5,575   | 6,218     | 6,879     | 7,601     | 14   | 27  | 41   | 55   |
| 52 | Hungary                          | 10,376        | 10,331  | 10,211    | 10,087    | 9,984     | 0    | -2  | -3   | -4   |
| 53 | Iceland                          | 255           | 267     | 281       | 297       | 320       | 5    | 10  | 16   | 26   |
| 54 | India                            | 873,785       | 964,486 | 1,053,898 | 1,140,043 | 1,224,614 | 10   | 21  | 30   | 40   |
| 55 | Indonesia                        | 184,346       | 199,400 | 213,395   | 227,303   | 239,871   | 8    | 16  | 23   | 30   |
| 56 | Iran (Islamic Republic of)       | 54,871        | 59,757  | 65,342    | 69,732    | 73,974    | 9    | 19  | 27   | 35   |
| 57 | Iraq                             | 17,374        | 20,288  | 23,857    | 27,359    | 31,672    | 17   | 37  | 57   | 82   |
| 58 | Ireland                          | 3,531         | 3,611   | 3,804     | 4,158     | 4,470     | 2    | 8   | 18   | 27   |
| 59 | Israel                           | 4,500         | 5,332   | 6,015     | 6,605     | 7,418     | 18   | 34  | 47   | 65   |
| 60 | Italy                            | 56,832        | 56,968  | 56,986    | 58,671    | 60,551    | 0    | 0   | 3    | 7    |
| 61 | Jamaica                          | 2,365         | 2,462   | 2,582     | 2,682     | 2,741     | 4    | 9   | 13   | 16   |
| 62 | Japan                            | 122,251       | 124,487 | 125,720   | 126,393   | 126,536   | 2    | 3   | 3    | 4    |
| 63 | Jordan                           | 3,416         | 4,382   | 4,827     | 5,342     | 6,187     | 28   | 41  | 56   | 81   |
| 64 | Kazakhstan*                      | 16,513        | 15,926  | 14,957    | 15,172    | 16,026    | -4   | -9  | -8   | -3   |
| 65 | Kenya                            | 23,447        | 27,426  | 31,254    | 35,615    | 40,513    | 17   | 33  | 52   | 73   |
| 66 | Kuwait                           | 2,088         | 1,628   | 1,941     | 2,264     | 2,737     | -22  | -7  | 8    | 31   |
| 67 | Kyrgyzstan*                      | 4,442         | 4,592   | 4,955     | 5,042     | 5,334     | 3    | 12  | 14   | 20   |
| 68 | Lao People's Democratic Republic | 4,192         | 4,795   | 5,317     | 5,753     | 6,201     | 14   | 27  | 37   | 48   |
| 69 | Latvia                           | 2,664         | 2,492   | 2,385     | 2,306     | 2,252     | -6   | -10   | -13  | -15  |
| 70 | Lesotho                          | 1,639         | 1,795   | 1,964     | 2,066     | 2,171     | 9    | 20  | 26   | 32   |
| 71 | Liberia                          | 2,127         | 2,095   | 2,847     | 3,183     | 3,994     | -2   | 34  | 50   | 88   |
| 72 | Lithuania*                       | 3,698         | 3,629   | 3,500     | 3,416     | 3,324     | -2   | -5  | -8   | -10  |

## Population

|     | operation           |         |         |         |             |                     |            |                        |                                     |                              |
|-----|---------------------|---------|---------|---------|-------------|---------------------|------------|------------------------|-------------------------------------|------------------------------|
|     |                     |         |         |         | Pc<br>in th | pulation<br>ousands | with respe | Po<br>pe<br>ect to the | opulation<br>rcentage<br>e base yea | change<br>change<br>ar, 1990 |
|     | Country/Year        | 1990    | 1995    | 2000    | 2005        | 2010                | 1995       | 2000                   | 2005                                | 2010                         |
| 73  | Luxembourg          | 381     | 408     | 435     | 457         | 507                 | 7          | 14                     | 20                                  | 33                           |
| 74  | Malawi              | 9,381   | 9,883   | 11,229  | 12,823      | 14,901              | 5          | 20                     | 37                                  | 59                           |
| 75  | Malaysia            | 18,209  | 20,721  | 23,415  | 26,100      | 28,401              | 14         | 29                     | 43                                  | 56                           |
| 76  | Maldives            | 219     | 249     | 273     | 295         | 316                 | 13         | 24                     | 35                                  | 44                           |
| 77  | Mali                | 8,673   | 9,825   | 11,295  | 13,177      | 15,370              | 13         | 30                     | 52                                  | 77                           |
| 78  | Malta               | 368     | 387     | 397     | 409         | 417                 | 5          | 8                      | 11                                  | 13                           |
| 79  | Mauritania          | 1,996   | 2,292   | 2,643   | 3,047       | 3,460               | 15         | 32                     | 53                                  | 73                           |
| 80  | Mauritius           | 1,060   | 1,137   | 1,196   | 1,257       | 1,299               | 7          | 13                     | 19                                  | 23                           |
| 81  | Mexico              | 84,307  | 92,273  | 99,960  | 106,484     | 113,423             | 9          | 19                     | 26                                  | 35                           |
| 82  | Mongolia            | 2,193   | 2,306   | 2,411   | 2,547       | 2,756               | 5          | 10                     | 16                                  | 26                           |
| 83  | Morocco             | 24,781  | 26,929  | 28,793  | 30,392      | 31,951              | 9          | 16                     | 23                                  | 29                           |
| 84  | Mozambique          | 13,547  | 15,933  | 18,201  | 20,770      | 23,391              | 18         | 34                     | 53                                  | 73                           |
| 85  | Myanmar             | 39,268  | 42,133  | 44,958  | 46,321      | 47,963              | 7          | 14                     | 18                                  | 22                           |
| 86  | Namibia             | 1,415   | 1,651   | 1,896   | 2,080       | 2,283               | 17         | 34                     | 47                                  | 61                           |
| 87  | Nepal               | 19,081  | 21,595  | 24,401  | 27,282      | 29,959              | 13         | 28                     | 43                                  | 57                           |
| 88  | Netherlands         | 14,892  | 15,422  | 15,863  | 16,305      | 16,613              | 4          | 7                      | 9                                   | 12                           |
| 89  | New Zealand         | 3,398   | 3,675   | 3,858   | 4,134       | 4,368               | 8          | 14                     | 22                                  | 29                           |
| 90  | Nicaragua           | 4,121   | 4,637   | 5,074   | 5,424       | 5,788               | 13         | 23                     | 32                                  | 40                           |
| 91  | Niger               | 7,788   | 9,179   | 10,922  | 12,994      | 15,512              | 18         | 40                     | 67                                  | 99                           |
| 92  | Nigeria             | 97,552  | 110,015 | 123,689 | 139,823     | 158,423             | 13         | 27                     | 43                                  | 62                           |
| 93  | Norway              | 4,241   | 4,359   | 4,491   | 4,623       | 4,883               | 3          | 6                      | 9                                   | 15                           |
| 94  | Pakistan            | 111,845 | 127,347 | 144,522 | 158,645     | 173,593             | 14         | 29                     | 42                                  | 55                           |
| 95  | Panama              | 2,416   | 2,677   | 2,956   | 3,238       | 3,517               | 11         | 22                     | 34                                  | 46                           |
| 96  | Papua New Guinea    | 4,158   | 4,716   | 5,379   | 6,095       | 6,858               | 13         | 29                     | 47                                  | 65                           |
| 97  | Paraguay            | 4,244   | 4,795   | 5,344   | 5,898       | 6,455               | 13         | 26                     | 39                                  | 52                           |
| 98  | Peru                | 21,686  | 23,827  | 25,862  | 27,559      | 29,077              | 10         | 19                     | 27                                  | 34                           |
| 99  | Philippines         | 61,629  | 69,255  | 77,310  | 85,546      | 93,261              | 12         | 25                     | 39                                  | 51                           |
| 100 | Poland              | 38,056  | 38,392  | 38,302  | 38,165      | 38,277              | 1          | 1                      | 0                                   | 1                            |
| 101 | Portugal            | 9,925   | 10,125  | 10,336  | 10,544      | 10,676              | 2          | 4                      | 6                                   | 8                            |
| 102 | Qatar               | 474     | 501     | 591     | 821         | 1,759               | 6          | 25                     | 73                                  | 271                          |
| 103 | Republic of Korea   | 42,980  | 44,662  | 45,988  | 47,044      | 48,184              | 4          | 7                      | 9                                   | 12                           |
| 104 | Republic of Moldova | 4,364   | 4,339   | 4,107   | 3,767       | 3,573               | -1         | -6                     | -14                                 | -18                          |
| 105 | Romania             | 23,207  | 22,681  | 22,192  | 21,772      | 21,486              | -2         | -4                     | -6                                  | -7                           |
| 106 | Russian Federation* | 148,674 | 148,699 | 146,758 | 143,843     | 142,958             | 0          | -1                     | -3                                  | -4                           |
| 107 | Rwanda              | 7,110   | 5,570   | 8,098   | 9,202       | 10,624              | -22        | 14                     | 29                                  | 49                           |
| 108 | Saudi Arabia        | 16,139  | 18,492  | 20,045  | 24,041      | 27,448              | 15         | 24                     | 49                                  | 70                           |

