A Dynamic Analysis of Human Welfare in a Warming Planet

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Climate science indicates that climate stabilization requires low GHG emissions. Is this consistent with nondecreasing human welfare?

Our welfare or utility index emphasizes education, knowledge, and the environment. We construct and calibrate a multigenerational model with intertemporal links provided by education, physical capital, knowledge and the environment.

We reject discounted utilitarianism and adopt, first, the Pure Sustainability Optimization (or Intergenerational Maximin) criterion, and, second, the Sustainable Growth Optimization criterion, that maximizes the utility of the first generation subject to a given future rate of growth. We apply these criteria to our calibrated model via a novel algorithm inspired by the turnpike property.

The computed paths yield levels of utility higher than the level at reference year 2000 for all generations. They require the doubling of the fraction of labor resources devoted to the creation of knowledge relative to the reference level, whereas the fractions of labor allocated to consumption and leisure are similar to the reference ones. On the other hand, higher growth rates require substantial increases in the fraction of labor devoted to education, together with moderate increases in the fractions of labor devoted to knowledge and the investment in physical capital.

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**JEL classification numbers**: D63, O40, O41, Q50, Q54, Q56.
1. Introduction

Since the late 1980’s, scientists have become increasingly concerned with the effect of the emission of greenhouse gases (GHGs) on global temperature. The Intergovernmental Panel on Climate Change (IPCC) has now issued four reports, documenting the conjecture, expressed with increasing levels of confidence, that recent increases in global temperature are primarily anthropogenic in origin, attributable in the main, but not solely, to the burning of fossil fuels. Much has been written about strategies of mitigation of these emissions, and/or adaptation to the higher temperatures that will ensue if we extrapolate according to their present rate of growth.

In this article, we study the problem of intergenerational equity in a world that is constrained to limit GHG emissions in order to keep global temperature at an acceptably low level. We construct and calibrate a dynamic model involving economic and environmental variables. We eschew the specification of a physical model of emission-stock interactions, and consider instead a particular path for the environmental variables, which entails very low emissions after 2050, and realistically appears to be feasible given present knowledge of climate dynamics. The economic variables are then endogenous in our optimization program. We develop a computational algorithm based on the turnpike property, and compute paths of resource allocation which, in a society which consists of a representative agent for each generation beginning with the present one, optimizes an objective function that sustains growth in human welfare forever, for exogenously specified rates of growth, taken to include zero as one possibility.

We show that positive rates of growth in human welfare are possible, while the first generation experiences a utility level higher than the reference level. For growth rates very close to zero, the computed paths involve investments in knowledge at noticeably higher levels than in the past: the fraction of labor resources devoted to the creation of knowledge must be doubled, whereas the fractions of labor allocated to consumption and leisure are similar to those of the reference level.

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On the other hand, higher growth rates, while also feasible, require substantial increases in the fraction of labor devoted to education, together with moderate increases in the fractions of labor devoted to knowledge and the investment in physical capital. We test for the robustness of the model calibration, and find qualitatively similar results.

We now summarize what is new about our approach, and how it contrasts with the influential works of William Nordhaus (2008a) and Nicholas Stern (2007).

The society in our model consists of an infinite set of generations, each represented by a single agent. The agents’ utility function, and the set of feasible paths of resource allocation, are specified as follows.

- The representative agent’s utility – welfare, or quality-of-life function – has four arguments: (i) consumption of a produced commodity, (ii) educated leisure time, which is raw leisure valued by the agent’s level of education or skill, (iii) the quality of the biosphere at the time the agent lives, a public good, and (iv) the level, or stock, of human knowledge, a public good.

- There are three production sectors: commodity production uses as inputs skilled labor, capital, accumulated human knowledge, biospheric quality, and the level of GHG emissions permitted. The production of knowledge is purely labor intensive using only skilled labor and past knowledge (think corporate research and development, and university research). The education of children is purely labor intensive, using only the skilled labor of teachers.

- There are four conduits of intergenerational transmission: capital passes from one generation to the next, after investment and depreciation; knowledge passes in like manner, with depreciation; the stock of biospheric quality augmented by emissions of the present generation passes to the next; and adult teachers educate children who become skilled workers and consumers at the next date.