|     |                                    |         |         |         | Pc<br>in th | pulation<br>ousands | with resp | Pc<br>per<br>ect to the | pulation (<br>rcentage (<br>e base yea | change<br>change<br>ar, 1990 |
|-----|------------------------------------|---------|---------|---------|-------------|---------------------|-----------|-------------------------|--|------------------------------|
|     | Country/Year                       | 1990    | 1995    | 2000    | 2005        | 2010                | 1995      | 2000                    | 2005                                   | 2010                         |
| 109 | Senegal                            | 7,242   | 8,369   | 9,506   | 10,872      | 12,434              | 16        | 31                      | 50                                     | 72                           |
| 110 | Serbia                             | 9,569   | 10,204  | 10,134  | 9,856       | 9,856               | 7         | 6                       | 3                                      | 3                            |
| 111 | Sierra Leone                       | 3,982   | 3,898   | 4,143   | 5,153       | 5,868               | -2        | 4                       | 29                                     | 47                           |
| 112 | Singapore                          | 3,017   | 3,482   | 3,919   | 4,266       | 5,086               | 15        | 30                      | 41                                     | 69                           |
| 113 | Slovakia**                         | 5,315   | 5,369   | 5,405   | 5,415       | 5,462               | 1         | 2                       | 2                                      | 3                            |
| 114 | Slovenia*                          | 1,935   | 1,966   | 1,985   | 2,002       | 2,030               | 2         | 3                       | 3                                      | 5                            |
| 115 | South Africa                       | 36,794  | 41,402  | 44,760  | 47,793      | 50,133              | 13        | 22                      | 30                                     | 36                           |
| 116 | Spain                              | 38,889  | 39,427  | 40,288  | 43,395      | 46,077              | 1         | 4                       | 12                                     | 18                           |
| 117 | Sri Lanka                          | 17,337  | 18,229  | 18,745  | 19,843      | 20,860              | 5         | 8                       | 14                                     | 20                           |
| 118 | Sudan (former)                     | 26,494  | 30,141  | 34,188  | 38,410      | 43,552              | 14        | 29                      | 45                                     | 64                           |
| 119 | Swaziland                          | 863     | 964     | 1,064   | 1,105       | 1,186               | 12        | 23                      | 28                                     | 37                           |
| 120 | Sweden                             | 8,559   | 8,827   | 8,860   | 9,029       | 9,380               | 3         | 4                       | 5                                      | 10                           |
| 121 | Switzerland                        | 6,674   | 7,019   | 7,168   | 7,415       | 7,664               | 5         | 7                       | 11                                     | 15                           |
| 122 | Syrian Arab Republic               | 12,324  | 14,171  | 15,989  | 18,484      | 20,411              | 15        | 30                      | 50                                     | 66                           |
| 123 | Tajikistan*                        | 5,419   | 5,775   | 6,173   | 6,453       | 6,879               | 7         | 14                      | 19                                     | 27                           |
| 124 | Thailand                           | 57,072  | 59,650  | 63,155  | 66,698      | 69,122              | 5         | 11                      | 17                                     | 21                           |
| 125 | Тодо                               | 3,666   | 4,085   | 4,794   | 5,408       | 6,028               | 11        | 31                      | 48                                     | 64                           |
| 126 | Trinidad and Tobago                | 1,215   | 1,261   | 1,292   | 1,315       | 1,341               | 4         | 6                       | 8                                      | 10                           |
| 127 | Tunisia                            | 8,215   | 8,936   | 9,456   | 9,912       | 10,481              | 9         | 15                      | 21                                     | 28                           |
| 128 | Turkey                             | 54,130  | 58,865  | 63,628  | 68,143      | 72,752              | 9         | 18                      | 26                                     | 34                           |
| 129 | Uganda                             | 17,700  | 20,831  | 24,213  | 28,431      | 33,425              | 18        | 37                      | 61                                     | 89                           |
| 130 | Ukraine*                           | 51,666  | 51,122  | 48,892  | 46,924      | 45,448              | -1        | -5                      | -9                                     | -12                          |
| 131 | United Arab Emirates               | 1,809   | 2,349   | 3,033   | 4,069       | 7,512               | 30        | 68                      | 125                                    | 315                          |
| 132 | United Kingdom                     | 57,214  | 57,997  | 58,874  | 60,203      | 62,036              | 1         | 3                       | 5                                      | 8                            |
| 133 | United Republic of Tanzania        | 25,479  | 29,944  | 34,038  | 38,831      | 44,841              | 18        | 34                      | 52                                     | 76                           |
| 134 | United States of America           | 253,339 | 266,324 | 282,496 | 296,820     | 310,384             | 5         | 12                      | 17                                     | 23                           |
| 135 | Uruguay                            | 3,109   | 3,223   | 3,319   | 3,323       | 3,369               | 4         | 7                       | 7                                      | 8                            |
| 136 | Venezuela (Bolivarian Republic of) | 19,685  | 22,035  | 24,348  | 26,664      | 28,980              | 12        | 24                      | 35                                     | 47                           |
| 137 | Vietnam                            | 67,102  | 74,008  | 78,758  | 83,161      | 87,848              | 10        | 17                      | 24                                     | 31                           |
| 138 | Yemen                              | 11,948  | 15,148  | 17,723  | 20,649      | 24,053              | 27        | 48                      | 73                                     | 101                          |
| 139 | Zambia                             | 7,860   | 8,919   | 10,202  | 11,462      | 13,089              | 13        | 30                      | 46                                     | 67                           |
| 140 | Zimbabwe                           | 10,469  | 11,685  | 12,509  | 12,571      | 12,571              | 12        | 19                      | 20                                     | 20                           |

## GDP

GDP level in millions of constant 2005 US\$

|    | Country/Year                     | 1990    | 1995    | 2000      | 2005      | 2010      |
|----|----------------------------------|---------|---------|-----------|-----------|-----------|
| 1  | Afghanistan                      | 5,025   | 3,977   | 3,068     | 6,840     | 9,840     |
| 2  | Albania                          | 5,322   | 4,782   | 6,136     | 8,159     | 10,488    |
| 3  | Algeria                          | 68,873  | 69,761  | 81,385    | 103,234   | 117,517   |
| 4  | Argentina                        | 109,779 | 146,181 | 166,010   | 183,196   | 253,746   |
| 5  | Armenia                          | 4,062   | 2,145   | 2,754     | 4,900     | 5,915     |
| 6  | Australia                        | 451,430 | 531,385 | 644,740   | 764,765   | 874,477   |
| 7  | Austria                          | 216,398 | 240,294 | 280,623   | 304,984   | 327,206   |
| 8  | Bahrain                          | 6,446   | 8,224   | 10,049    | 13,459    | 17,821    |
| 9  | Bangladesh                       | 27,881  | 34,316  | 44,235    | 57,628    | 77,891    |
| 10 | Barbados                         | 3,046   | 2,917   | 3,383     | 3,685     | 3,719     |
| 11 | Belgium                          | 279,844 | 302,853 | 348,623   | 377,253   | 399,921   |
| 12 | Belize                           | 518     | 640     | 857       | 1,115     | 1,259     |
| 13 | Benin                            | 2,290   | 2,814   | 3,565     | 4,358     | 5,231     |
| 14 | Bolivia (Plurinational State of) | 5,665   | 6,925   | 8,201     | 9,549     | 11,954    |
| 15 | Botswana                         | 4,275   | 5,396   | 7,937     | 10,256    | 11,845    |
| 16 | Brazil                           | 598,399 | 696,020 | 768,855   | 882,044   | 1,092,556 |
| 17 | Bulgaria                         | 26,246  | 22,996  | 22,131    | 28,894    | 32,932    |
| 18 | Burundi                          | 1,190   | 1,070   | 1,006     | 1,117     | 1,404     |
| 19 | Cambodia                         | 2,086   | 2,851   | 4,027     | 6,293     | 8,694     |
| 20 | Cameroon                         | 11,706  | 10,993  | 13,827    | 16,588    | 19,027    |
| 21 | Canada                           | 749,884 | 816,662 | 999,927   | 1,133,757 | 1,203,888 |
| 22 | Central African Republic         | 1,268   | 1,328   | 1,381     | 1,350     | 1,611     |
| 23 | Chile                            | 51,765  | 78,512  | 96,236    | 118,250   | 138,703   |
| 24 | China                            | 531,890 | 948,490 | 1,433,854 | 2,283,671 | 3,883,522 |
| 25 | Colombia                         | 94,460  | 117,224 | 122,698   | 146,566   | 183,182   |
| 26 | Congo                            | 4,330   | 4,428   | 4,988     | 6,087     | 7,878     |
| 27 | Costa Rica                       | 9,845   | 12,846  | 16,341    | 19,965    | 24,760    |
| 28 | Côte d'Ivoire                    | 12,815  | 13,878  | 16,330    | 16,354    | 18,026    |
| 29 | Croatia*                         | 33,241  | 30,530  | 36,034    | 44,821    | 46,895    |
| 30 | Cuba                             | 38,539  | 26,732  | 33,378    | 42,644    | 55,262    |
| 31 | Cyprus                           | 9,291   | 11,983  | 14,427    | 16,902    | 19,019    |
| 32 | Czech Republic**                 | 89,680  | 97,174  | 106,445   | 130,066   | 148,578   |
| 33 | Democratic Republic of the Congo | 10,239  | 7,100   | 5,817     | 7,166     | 9,400     |
| 34 | Denmark                          | 187,355 | 210,307 | 242,097   | 257,676   | 258,095   |
| 35 | Dominican Republic               | 15,678  | 20,204  | 28,206    | 33,542    | 47,247    |
| 36 | Ecuador                          | 22,659  | 26,788  | 28,691    | 36,942    | 43,859    |