- One very important production function is not explicitly modeled: the evolution of biospheric quality from emissions. One might postulate a law of motion for this process similar to the laws of motion of capital and knowledge – that biospheric quality at date $t+1$ consists of biospheric quality at date $t$, partially rejuvenated by natural processes that absorb carbon dioxide, plus the impact of new emissions of GHGs. However, the scientific view on the nature of this law of motion is very much
in flux, and so we have elected not to imply a false precision by inserting such a law into our model. In place of doing so, we simply take the path of emissions and concomitant atmospheric concentration of carbon dioxide proposed by the IPCC AR 4 (2007, Working Group I, The Physical Science Basis, Chapter 10) as one which, they conjecture, will suffice to stabilize the atmospheric concentration of carbon at 450 ppm CO$_2$e, and we constrain our production sector not to emit more than is allowed on this path. That is to say: we do not optimize over possible paths of future emissions, because we believe the knowledge to do so does not exist at present.

- Our exercise is entirely normative: we choose the path to maximize the utility of the first generation, subject to guaranteeing a rate of growth of utility of $g$ for all future generations. We compute this path for various values of $g$. The path with $g = 0$ we call ‘pure sustainability optimization,’ as it sustains human welfare forever at the highest possible level. The paths with $g > 0$ we call ‘sustainable growth optimization’ We do not propose a rule for adjudicating among various values of $g$: but our calculations suggest that values of $g$ of 2% per (64% per generation) are more ethically attractive than the optimal path at $g = 0$.

- As our approach is purely normative, we do not propose an economic equilibrium model, nor do we attempt to predict what the path would be in the absence of policy (what is often called the business as usual path).

- Technological change is modeled by the presence of knowledge, accumulated through investment in R&D, as an input into commodity production. Thus knowledge can substitute for capital, labor, and emissions through the process of technological change.

What is the output of the model which interests us? First, we seek to understand what rates of growth of human welfare can be sustained, given the constraints on emissions recommended by the IPCC. Second, we wish to understand the trade-offs implied by choosing to grow at higher rates: for instance, it turns out to be feasible to support welfare growth of 64% per generation with our calibration, but the cost will be lower welfare than the first generation would enjoy under a 0% growth scenario. What is the magnitude of this trade-off? Third, we wish to understand how labor should be allocated among its four uses for various values of $g$: labor allocated to commodity
production, to educating children, to research and knowledge production, and to leisure. Should we radically re-allocate labor from its present uses?

We now contrast our approach those of with Nordhaus (2008a) and Stern (2007).

- Nordhaus (2008a) also carries out a normative exercise of maximizing an intergenerational social welfare function. He does not fix a path of emissions. Instead, he proposes a law of motion of biospheric degradation, and optimizes over not only the paths of consumption, investment, and capital, but also of emissions. As we note in Section 5.2 below, his solution paths entail, for the next two centuries, emission levels substantially higher than the IPCC-inspired paths that we adopt.

- The utility function of his representative agents consists only of consumption of a produced commodity. Accordingly, emissions and biospheric quality affect human welfare only indirectly, through their impact on production.

- Nordhaus proposes an exogenous path of technological change. There is no knowledge-production sector in his model. Neither is there an education sector in Nordhaus (2008a).

- Most importantly, the social welfare function in Nordhaus (2008a) is discounted utilitarian. He maximizes the discounted sum of generational utility levels, where the discount rate is calibrated from the rate of time impatience of existing consumers, calculated via the Ramsey equation.

- The Stern (2007) report does not carry out a full optimization exercise. It compares only two paths: ‘business as usual,’ against an alternative path that cuts back severely on emissions. The criterion used to compare these two paths is discounted utilitarian. But the objective differs from Nordhaus because Stern chooses a much smaller discount rate (larger discount factor) than Nordhaus. Rather than calibrating the discount rate from the Ramsey equation – and thus from the rate of impatience of market consumers – Stern (2007) discounts future utility only because future generations might not exist, due to a small probability, at each date, of the disappearance of the human species.

There are three principal differences between our work and that of Nordhaus and Stern.
Our objective is to sustain the growth of human welfare, at some specified rate of growth, rather than maximizing the discounted sum of generational utilities. We lack the space in the present paper to argue why we view our approach as superior: but we refer the reader to extended discussions of this matter in two accompanying papers, Llavador et al. (2009), and Roemer (in press). We ask the reader to note that sustaining growth is heard much more in both scientific and popular discussions than maximizing the sum of discounted utilities. While the latter has a long history in economic theory, it has little popular resonance. This, however, is not the main basis of our critique in the just mentioned papers.

We include four arguments in the utility function, not one. This is more realistic, we believe, and also provides more possibilities for substitution in order to maintain growth of human welfare. While Nordhaus (2008a) claims that his ‘consumption’ can be interpreted as including myriad goods, this is incorrect. For the production of different goods (leisure, education, knowledge) impact very differently upon biospheric quality through their emission of GHGs. Nordhaus’s aggregation would be valid only if all relevant goods impacted upon biospheric quality in the same way.

We do not optimize over paths of emissions, and we have explained our choice not to do so above.\(^2\)

Macroeconomists are used to arguing over what the discount rate should be: indeed, this is the main topic of disagreement between Nordhaus and Stern (see Section 5.3 below). A discount rate does not appear in our model, and the reader may wonder why – are we avoiding an important issue? The answer is that it is possible to insert a discount rate into our model, along the lines of Stern: the utility of future generations can be discounted because they may not exist. This topic is treated in Llavador et al. (2009). It turns out, however, that the discount rate has a very different impact on the outcome of optimization in our ‘sustainability of welfare growth’ approach than in the discounted-utilitarian approach.

Because of the difficulty of obtaining reliable global data, we have calibrated our model with US data only. Thus, the agents in our model must be interpreted as US residents. However, the IPCC emissions paths refer to global emissions, with concomitant atmospheric concentrations of

\(^2\) Not even in the sense of cost minimization conditional to a given stock or mitigation path, as do Lawrence Goulder and Koshy Mathai (2000) or OECD (2008, Ch. 7).
CO\textsubscript{2e}. We must therefore propose a way to allocate emissions to the United States which conforms to the IPCC global emissions path that we take as our constraint.

For each growth scenario we study, we calculate two optimal paths of resource allocation: the first assumes that the US continues to emit 24\% of global emissions forever, and the second assumes that US emits only its per capita share of global emissions forever. Obviously, the first is an optimistic, and the second a pessimistic, path as far as the welfare of US residents is concerned. These paths, we believe, give upper and lower bounds, respectively, on what US residents can expect in the political agreement that will eventually transpire among nations to allocate rights to emit GHGs to countries. That is: it is unreasonable to suppose that the US will emit more than its present share of GHGs in the future; and even if the US is allocated permits according to a global per capita share rule, it is almost surely the case that trading in permits will result in the US’s emitting more than its per capita share (hence, the second scenario provides a lower welfare bound). We have limited our exercise to using US data not due to an ethical parochialism, but solely in order to be able to calibrate a model with some confidence.

The paper is organized as follows. Section 2 presents the formal details of the model. Section 3 explains our strategy for calculating optimal paths. Section 4 presents the results. Section 5 discusses relation to the literature in more detail, and Section 6 summarizes and concludes.

2. Our approach

2.1. The utility function

A large segment of the literature (e. g., Nordhaus, 2008a) postulates an individual or generational utility function with the consumption of a single, produced good as its only argument (sometimes augmented by leisure time): Improvements in knowledge, education, and the environment are then important only in so far as they make possible the production of consumption goods with less labor time or capital.