GDP change percentage change with respect to the base year, 1990

### GDP per capita in millions of constant 2005 US\$

|        |        |        |        |        |      |      | •    | <b>3</b> |
|--------|--------|--------|--------|--------|------|------|------|----------|
| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010     |
| 386    | 201    | 134    | 248    | 313    | -48  | -65  | -36  | -19      |
| 1,618  | 1,522  | 1,998  | 2,597  | 3,273  | -6   | 23   | 61   | 102      |
| 2,722  | 2,466  | 2,665  | 3,139  | 3,313  | -9   | -2   | 15   | 22       |
| 3,363  | 4,194  | 4,495  | 4,736  | 6,279  | 25   | 34   | 41   | 87       |
| 1,146  | 665    | 895    | 1,598  | 1,913  | -42  | -22  | 39   | 67       |
| 26,405 | 29,328 | 33,643 | 37,482 | 39,270 | 11   | 27   | 42   | 49       |
| 28,212 | 30,280 | 35,057 | 37,048 | 38,983 | 7    | 24   | 31   | 38       |
| 13,077 | 14,710 | 15,745 | 18,569 | 14,123 | 12   | 20   | 42   | 8        |
| 265    | 292    | 341    | 410    | 524    | 10   | 29   | 55   | 98       |
| 11,737 | 11,073 | 12,645 | 13,623 | 13,605 | -6   | 8    | 16   | 16       |
| 28,128 | 30,045 | 34,260 | 36,225 | 37,334 | 7    | 22   | 29   | 33       |
| 2,723  | 2,904  | 3,422  | 3,968  | 4,039  | 7    | 26   | 46   | 48       |
| 480    | 498    | 547    | 571    | 591    | 4    | 14   | 19   | 23       |
| 851    | 927    | 987    | 1,044  | 1,204  | 9    | 16   | 23   | 41       |
| 3,092  | 3,403  | 4,515  | 5,468  | 5,902  | 10   | 46   | 77   | 91       |
| 3,999  | 4,300  | 4,408  | 4,743  | 5,604  | 8    | 10   | 19   | 40       |
| 2,976  | 2,752  | 2,764  | 3,734  | 4,394  | -8   | -7   | 25   | 48       |
| 213    | 176    | 158    | 154    | 167    | -17  | -26  | -28  | -21      |
| 219    | 255    | 324    | 471    | 615    | 17   | 48   | 115  | 181      |
| 961    | 789    | 882    | 945    | 971    | -18  | -8   | -2   | 1        |
| 27,071 | 27,870 | 32,606 | 35,119 | 35,391 | 3    | 20   | 30   | 31       |
| 432    | 399    | 373    | 336    | 366    | -8   | -14  | -22  | -15      |
| 3,925  | 5,449  | 6,241  | 7,254  | 8,105  | 39   | 59   | 85   | 106      |
| 464    | 781    | 1,130  | 1,746  | 2,895  | 68   | 143  | 276  | 523      |
| 2,845  | 3,216  | 3,086  | 3,405  | 3,957  | 13   | 8    | 20   | 39       |
| 1,813  | 1,620  | 1,591  | 1,723  | 1,949  | -11  | -12  | -5   | 7        |
| 3,206  | 3,703  | 4,169  | 4,633  | 5,315  | 15   | 30   | 44   | 66       |
| 1,024  | 946    | 985    | 907    | 913    | -8   | -4   | -11  | -11      |
| 7,309  | 6,539  | 7,998  | 10,090 | 10,650 | -11  | 9    | 38   | 46       |
| 3,646  | 2,452  | 3,006  | 3,789  | 4,909  | -33  | -18  | 4    | 35       |
| 12,118 | 14,008 | 15,294 | 16,369 | 17,233 | 16   | 26   | 35   | 42       |
| 8,695  | 9,417  | 10,392 | 12,726 | 14,160 | 8    | 20   | 46   | 63       |
| 281    | 161    | 117    | 125    | 143    | -43  | -58  | -56  | -49      |
| 36,443 | 40,186 | 45,341 | 47,546 | 46,502 | 10   | 24   | 30   | 28       |
| 2,179  | 2,552  | 3,283  | 3,621  | 4,759  | 17   | 51   | 66   | 118      |
| 2,208  | 2,353  | 2,324  | 2,751  | 3,032  | 7    | 5    | 25   | 37       |

## GDP

GDP level in millions of constant 2005 US\$

|    | Country/Year                     | 1990      | 1995      | 2000      | 2005      | 2010      |
|----|----------------------------------|-----------|-----------|-----------|-----------|-----------|
| 37 | Egypt                            | 48,637    | 61,120    | 78,838    | 94,461    | 127,480   |
| 38 | El Salvador                      | 9,702     | 13,093    | 15,219    | 17,094    | 18,352    |
| 39 | Estonia                          | 10,163    | 7,121     | 9,841     | 13,903    | 13,898    |
| 40 | Fiji                             | 2,091     | 2,319     | 2,737     | 3,006     | 2,957     |
| 41 | Finland                          | 140,212   | 135,908   | 171,860   | 195,626   | 206,858   |
| 42 | France                           | 1,623,841 | 1,725,643 | 1,973,037 | 2,136,555 | 2,208,616 |
| 43 | Gabon                            | 7,511     | 8,756     | 8,713     | 9,459     | 10,432    |
| 44 | Gambia                           | 388       | 435       | 535       | 630       | 797       |
| 45 | Germany                          | 2,216,250 | 2,448,688 | 2,685,203 | 2,766,254 | 2,945,784 |
| 46 | Ghana                            | 8,719     | 10,740    | 13,335    | 17,198    | 23,251    |
| 47 | Greece                           | 156,255   | 166,243   | 196,959   | 240,076   | 243,230   |
| 48 | Guatemala                        | 15,661    | 19,313    | 23,442    | 27,211    | 32,481    |
| 49 | Guyana                           | 788       | 1,110     | 1,270     | 1,315     | 1,615     |
| 50 | Haiti                            | 4,308     | 3,504     | 3,912     | 3,807     | 3,963     |
| 51 | Honduras                         | 5,625     | 6,695     | 7,773     | 9,757     | 11,576    |
| 52 | Hungary                          | 86,539    | 77,802    | 89,959    | 110,322   | 109,265   |
| 53 | Iceland                          | 10,277    | 10,413    | 13,216    | 16,286    | 16,399    |
| 54 | India                            | 350,886   | 451,010   | 597,744   | 837,299   | 1,251,603 |
| 55 | Indonesia                        | 150,091   | 218,808   | 226,917   | 285,869   | 377,282   |
| 56 | Iran (Islamic Republic of)       | 105,572   | 127,013   | 155,413   | 205,586   | 241,248   |
| 57 | Iraq                             | 12,257    | 6,040     | 17,713    | 18,164    | 25,354    |
| 58 | Ireland                          | 82,700    | 103,382   | 159,786   | 203,280   | 202,325   |
| 59 | Israel                           | 68,727    | 94,311    | 120,913   | 133,968   | 164,136   |
| 60 | Italy                            | 1,453,346 | 1,548,111 | 1,700,969 | 1,777,694 | 1,743,957 |
| 61 | Jamaica                          | 8,997     | 10,407    | 10,383    | 11,163    | 11,061    |
| 62 | Japan                            | 3,794,070 | 4,068,395 | 4,265,774 | 4,552,191 | 4,578,543 |
| 63 | Jordan                           | 5,833     | 7,894     | 9,244     | 12,589    | 16,732    |
| 64 | Kazakhstan*                      | 44,727    | 30,835    | 34,877    | 57,124    | 76,805    |
| 65 | Kenya                            | 12,995    | 14,059    | 15,570    | 18,739    | 23,450    |
| 66 | Kuwait                           | 39,310    | 51,472    | 54,707    | 80,798    | 90,039    |
| 67 | Kyrgyzstan*                      | 2,825     | 1,555     | 2,043     | 2,460     | 3,029     |
| 68 | Lao People's Democratic Republic | 1,096     | 1,495     | 2,016     | 2,739     | 4,380     |
| 69 | Latvia                           | 15,413    | 8,327     | 10,735    | 15,938    | 15,399    |
| 70 | Lesotho                          | 795       | 972       | 1,178     | 1,355     | 1,638     |
| 71 | Liberia                          | 579       | 162       | 730       | 578       | 803       |
| 72 | Lithuania*                       | 23,660    | 14,436    | 17,939    | 26,100    | 27,466    |