In fact, both the consumption of goods and the availability of natural capital positively affect human welfare. Indeed, the spectacular increase of consumption in developed economies during the last century has undoubtedly provided a major welfare improvement (D. G. Johnson, 2000). But, in our view, two other factors have also had major impacts. First are the improvements in life expectancy, health status and infant survival, partly due to the rise in consumption, but to a large
extent due to medical discoveries, and their implementation by the public health system. Second is the improvement in literacy and, more generally, in the amount of education received by the average person, which has enhanced not only the productivity of labor but also utility: the contribution of leisure to utility increases as leisure time embodies higher levels of human capital, see Salvador Ortigueira (1999) and Martin Wolf (2007), as well as J. J. Heckman (1976) and Robert T. Michael (1972). In Wolf’s words:

“The ends people desire are, instead, what makes the means they employ valuable. Ends should always come above the means people use. The question in education is whether it, too, can be an end in itself and not merely a means to some other end – a better job, a more attractive mate or even, that holiest of contemporary grails, a more productive economy. The answer has to be yes. The search for understanding is as much a defining characteristic of humanity as is the search for beauty. It is, indeed, far more of a defining characteristic than the search for food or for a mate. Anybody who denies its intrinsic value also denies what makes us most fully human.”

Our approach follows the spirit of the Human Development Index produced by the United Nations Development Program, which considers three dimensions, namely (a) life expectancy, (b) education, and (c) consumption (GDP per capita). On the other hand, as we discuss in Section 2.2 below, the welfare or the consumption of a generation’s children is not an argument in the utility function.

The first argument in the utility function is consumption. But we emphasize other factors as well:

(i) Education, which modifies the value of leisure time to the individual;

(ii) Knowledge, in the form of society’s stock of culture and science, which directly increases the value of life (in addition to any indirect effects through productivity), via improvements in health and life expectancy, and because an understanding of how the world works and an appreciation of culture are intrinsic to human well-being,

(iii) An undegraded biosphere, which is valuable to humans for its direct impact on physical and mental health.

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3 Jim Oeppen and James Vaupel (2002, p. 1029) report that “female life expectancy in the record-holding country has risen for 160 years at a steady pace of almost 3 months per year.”
4 Increases in the human capital of the parents can also improve the quality of their child-rearing services, a component of the parents’ “leisure.”
5 This is captured in the Cost-Benefit literature on global warming by the computation of the so-called “noneconomic effects.”
Hence, consumption, educated leisure, the stock of human knowledge, and the quality of the biosphere are arguments in the utility function. The first two arguments are private goods, and the last two are public goods.

We abstract from all conflicts except for the intergenerational one and, accordingly, we assume a representative agent in each generation. We assume that a generation lives for 25 years, and we formally postulate the following utility function of Generation $t$, $t \geq 1$:

$$\tilde{\Lambda}(c_t, x_t, S_t^n, S_t^m) \equiv (c_t)^{\alpha_c} (x_t)^{\alpha_l} (S_t^n)^{\alpha_n} (\dot{S}_t^m - S_t^m)^{\alpha_m},$$

where the exponents are positive and normalized such that $\alpha_c + \alpha_l + \alpha_m + \alpha_n = 1$, and where:

- $c_t$ = annual average consumption per capita by Generation $t$;
- $x_t$ = annual average leisure per capita, in efficiency units, by Generation $t$;
- $S_t^n$ = stock of knowledge per capita, which enters Generation $t$’s utility function and production function, understood as located in the last year of life of Generation $t$;
- $S_t^m$ = total CO2 in the atmosphere above the equilibrium pre-industrial level, in GtC, which is understood as located in the last year of life of Generation $t$;\(^6\) and
- $\dot{S}_t^m$ = “catastrophic” level of CO2 in the atmosphere above the pre-industrial level.

2.2. Optimization programs: Sustainable utility levels and sustainable growth

We are concerned with human sustainability, which requires maintaining human welfare, rather than green sustainability, which may be defined as keeping the quality of the biosphere constant.\(^7\) This objective can be justified by appealing to the Maximin principle, see Roemer (1998, 2007). It can be argued, and this is Rawls’s position when justifying the (contemporaneous) “difference principle,” that it is the quality of life of each person that should enter the Maximin calculus, rather than subjective happiness, which generally includes the satisfaction that the individual derives from the welfare of other people, such as her children.

\(^6\) The preindustrial values for the CO2 stock are taken to be 595.5 GtC or 280 ppm. To convert our $S_t^m$ into CO2 ppm, add 595.5 to $S_t^m$ and multiply by 0.47. To convert a number of CO2 ppm into our $S_t^m$, subtract 280 from it and multiply by 2.13. The presence of the stock of CO2 in the utility function captures our view that environmental deterioration is a public bad in consumption (as well as in production), contrary to the modeling of Nordhaus (1994, 2008a) and Nordhaus and Joseph Boyer (2000), where it is only a public bad in production.

\(^7\) See Eric Neumayer (1999) and the articles collected in Geir Asheim (2007) for the analysis of the various notions of sustainability.
Maximizing the utility of the worst-off generation will often require the maximization of the utility of the first generation subject to maintaining that utility for all future generations, so that there is no utility growth after the first generation. Formally, the optimization program is of the following type.

**Pure Sustainability Optimization Program:**

\[
\max \Lambda \text{ subject to } (c_t)^{\alpha_x} (x_t^Y)^{\alpha_w} (S_t^m)^{\alpha_m} \left( \hat{S}_t^m - S_t^m \right)^{\alpha_\nu} \geq \Lambda, \quad t \geq 1,
\]

and subject to the feasibility conditions given by specific production relations, laws of motion of the stocks and resource constraints, and with the initial conditions given by the relevant stock values in the base year (2000).

At a solution of the Pure Sustainability Optimization Program, the path of the utility will typically be stationary, and it can be (at least asymptotically) supported by stationary paths in all the arguments of the utility function.

Alternatively, the planner may seek a positive rate of growth in the utility of future generations at the cost of reducing the utility of Generation 1. It is, however, not obvious how to justify sacrifices of the worst-off present generation for the sake of improving the already higher welfare levels of future ones.

One might argue that parents want their children to have a higher quality of life than they do. Thus, welfare growth might be supported by all parents over the Pure Sustainability Optimization solution. An alternative justification for altruism towards future generations would appeal to *growth as a public good*: we may feel justifiably proud of mankind’s recent gains in, say, extraterrestrial travel, or average life expectancy, and wish them to continue into the far future even at a personal cost.

Indeed, there is an asymmetry in the way we feel about contemporaneous vs. temporally disjoint inequality: a person in a poor country may not wish to sacrifice her utility for the sake of improving that of a person in a richer country, while at the same time be willing to make some sacrifices for the welfare of unrelated, yet-to-be born individuals who will as a consequence be richer than she.

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8 But not always: see Silvestre (2002).
9 Recall that we assume away intragenerational inequality, thereby depriving economic growth of a role in alleviating contemporaneous poverty. This important topic has high priority in our research agenda.
Assume that society wants to achieve a sustained rate $\rho$ of growth in future utility: instead of maximizing the utility of the worst-off generation, it aims at the maximization of the utility of the first generation, subject to the condition that utility subsequently grows at a given rate $\rho$ per generation. The optimization program then becomes:

**Sustainable Growth Optimization Program**

\[
\max \Lambda \quad \text{subject to:} \quad \left( c_i \right)^{\alpha_c} \left( x'_i \right)^{\alpha_e} \left( S^e_i \right)^{\alpha_e} \left( S^m_i - S^c_i \right)^{\alpha_m} \geq \left( 1 + \rho \right)^{t-1} \Lambda, \quad t \geq 1,
\]

for $\rho \geq 0$ given, and subject again to the feasibility and initial conditions.

Note that the Pure Sustainability Optimization Program can be written in this form by letting $\rho = 0$.

At a solution to this program, the utility grows at a constant rate, but it is impossible to have steady positive growth of all variables because of the finite capacity $\hat{S}^m$ of the biosphere.