GDP change percentage change with respect to the base year, 1990

### GDP per capita in millions of constant 2005 US\$

| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
|--------|--------|--------|--------|--------|------|------|------|------|
| 856    | 985    | 1,165  | 1,273  | 1,571  | 15   | 36   | 49   | 84   |
| 1,819  | 2,284  | 2,562  | 2,825  | 2,963  | 26   | 41   | 55   | 63   |
| 6,483  | 4,943  | 7,179  | 10,330 | 10,363 | -24  | 11   | 59   | 60   |
| 2,870  | 2,990  | 3,372  | 3,655  | 3,436  | 4    | 18   | 27   | 20   |
| 28,119 | 26,608 | 33,220 | 37,302 | 38,560 | -5   | 18   | 33   | 37   |
| 28,635 | 29,832 | 33,414 | 35,028 | 35,176 | 4    | 17   | 22   | 23   |
| 8,084  | 8,053  | 7,053  | 6,901  | 6,929  | 0    | -13  | -15  | -14  |
| 401    | 386    | 412    | 419    | 461    | -4   | 3    | 4    | 15   |
| 28,019 | 29,888 | 32,608 | 33,514 | 35,792 | 7    | 16   | 20   | 28   |
| 589    | 632    | 696    | 795    | 953    | 7    | 18   | 35   | 62   |
| 15,379 | 15,578 | 17,927 | 21,468 | 21,412 | 1    | 17   | 40   | 39   |
| 1,755  | 1,928  | 2,086  | 2,140  | 2,257  | 10   | 19   | 22   | 29   |
| 1,087  | 1,525  | 1,732  | 1,763  | 2,141  | 40   | 59   | 62   | 97   |
| 605    | 445    | 453    | 407    | 397    | -26  | -25  | -33  | -34  |
| 1,151  | 1,201  | 1,250  | 1,418  | 1,523  | 4    | 9    | 23   | 32   |
| 8,340  | 7,531  | 8,810  | 10,937 | 10,944 | -10  | 6    | 31   | 31   |
| 40,335 | 38,929 | 46,996 | 54,884 | 51,224 | -3   | 17   | 36   | 27   |
| 402    | 468    | 567    | 734    | 1,022  | 16   | 41   | 83   | 155  |
| 814    | 1,097  | 1,063  | 1,258  | 1,573  | 35   | 31   | 54   | 93   |
| 1,924  | 2,125  | 2,378  | 2,948  | 3,261  | 10   | 24   | 53   | 70   |
| 706    | 298    | 742    | 664    | 801    | -58  | 5    | -6   | 13   |
| 23,420 | 28,630 | 42,007 | 48,888 | 45,264 | 22   | 79   | 109  | 93   |
| 15,273 | 17,687 | 20,102 | 20,284 | 22,126 | 16   | 32   | 33   | 45   |
| 25,573 | 27,175 | 29,849 | 30,299 | 28,802 | 6    | 17   | 18   | 13   |
| 3,804  | 4,228  | 4,022  | 4,163  | 4,035  | 11   | 6    | 9    | 6    |
| 31,035 | 32,681 | 33,931 | 36,016 | 36,184 | 5    | 9    | 16   | 17   |
| 1,708  | 1,801  | 1,915  | 2,357  | 2,704  | 5    | 12   | 38   | 58   |
| 2,709  | 1,936  | 2,332  | 3,765  | 4,792  | -29  | -14  | 39   | 77   |
| 554    | 513    | 498    | 526    | 579    | -8   | -10  | -5   | 4    |
| 18,830 | 31,618 | 28,188 | 35,688 | 32,900 | 68   | 50   | 90   | 75   |
| 636    | 339    | 412    | 488    | 568    | -47  | -35  | -23  | -11  |
| 261    | 312    | 379    | 476    | 706    | 19   | 45   | 82   | 170  |
| 5,786  | 3,341  | 4,501  | 6,913  | 6,838  | -42  | -22  | 19   | 18   |
| 485    | 542    | 600    | 656    | 754    | 12   | 24   | 35   | 55   |
| 272    | 77     | 256    | 182    | 201    | -72  | -6   | -33  | -26  |
| 6,398  | 3,978  | 5,126  | 7,641  | 8,264  | -38  | -20  | 19   | 29   |

## GDP

GDP level in millions of constant 2005 US\$

|     | Country/Year        | 1990    | 1995    | 2000    | 2005    | 2010      |
|-----|---------------------|---------|---------|---------|---------|-----------|
| 73  | Luxembourg          | 19,315  | 23,446  | 31,570  | 37,659  | 41,300    |
| 74  | Malawi              | 1,929   | 2,044   | 2,449   | 2,755   | 3,980     |
| 75  | Malaysia            | 55,084  | 86,608  | 109,442 | 137,954 | 171,826   |
| 76  | Maldives            | 286     | 397     | 592     | 750     | 1,025     |
| 77  | Mali                | 2,700   | 3,365   | 4,026   | 5,486   | 6,899     |
| 78  | Malta               | 3,412   | 4,456   | 5,721   | 5,981   | 6,653     |
| 79  | Mauritania          | 1,287   | 1,504   | 1,742   | 2,184   | 2,829     |
| 80  | Mauritius           | 3,331   | 4,211   | 5,516   | 6,489   | 8,121     |
| 81  | Mexico              | 547,801 | 590,980 | 770,735 | 846,095 | 922,307   |
| 82  | Mongolia            | 1,851   | 1,604   | 1,843   | 2,523   | 3,454     |
| 83  | Morocco             | 36,941  | 38,837  | 46,686  | 59,524  | 75,410    |
| 84  | Mozambique          | 2,340   | 2,746   | 4,309   | 6,579   | 9,324     |
| 85  | Myanmar             | 3,294   | 4,374   | 6,514   | 11,931  | 20,310    |
| 86  | Namibia             | 3,956   | 4,971   | 5,708   | 7,261   | 8,852     |
| 87  | Nepal               | 4,271   | 5,499   | 6,960   | 8,259   | 10,261    |
| 88  | Netherlands         | 437,826 | 490,389 | 597,952 | 638,471 | 685,082   |
| 89  | New Zealand         | 69,760  | 81,075  | 93,774  | 113,058 | 121,298   |
| 90  | Nicaragua           | 2,989   | 3,265   | 4,171   | 4,872   | 5,563     |
| 91  | Niger               | 2,143   | 2,307   | 2,702   | 3,369   | 4,290     |
| 92  | Nigeria             | 56,419  | 57,836  | 67,851  | 112,248 | 122,349   |
| 93  | Norway              | 188,114 | 225,884 | 270,651 | 302,013 | 315,275   |
| 94  | Pakistan            | 58,161  | 72,921  | 85,596  | 109,213 | 134,328   |
| 95  | Panama              | 7,643   | 9,986   | 12,524  | 15,465  | 22,978    |
| 96  | Papua New Guinea    | 2,790   | 4,213   | 4,374   | 4,866   | 6,421     |
| 97  | Paraguay            | 5,631   | 6,650   | 6,587   | 7,473   | 9,764     |
| 98  | Peru                | 43,529  | 57,183  | 64,657  | 79,389  | 112,100   |
| 99  | Philippines         | 62,103  | 69,129  | 82,358  | 103,072 | 131,138   |
| 100 | Poland              | 180,139 | 200,601 | 261,094 | 303,912 | 382,761   |
| 101 | Portugal            | 137,268 | 149,369 | 183,643 | 191,176 | 196,251   |
| 102 | Qatar               | 15,215  | 17,344  | 29,927  | 44,530  | 104,255   |
| 103 | Republic of Korea   | 360,297 | 526,721 | 678,270 | 844,866 | 1,017,571 |
| 104 | Republic of Moldova | 6,101   | 2,405   | 2,122   | 2,988   | 3,497     |
| 105 | Romania             | 85,267  | 76,580  | 75,104  | 99,173  | 111,921   |
| 106 | Russian Federation* | 711,344 | 524,113 | 567,392 | 764,016 | 905,248   |
| 107 | Rwanda              | 1,712   | 1,099   | 1,783   | 2,581   | 3,854     |
| 108 | Saudi Arabia        | 200,420 | 230,832 | 262,042 | 315,583 | 359,749   |

GDP change percentage change with respect to the base year, 1990

### GDP per capita in millions of constant 2005 US\$

|        |        |        |        |        |      |      | •    |      |
|--------|--------|--------|--------|--------|------|------|------|------|
| 1990   | 1995   | 2000   | 2005   | 2010   | 1995 | 2000 | 2005 | 2010 |
| 50,666 | 57,533 | 72,493 | 82,370 | 81,388 | 14   | 43   | 63   | 61   |
| 206    | 207    | 218    | 215    | 267    | 1    | 6    | 5    | 30   |
| 3,025  | 4,180  | 4,674  | 5,286  | 6,050  | 38   | 55   | 75   | 100  |
| 1,304  | 1,596  | 2,168  | 2,540  | 3,243  | 22   | 66   | 95   | 149  |
| 311    | 343    | 356    | 416    | 449    | 10   | 14   | 34   | 44   |
| 9,285  | 11,527 | 14,396 | 14,612 | 15,972 | 24   | 55   | 57   | 72   |
| 645    | 656    | 659    | 717    | 818    | 2    | 2    | 11   | 27   |
| 3,143  | 3,703  | 4,612  | 5,163  | 6,251  | 18   | 47   | 64   | 99   |
| 6,498  | 6,405  | 7,710  | 7,946  | 8,132  | -1   | 19   | 22   | 25   |
| 844    | 695    | 764    | 991    | 1,253  | -18  | -9   | 17   | 48   |
| 1,491  | 1,442  | 1,621  | 1,959  | 2,360  | -3   | 9    | 31   | 58   |
| 173    | 172    | 237    | 317    | 399    | 0    | 37   | 83   | 131  |
| 84     | 104    | 145    | 258    | 423    | 24   | 73   | 207  | 405  |
| 2,796  | 3,010  | 3,011  | 3,491  | 3,877  | 8    | 8    | 25   | 39   |
| 224    | 255    | 285    | 303    | 342    | 14   | 27   | 35   | 53   |
| 29,401 | 31,798 | 37,695 | 39,157 | 41,238 | 8    | 28   | 33   | 40   |
| 20,530 | 22,061 | 24,306 | 27,348 | 27,769 | 7    | 18   | 33   | 35   |
| 725    | 704    | 822    | 898    | 961    | -3   | 13   | 24   | 33   |
| 275    | 251    | 247    | 259    | 277    | -9   | -10  | -6   | 0    |
| 578    | 526    | 549    | 803    | 772    | -9   | -5   | 39   | 34   |
| 44,351 | 51,819 | 60,267 | 65,324 | 64,564 | 17   | 36   | 47   | 46   |
| 520    | 573    | 592    | 688    | 774    | 10   | 14   | 32   | 49   |
| 3,164  | 3,730  | 4,237  | 4,776  | 6,534  | 18   | 34   | 51   | 107  |
| 671    | 893    | 813    | 798    | 936    | 33   | 21   | 19   | 40   |
| 1,327  | 1,387  | 1,233  | 1,267  | 1,513  | 5    | -7   | -5   | 14   |
| 2,007  | 2,400  | 2,500  | 2,881  | 3,855  | 20   | 25   | 44   | 92   |
| 1,008  | 998    | 1,065  | 1,205  | 1,406  | -1   | 6    | 20   | 40   |
| 4,734  | 5,225  | 6,817  | 7,963  | 10,000 | 10   | 44   | 68   | 111  |
| 13,830 | 14,753 | 17,767 | 18,132 | 18,383 | 7    | 28   | 31   | 33   |
| 32,118 | 34,593 | 50,642 | 54,240 | 59,277 | 8    | 58   | 69   | 85   |
| 8,383  | 11,793 | 14,749 | 17,959 | 21,119 | 41   | 76   | 114  | 152  |
| 1,398  | 554    | 517    | 793    | 979    | -60  | -63  | -43  | -30  |
| 3,674  | 3,376  | 3,384  | 4,555  | 5,209  | -8   | -8   | 24   | 42   |
| 4,785  | 3,525  | 3,866  | 5,311  | 6,332  | -26  | -19  | 11   | 32   |
| 241    | 197    | 220    | 281    | 363    | -18  | -9   | 16   | 51   |
| 12,418 | 12,483 | 13,072 | 13,127 | 13,107 | 1    | 5    | 6    | 6    |

## GDP

GDP level in millions of constant 2005 US\$

|     | Country/Year                       | 1990      | 1995      | 2000       | 2005       | 2010       |
|-----|------------------------------------|-----------|-----------|------------|------------|------------|
| 109 | Senegal                            | 5,127     | 5,498     | 6,934      | 8,708      | 10,298     |
| 110 | Serbia                             | 38,758    | 18,559    | 19,632     | 25,231     | 29,523     |
| 111 | Sierra Leone                       | 1,799     | 1,393     | 817        | 1,491      | 1,904      |
| 112 | Singapore                          | 49,663    | 74,851    | 99,282     | 125,429    | 170,969    |
| 113 | Slovakia**                         | 27,500    | 31,501    | 37,697     | 47,895     | 60,184     |
| 114 | Slovenia*                          | 22,679    | 24,178    | 29,901     | 35,718     | 39,034     |
| 115 | South Africa                       | 170,914   | 178,412   | 204,703    | 247,052    | 288,441    |
| 116 | Spain                              | 730,335   | 787,041   | 962,433    | 1,130,170  | 1,180,660  |
| 117 | Sri Lanka                          | 12,021    | 15,678    | 20,044     | 24,406     | 33,252     |
| 118 | Sudan (former)                     | 14,180    | 17,935    | 25,262     | 35,183     | 51,663     |
| 119 | Swaziland                          | 1,715     | 1,983     | 2,321      | 2,596      | 2,889      |
| 120 | Sweden                             | 263,843   | 272,953   | 324,508    | 370,580    | 397,080    |
| 121 | Switzerland                        | 313,927   | 315,491   | 349,046    | 372,477    | 411,657    |
| 122 | Syrian Arab Republic               | 12,567    | 18,570    | 22,208     | 28,397     | 36,043     |
| 123 | Tajikistan*                        | 3,557     | 1,455     | 1,457      | 2,312      | 3,166      |
| 124 | Thailand                           | 88,907    | 134,468   | 137,515    | 176,352    | 210,077    |
| 125 | Тодо                               | 1,817     | 1,793     | 1,989      | 2,110      | 2,462      |
| 126 | Trinidad and Tobago                | 7,025     | 7,527     | 10,834     | 15,982     | 19,186     |
| 127 | Tunisia                            | 16,450    | 19,885    | 26,040     | 32,272     | 40,476     |
| 128 | Turkey                             | 269,684   | 315,856   | 386,584    | 482,986    | 564,315    |
| 129 | Uganda                             | 3,726     | 5,337     | 7,296      | 10,040     | 13,729     |
| 130 | Ukraine*                           | 125,433   | 65,598    | 59,344     | 86,142     | 90,700     |
| 131 | United Arab Emirates               | 81,923    | 101,625   | 139,151    | 180,617    | 211,214    |
| 132 | United Kingdom                     | 1,525,105 | 1,654,441 | 1,979,324  | 2,280,538  | 2,330,011  |
| 133 | United Republic of Tanzania        | 6,852     | 8,326     | 10,310     | 14,492     | 20,154     |
| 134 | United States of America           | 7,962,600 | 9,019,900 | 11,158,100 | 12,564,300 | 13,017,000 |
| 135 | Uruguay                            | 12,653    | 15,453    | 17,205     | 17,363     | 23,493     |
| 136 | Venezuela (Bolivarian Republic of) | 104,316   | 123,574   | 128,279    | 145,513    | 172,851    |
| 137 | Vietnam                            | 17,751    | 26,331    | 36,846     | 52,917     | 74,268     |
| 138 | Yemen                              | 7,276     | 9,495     | 14,188     | 17,872     | 22,927     |
| 139 | Zambia                             | 5,384     | 5,034     | 5,759      | 7,271      | 9,890      |
| 140 | Zimbabwe                           | 7.372     | 7.826     | 7.749      | 6.223      | 6.462      |