### 2.3. Economic constraints

Feasible paths are characterized by **economic constraints** and by **environmental stock-flow relations**. We adopt the following economic constraints. Recall that $t = 1, 2, \ldots$ is measured in generations (25 years).

\[
f(x^e_i, S^k_i, S^n_i, e_i, S^m_i) = k_i (x^e_i)^{\theta_c} (S^k_i)^{\theta_k} (S^n_i)^{\theta_n} (e_i)^{\theta_e} (S^m_i)^{\theta_m} \geq c_i + i, \quad t \geq 1,
\]

with $\theta_c > 0, \theta_k > 0, \theta_n > 0, \theta_c + \theta_k + \theta_n = 1, \theta_e > 0, \theta_m < 0$,

(Aggregate production function $f$)

\[(1 - \delta^k)S^k_{t-1} + k_s n_i \geq S^k_i, \quad t \geq 1, \quad \text{(Law of motion of physical capital)}\]

\[(1 - \delta^n)S^n_{t-1} + k_s n_i \geq S^n_i, \quad t \geq 1, \quad \text{(Law of motion of the stock of knowledge)}\]

\[x^e_i + x^c_i + x^n_i + x'_i \equiv x_i, \quad t \geq 1, \quad \text{(Allocation of efficiency units of labor)}\]

\[k_s x^e_{t-1} \geq x_i, \quad t \geq 1, \quad \text{(Education production function)}\]

with initial conditions $(x^e_0, S^k_0, S^n_0)$, where $c_i, x'_i, S^e_i$ and $S^m_i$ have been defined in Section 2.1 above, and where:

- $x^e_i =$ average annual efficiency units of labor per capita devoted to the production of output by Generation $t$;
- $e_i =$ average annual emissions of CO$_2$ in GtC by Generation $t$. 
$S_t^k = \text{capital stock per capita available to Generation } t$;

$i_t = \text{average annual investment per capita by Generation } t$;

$x_t^n = \text{average annual efficiency units of labor per capita devoted to the production of knowledge by Generation } t$,

$x_t^e = \text{average annual efficiency units of labor per capita devoted to education by Generation } t$;

$x_t = \text{average annual efficiency units of time (labor and leisure) per capita available to Generation } t$.

We call emissions $e_t$ and concentrations $S_t^m$ \textit{environmental variables}, whereas the remaining variables are called \textit{economic}.

The following remarks compare our technology to some of those postulated in the growth literature.

\textbf{Remark 1.} The labor input in production, $x_t^e$, is measured in efficiency units of labor, which may be viewed as the number of labor-time units (“hours”) multiplied by the amount of human capital embodied in one labor-time unit (as is customary since Hirofumi Uzawa, 1965 and Robert Lucas, 1988). Hence, because we assume that $\theta_c + \theta_k + \theta_n = 1$, our production function displays decreasing returns to “capital” when construed to consist of physical and human capital. But returns would be constant if we broadened the notion of “capital” to include also the stock of knowledge.

\textbf{Remark 2.} We assume that the production of new knowledge requires only efficiency labor (dedicated to R&D, or to “learning by not doing”), but that knowledge depreciates at a positive rate. These assumptions are in line with a large segment of the growth literature.

\textbf{Remark 3.} Our education production function, $x_t = k_4 x_{t-1}^e$, states that the education of a young generation requires only efficiency labor of the previous generation. If we normalize to unity the total labor-leisure time available to Generation $t$, then $x_t$ can be interpreted as the amount of human capital per time unit in Generation $t$. Because our model is generational ($t$ is a generation), instead of being an infinitely lived consumer (for whom $t$ is just a moment in her life), our education production function cannot be interpreted in exactly the same manner as in many existing models of investment in human capital, which, in addition, are often cast in continuous time. More specifically, our formulation displays the following features.
(a) As in Uzawa (1965) and Lucas (1988), we do not include physical capital as an input in the production of education. This contrasts with Sergio Rebelo (1991) and Robert Barro and Xavier Sala-i-Martin (1999, p. 179). In the notation and wording of Barro and Sala-i-Martin, their “human capital production function” is,

\[ H = B[(1 - v)K]^\eta[(1 - u)H]^\eta - \delta H, \]

where \( H \) is the amount of human capital, \((1 - v)K \) is the amount of physical capital used in education, \((1 - u)\) is the fraction of human capital used in education, and \( B, \eta, \) and \( \delta \) are parameters, the last one being the human-capital depreciation factor.