GDP change percentage change with respect to the base year, 1990

### GDP per capita in millions of constant 2005 US\$

| 2010 | 2005 | 2000 | 1995 | 2010   | 2005   | 2000   | 1995   | 1990   |
|------|------|------|------|--------|--------|--------|--------|--------|
| 17   | 13   | 3    | -7   | 828    | 801    | 729    | 657    | 708    |
| -26  | -37  | -52  | -55  | 2,995  | 2,560  | 1,937  | 1,819  | 4,051  |
| -28  | -36  | -56  | -21  | 324    | 289    | 197    | 357    | 452    |
| 104  | 79   | 54   | 31   | 33,613 | 29,402 | 25,332 | 21,499 | 16,463 |
| 113  | 71   | 35   | 13   | 11,019 | 8,844  | 6,975  | 5,867  | 5,174  |
| 64   | 52   | 29   | 5    | 19,232 | 17,840 | 15,060 | 12,297 | 11,719 |
| 24   | 11   | -2   | -7   | 5,754  | 5,169  | 4,573  | 4,309  | 4,645  |
| 36   | 39   | 27   | 6    | 25,624 | 26,044 | 23,889 | 19,962 | 18,780 |
| 130  | 77   | 54   | 24   | 1,594  | 1,230  | 1,069  | 860    | 693    |
| 122  | 71   | 38   | 11   | 1,186  | 916    | 739    | 595    | 535    |
| 23   | 18   | 10   | 4    | 2,436  | 2,349  | 2,182  | 2,058  | 1,988  |
| 37   | 33   | 19   | 0    | 42,334 | 41,042 | 36,626 | 30,923 | 30,827 |
| 14   | 7    | 4    | -4   | 53,711 | 50,233 | 48,696 | 44,950 | 47,039 |
| 73   | 51   | 36   | 29   | 1,766  | 1,536  | 1,389  | 1,310  | 1,020  |
| -30  | -45  | -64  | -62  | 460    | 358    | 236    | 252    | 656    |
| 95   | 70   | 40   | 45   | 3,039  | 2,644  | 2,177  | 2,254  | 1,558  |
| -18  | -21  | -16  | -11  | 408    | 390    | 415    | 439    | 496    |
| 147  | 110  | 45   | 3    | 14,302 | 12,150 | 8,385  | 5,967  | 5,780  |
| 93   | 63   | 38   | 11   | 3,862  | 3,256  | 2,754  | 2,225  | 2,002  |
| 56   | 42   | 22   | 8    | 7,757  | 7,088  | 6,076  | 5,366  | 4,982  |
| 95   | 68   | 43   | 22   | 411    | 353    | 301    | 256    | 210    |
| -18  | -24  | -50  | -47  | 1,996  | 1,836  | 1,214  | 1,283  | 2,428  |
| -38  | -2   | 1    | -4   | 28,118 | 44,385 | 45,871 | 43,272 | 45,296 |
| 41   | 42   | 26   | 7    | 37,559 | 37,881 | 33,620 | 28,526 | 26,656 |
| 67   | 39   | 13   | 3    | 449    | 373    | 303    | 278    | 269    |
| 33   | 35   | 26   | 8    | 41,938 | 42,330 | 39,498 | 33,868 | 31,431 |
| 71   | 28   | 27   | 18   | 6,974  | 5,226  | 5,184  | 4,794  | 4,070  |
| 13   | 3    | -1   | 6    | 5,965  | 5,457  | 5,268  | 5,608  | 5,299  |
| 220  | 141  | 77   | 34   | 845    | 636    | 468    | 356    | 265    |
| 57   | 42   | 31   | 3    | 953    | 866    | 801    | 627    | 609    |
| 10   | -7   | -18  | -18  | 756    | 634    | 565    | 564    | 685    |
| -27  | -30  | -12  | -5   | 514    | 495    | 619    | 670    | 704    |

# Country classification by region

|    | Africa                           | Asia                             | Europe              |
|----|----------------------------------|----------------------------------|---------------------|
| 1  | Eastern Africa                   | Eastern Asia                     | Eastern Europe      |
| 2  | Burundi                          | China                            | Bulgaria            |
| 3  | Kenya                            | Japan                            | Czech Republic      |
| 4  | Malawi                           | Mongolia                         | Hungary             |
| 5  | Mauritius                        | Republic of Korea                | Poland              |
| 6  | Mozambique                       | South-Central Asia               | Republic of Moldova |
| 7  | Rwanda                           | Afghanistan                      | Romania             |
| 8  | Uganda                           | Bangladesh                       | Russian Federation  |
| 9  | United Republic of Tanzania      | India                            | Slovakia            |
| 10 | Zambia                           | Iran (Islamic Republic of)       | Ukraine             |
| 11 | Zimbabwe                         | Kazakhstan                       | Northern Europe     |
| 12 | Middle Africa                    | Kyrgyzstan                       | Denmark             |
| 13 | Cameroon                         | Maldives                         | Estonia             |
| 14 | Central African Republic         | Nepal                            | Finland             |
| 15 | Congo                            | Pakistan                         | Iceland             |
| 16 | Democratic Republic of the Congo | Sri Lanka                        | Ireland             |
| 17 | Gabon                            | Tajikistan                       | Latvia              |
| 18 | Northern Africa                  | South–Eastern Asia               | Lithuania           |
| 19 | Algeria                          | Cambodia                         | Norway              |
| 20 | Egypt                            | Indonesia                        | Sweden              |
| 21 | Morocco                          | Lao People's Democratic Republic | United Kingdom      |
| 22 | Sudan (former)                   | Malaysia                         | Southern Europe     |
| 23 | Tunisia                          | Myanmar                          | Albania             |
| 24 | Southern Africa                  | Philippines                      | Croatia             |
| 25 | Botswana                         | Singapore                        | Greece              |
| 26 | Lesotho                          | Thailand                         | Italy               |
| 27 | Namibia                          | Vietnam                          | Malta               |
| 28 | South Africa                     | Western Asia                     | Portugal            |
| 29 | Swaziland                        | Armenia                          | Serbia              |
| 30 | Western Africa                   | Bahrain                          | Slovenia            |
| 31 | Benin                            | Cyprus                           | Spain               |
| 32 | Côte d'Ivoire                    | Iraq                             | Western Europe      |
| 33 | Gambia                           | Israel                           | Austria             |
| 34 | Ghana                            | Jordan                           | Belgium             |
| 35 | Liberia                          | Kuwait                           | France              |
| 36 | Mali                             | Qatar                            | Germany             |
| 37 | Mauritania                       | Saudi Arabia                     | Luxembourg          |
| 38 | Niger                            | Syrian Arab Republic             | Netherlands         |
| 39 | Nigeria                          | Turkey                           | Switzerland         |
| 40 | Senegal                          | United Arab Emirates             |                     |
| 41 | Sierra Leone                     | Yemen                            |                     |
| 42 | Тодо                             |                                  |                     |

|    | Latin America and the Caribbean    | Northern America         | Oceania               |  |  |
|----|------------------------------------|--------------------------|-----------------------|--|--|
| 43 | Caribbean                          | Northern America         | Australia/New Zealand |  |  |
| 44 | Barbados                           | Canada                   | Australia             |  |  |
| 45 | Cuba                               | United States of America | New Zealand           |  |  |
| 46 | Dominican Republic                 |                          | Melanesia             |  |  |
| 47 | Haiti                              |                          | Fiji                  |  |  |
| 48 | Jamaica                            |                          | Papua New Guinea      |  |  |
| 49 | Trinidad and Tobago                |                          |                       |  |  |
| 50 | Central America                    |                          |                       |  |  |
| 51 | Belize                             |                          |                       |  |  |
| 52 | Costa Rica                         |                          |                       |  |  |
| 53 | El Salvador                        |                          |                       |  |  |
| 54 | Guatemala                          |                          |                       |  |  |
| 55 | Honduras                           |                          |                       |  |  |
| 56 | Mexico                             |                          |                       |  |  |
| 57 | Nicaragua                          |                          |                       |  |  |
| 58 | Panama                             |                          |                       |  |  |
| 59 | South America                      |                          |                       |  |  |
| 60 | Argentina                          |                          |                       |  |  |
| 61 | Bolivia (Plurinational State of)   |                          |                       |  |  |
| 62 | Brazil                             |                          |                       |  |  |
| 63 | Chile                              |                          |                       |  |  |
| 64 | Colombia                           |                          |                       |  |  |
| 65 | Ecuador                            |                          |                       |  |  |
| 66 | Guyana                             |                          |                       |  |  |
| 67 | Paraguay                           |                          |                       |  |  |
| 68 | Peru                               |                          |                       |  |  |
| 69 | Uruguay                            |                          |                       |  |  |
| 70 | Venezuela (Bolivarian Republic of) |                          |                       |  |  |

# Country classification by income levels

|    | High income              | Upper middle income                | Lower middle income              | Low income                       |
|----|--------------------------|------------------------------------|----------------------------------|----------------------------------|
| 1  | Australia                | Algeria                            | Albania                          | Afghanistan                      |
| 2  | Austria                  | Argentina                          | Armenia                          | Bangladesh                       |
| 3  | Bahrain                  | Botswana                           | Belize                           | Benin                            |
| 4  | Barbados                 | Brazil                             | Bolivia (Plurinational State of) | Burundi                          |
| 5  | Belgium                  | Bulgaria                           | Cameroon                         | Cambodia                         |
| 6  | Canada                   | Chile                              | Congo                            | Central African Republic         |
| 7  | Croatia                  | China                              | Côte d'Ivoire                    | Democratic Republic of the Congo |
| 8  | Cyprus                   | Colombia                           | Egypt                            | Gambia                           |
| 9  | Czech Republic           | Costa Rica                         | El Salvador                      | Haiti                            |
| 10 | Denmark                  | Cuba                               | Fiji                             | Kenya                            |
| 11 | Estonia                  | Dominican Republic                 | Ghana                            | Kyrgyzstan                       |
| 12 | Finland                  | Ecuador                            | Guatemala                        | Liberia                          |
| 13 | France                   | Gabon                              | Guyana                           | Malawi                           |
| 14 | Germany                  | Iran (Islamic Republic of)         | Honduras                         | Mali                             |
| 15 | Greece                   | Jamaica                            | India                            | Mauritania                       |
| 16 | Hungary                  | Jordan                             | Indonesia                        | Mozambique                       |
| 17 | Iceland                  | Kazakhstan                         | Iraq                             | Myanmar                          |
| 18 | Ireland                  | Latvia                             | Lao People's Democratic Republic | Nepal                            |
| 19 | Israel                   | Lithuania                          | Lesotho                          | Niger                            |
| 20 | Italy                    | Malaysia                           | Mongolia                         | Rwanda                           |
| 21 | Japan                    | Maldives                           | Morocco                          | Sierra Leone                     |
| 22 | Kuwait                   | Mauritius                          | Nicaragua                        | Tajikistan                       |
| 23 | Luxembourg               | Mexico                             | Nigeria                          | Тодо                             |
| 24 | Malta                    | Namibia                            | Pakistan                         | Uganda                           |
| 25 | Netherlands              | Panama                             | Papua New Guinea                 | United Republic of Tanzania      |
| 26 | New Zealand              | Peru                               | Paraguay                         | Zimbabwe                         |
| 27 | Norway                   | Romania                            | Philippines                      |                                  |
| 28 | Poland                   | Russian Federation                 | Republic of Moldova              |                                  |
| 29 | Portugal                 | Serbia                             | Senegal                          |                                  |
| 30 | Qatar                    | South Africa                       | Sri Lanka                        |                                  |
| 31 | Republic of Korea        | Thailand                           | Sudan (former)                   |                                  |
| 32 | Saudi Arabia             | Tunisia                            | Swaziland                        |                                  |
| 33 | Singapore                | Turkey                             | Syrian Arab Republic             |                                  |
| 34 | Slovakia                 | Uruguay                            | Ukraine                          |                                  |
| 35 | Slovenia                 | Venezuela (Bolivarian Republic of) | Viet Nam                         |                                  |
| 36 | Spain                    |                                    | Yemen                            |                                  |
| 37 | Sweden                   |                                    | Zambia                           |                                  |
| 38 | Switzerland              |                                    |                                  |                                  |
| 39 | Trinidad and Tobago      |                                    |                                  |                                  |
| 40 | United Arab Emirates     |                                    |                                  |                                  |
| 41 | United Kingdom           |                                    |                                  |                                  |
| 42 | United States of America |                                    |                                  |                                  |

# Glossary of terms

Absolute decoupling: refers to a situation of an overall reduction in required material inputs or pollution outputs, even while the economy grows, whether through productivity improvements or through a decrease in pollution, or a combination of both.

Adjusted net savings: a measure of net change in the value of a country's capital stocks, including produced, human, and natural capital.

Adjusted Inclusive Wealth Index (Wadj): a measure of wealth that results from the accounting for those factors that affect the size of a country's productive base. In this report, we take into account the following three components: carbon damages, oil capital gains, and total factor productivity.

**Biodiversity**: the variability among living organisms from all sources and the ecological complexes which they are part of, including diversity within species, between species, and of ecosystems.

**Biosphere**: a limited space made up of air, earth, and water, which is capable to support life.

**Carbon emissions**: the release of carbon dioxide gas into the atmosphere.

**Comprehensive wealth**: the shadow value of all capital assets in a country. See inclusive wealth.

**Decoupling**: a decline in the ratio of the amount used of a certain resource, or of the environmental impact, to the value generated or otherwise involved in the resource use or environmental impact. **Dematerialization**: an absolute decrease in the quantity of resources, measured by mass, used by an economy.

**Ecological capital**: ecosystems considered capital assets.

**Ecosystem**: a mesh of human and natural resources interacting with one another at a multitude of speeds and across often overlapping spatial scales.

**Ecosystem services**: ecosystem services cover the provision of ecosystem inputs, the assimilative capacity of the environment, and the provision of biodiversity.

**Educational attainment**: refers to the highest level of schooling which each student has attended and completed successfully. In this report it is represented by the average years of total schooling per person.

**Energy security**: the uninterrupted availability of energy sources at an affordable price.

**Environmental sustainability**: the maintenance of the minimum thresholds of natural capital that are required to sustain important environmental functions.

**Externalities**: the effects of activities on the well-being of people who have not been part of the negotiations that led to those activities.

**Food security**: the state that all people at all times have physical and economic access to sufficient, safe, and nutritious food to maintain a healthy and active life.

**Fossil fuels**: fuels such as natural gas, coal, and petroleum that are formed in the earth from plant or animal remains.

**Global genuine saving rate**: the world's gross savings plus educational expenditures minus produced capital depreciation and the values of natural resource depletion and the damages resulting from carbon emissions.

**Gross domestic product**: the market value of all final goods and services produced within an economy within a year.

Health capital: in this report, health capital comprises the three components: direct wellbeing, productivity, and longevity. As accepted empirical estimates of the first two components are lacking, it is here essentially measured by the extensions (reductions) in the individual's life expectancy. Such changes are basically analyzed by calculating the expected discounted years of life remaining of a given population.

**Human capital**: the knowledge, skills, competencies, and attributes embodied in individuals that facilitate the creation of personal, social, and economic well-being

Human Development Index: a measure of development designed by the United Nations Development Programme that combines indicators of life expectancy, educational attainment, and income.

**Inclusive investment**: the measurement of the physical changes in the assets of the economy, while holding shadow prices constant.

**Inclusive wealth**: the shadow value of all capital assets in a country. In this report comprehensive wealth and inclusive wealth are used interchangeably. **Inclusive Wealth Index (W)**: a measure of intertemporal social welfare to assess nations' social progress by taking human, natural, and produced capital into account.

**Infrastructure capital**: is a subset of produced capital, and is commonly defined as the built civic assets essential for the functioning of an economy and society.

**Produced capital**: traditionally considered as capital, consisting of roads, buildings, ports, machinery, and equipment. The latter manufactured assets, which remain within the production system for more than one accounting period – e.g. year –, are used to produce other goods and services.

**Microclimate**: the climate of a small, specific place within an area as contrasted with the climate of the entire area.

**Millennium Ecosystem Assessment**: a pioneering study from 2005 of the services humanity enjoys from ecosystems.

**Natural capital**: everything in nature (biotic and abiotic) capable of providing human beings with well-being, either directly or through the production process.

**Non-renewable resources**: natural resources that cannot be regenerated or grown at a sustainable rate to meet demand, including fossil fuels, metals, and minerals.

**Oil capital gains**: measure of a change in wealth due to changes in price given a fixed quantity of oil.

Manufactured capital: see produced capital.

**Relative decoupling**: refers to a situation where productivity/efficiency improvements have been realized but total inputs, or pollution outputs, continue to increase as economic output increases. **Renewable resources**: natural resources that, after exploitation, can return to their previous stock levels by natural processes of growth or replenishment. Conditionally renewable resources are those whose exploitation eventually reaches a level beyond which regeneration will become impossible. Such is the case with the clear cutting of tropical forests.

**Rental prices**: market prices minus production costs of resources.

**Satellite accounts**: provide a framework linked to the to central (national or regional) accounts, allowing attention to be focused on a certain field or aspect of economic and social life in the context of national accounts; common examples are satellite accounts for the environment, or tourism, or unpaid household work.

**Shadow price**: the shadow price of a capital asset is the forecast contribution of a marginal unit of this asset to human well-being.

**Social capital**: aspects of social structure that facilitate action, in terms of the importance of obligations and expectations, information channels, and social norms to education.

**Sustainable development**: development that meets the needs of the present without compromising the ability of future generations to meet their own needs.

**Sustainability gap:** indicates the degree of consumption of natural capital, either in the past or present, that is in excess of what is required for environmental sustainability.

**System of National Accounts**: is an internationally agreed-upon standard statistical framework that provides a comprehensive, consistent, and flexible set of macroeconomic accounts suitable for measuring, monitoring, and analyzing the economy and its constituents, so as to assist national policy planning processes. **Total factor productivity:** the proportion of output not explained by the amount of inputs used in production. It captures the effect of technical progress, the efficiency with which inputs are used, institutional conditions, and the impact of environmental factors such as climate.

Value of statistical life: an approach measuring a society's willingness to pay to avoid additional occurrences of death.

Value of statistical life year: the value of statistical life year is the value of statistical life divided by the expected discounted years of life remaining and thus represents a per-year valuation of the reduction in risk of mortality.

# Contributing organizations

#### UNU-IHDP

The International Human Dimensions Programme on Global Environmental Change (IHDP) was an interdisciplinary science programme that worked toward a better understanding of human interaction with and within their natural environment. Until its closure in June 2014, the programme facilitated dialogue between science and policy to ensure that research results fed into policy-planning and law-making processes, and offered education and training to future leaders in the field.

IHDP was founded by the International Council for Science (ICSU) and the International Social Science Council (ISSC) of UNESCO in 1996. Its Secretariat (UNU-IHDP) was hosted by the United Nations University (UNU) in Bonn who joined as third sponsor in 2007. IHDP's research was guided by an international Scientific Committee comprised of renowned scientists from various disciplinary and regional backgrounds.

#### UNEP

The United Nations Environment Programme (UNEP) is the voice for the environment within the United Nations system. Established in 1972, UNEP's mission is to provide leadership and encourage partnership in caring for the environment by inspiring, informing, and enabling nations and people to improve their quality of life without compromising that of future generations. UNEP acts as a catalyst, advocate, educator, and facilitator promoting the wise use and sustainable development of the global environment. UNEP works with a wide range of partners, including United Nations entities, international organizations, national governments, non-governmental organizations, the private sector, and the civil society.

UNEP's work involves providing support for environmental assessment and reporting; legal and institutional strengthening and environmental policy development; sustainable use and management of natural resources; integration of economic development and environmental protection; and promoting public participation in environmental management. For more information, please visit: www.unep.org.

### **UNESCO-MGIEP**

UNESCO-MGIEP is a specialized international UNESCO Category I Institute on Education for Peace and Sustainable Development (EPSD). Its mission is to build capacities of Member States and relevant stakeholders, and strengthen international policy to integrate EPSD into knowledge, education, and learning, in order to foster global citizenship. The institute was announced jointly by the President of India and the Director-General of UNESCO in 2011 and is located in New Delhi, India.

UNESCO-MGIEP's work supports the United Nations Secretary General's Education First Initiative, the post-2015 global education agenda, and the Sustainable Development Goals by strengthening education systems to nurture generations of global citizens with the understanding and skills required to build a more peaceful, sustainable, and inclusive future.

#### ASCENT

ASCENT is an innovative think tank in Nairobi, Kenya which focuses on sustainability, rule of law, social transformation, and economic empowerment through engagement of governments, private sector and civil society. The institute aims to move beyond traditional approaches, to harness the growth momentum for the youthful population of Africa to develop and support fresh ideas and technological adaptations that address scarcity and hardship as well as advance social progress.

The institute is inspired by the ancient African Adinkra adage Boa me na me mmoa wo, meaning, "help me help you", in other words, we believe that the best way to address challenges and provide a healthy and sustainable future in Africa is by creating an environment of cooperation, partnership and interdependence. www.africasustianability.org.

#### MIGHT

The Malaysian Industry-Government Group for High Technology (MIGHT) was formed in 1993 as Malaysia embarked on its drive to encourage its industries to accelerate the adoption of high technology, thus enabling the country to remain competitive in the face of growing globalization and trade liberalization.

MIGHT is an independent and not-for-profit organization driven by key players from both the public and private sectors. The synergies arising from these engagements form the basis for efforts in prospecting cutting-edge technological know-how and opportunities that can be nurtured into viable businesses for wealth creation.

#### Science to Action

Science to Action (S2A) is a national initiative announced by YAB Prime Minister of Malaysia in 2013, to be led by the Office of the Science Advisor and anchored by the Malaysian Industry-Government Group for High Technology (MIGHT). The objectives of the initiative are twofold: to identify S&T programs that will enable nation-building, and raising the profile of S&T in national development. Implementation of the S2A initiative will be undertaken parallel to the following targets:

- Global Competitiveness Index targeting Malaysia as a top 10 nation;
- Global Innovation Index Malaysia firmly recognized as a top 10 nation;
- Gross Domestic Expenditure on R&D increasing the country's GERD to 2%.

The S2A is a comprehensive effort that will enable Malaysia to sustain its growth beyond 2020. It is complementary in nature and innately supports various existing initiatives that have already been undertaken by the government, while at the same time ensuring that the nation is able to maximize its potential through the use of S&T. Ultimately, S2A will enable the nation to generate new and vibrant programs, through strategies that strengthen the STI governance and development ecosystem. It will also be focusing on selected target areas, as well as enabling active participations from the private sector to lead initiatives in STI.
## Ministry of Environment - Government of Japan

The Ministry of the Environment is the government body responsible for environmental protection in Japan. Established in 1971 as the Environment Agency, it has worked on policy-making in global environmental issues, pollution control, and nature conservation.

## **UNU-IAS**

The United Nations University Institute for the Advanced Study of Sustainability (UNU-IAS) is a leading research and teaching institute based in Tokyo, Japan. Its mission is to advance efforts towards a more sustainable future, through policy-oriented research and capacity development focused on sustainability and its social, economic and environmental dimensions. UNU-IAS serves the international community, making valuable and innovative contributions to high-level policymaking and debates within the UN system.

## The Science and Technology Alliance for Global Sustainability

The Science and Technology Alliance for Global Sustainability is a partnership of international organizations working strategically in critical science, education, technology, and innovation research areas. Through its members, the Alliance represents a global community of researchers, science policy-makers, and funders committed to transformative research that leads to strategic action on global sustainability.

Drawing on their respective networks and resources, Alliance members join forces to develop transformative research agendas that use critical scientific knowledge to drive strategic global sustainability actions by decision-makers, policy-shapers, practitioners, the private sector, and the civil society.

The Alliance's first flagship initiative is Future Earth, an international program of integrated, solutions-oriented research for global sustainability. Please visit www.futureearth. info for more information.