(b) We interpret the labor input in the production of education as that of teachers, rather than students. This departs from the interpretations by Lucas (1988) and Rebelo (1991), but it agrees with the comments in Uzawa (1965) and Barro and Sala-i-Martin (1999), e. g., the latter write (p. 179) “… a key aspect of education [is that] it relies heavily on educated people as an input.”

(c) We see the education of a generation as a social investment, in line with Lucas’s (1988, p. 19) dictum “…a general fact that I will emphasize again and again: that human capital accumulation is a social activity, involving groups of people, in a way that has no counterpart in the accumulation of physical capital.” Also, we adopt a broad view of educational achievement, which in particular bestows the ability to adapt to new technologies, as emphasized by Claudia Goldin and Lawrence Katz (2008).

(d) Our education production function can be viewed as a generational version of (1) for the parameter values \( \eta = 0 \) and \( \delta = 1 \) (since, in our model, all adults die at the end of each date), obtaining:

\[ H - H_{t-1} = B[(1 - u)H_{t-1}] - H_{t-1}, \ i.\ e., \ H_t = B[(1 - u)H_{t-1}], \]

which is precisely our education production function under the notational correspondence \( H_t \leftrightarrow x_t, \)

\( (1 - u) \leftrightarrow \frac{x_{t-1}^c}{x_{t-1}} \) and \( B \leftrightarrow k_4 \).

2.4. Environmental stocks and flows

Anthropogenic greenhouse gas (GHG) emissions have caused atmospheric concentrations with no precedents in the last half a million years (see, e. g., Pierre Friedlingstein and Susan Salomen, 2005). The unparalleled behavior of GHG concentrations has motivated a growing
literature that tries to predict the relationship among the paths of emissions, concentrations and global temperature changes.

Following a large segment of literature, we focus on CO$_2$ emissions and concentrations. Recent climate research has revised upwards the persistence of the effects of GHG emissions. Haaron Kheshgi, Steven Smith and James Edmonds (2005, p. 213) emphasize that emitted CO$_2$ “is not destroyed in the atmosphere, but redistributed amongst the reservoirs that actively exchange carbon: plants and soils, oceans and the atmosphere.” They argue that “for CO$_2$ to approach a constant concentration over finite time, CO$_2$ emissions must peak and then gradually approach zero over 1,000+ years, regardless of the concentration level.” Alvaro Montenegro et al. (2007, p.1) argue that “higher levels of atmospheric CO$_2$ remain in the atmosphere than predicted by previous experiments, and the average perturbation lifetime of emissions is much longer than the 300-400 years proposed by other studies.” Based on new evidence on the behavior of ocean temperatures after increases in emissions, H. Damon Matthews and Ken Caldeira (2008) show that temperatures will be rising long after the CO$_2$ concentration in the atmosphere has been stabilized and that in order “to achieve atmospheric carbon dioxide levels that lead to climate stabilization, the net addition of CO$_2$ to the atmosphere from human activities must be decreased to nearly zero.” Similar conclusions are reached by Friedlingstein and Solomon (2005).

Most of the more recent and detailed physical models have no steady states, in the strict sense, with positive emissions. But if emissions are steady at low enough levels, then the stock of GHG eventually grows very slowly, experiencing minor increases in a scale of thousands of years. The effects of climate change on human welfare can then be substantially attenuated via mitigation (e. g., the construction of levees) and adaptation (e. g., moving North). The stocks of GHG are then said to be “stabilized” even though, strictly speaking, they are not constant in the very long run. Here we assume a constant “long term” value of the stock of GHG, where “constant” is a simplification of “stabilized,” and where the “long term” scale refers to a few hundreds, but not thousands, of years.

2.5. Our postulated paths for the CO$_2$ emissions and stocks

Because of the complexity of the climate models proposed and the lack of a canonical physical model of the current state of climatology, we shun false precision and do not attempt to

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11 The long-term effects of non-CO$_2$ GHG emissions have been addressed by Marcus Sarofim et al. (2005).
specify the set of feasible flow-stock sequences \([(e_i, S^m_i))_{i=1}^\infty\]. Accordingly, we do not try to compute optimal paths for emissions and the environmental stock. Instead, we adopt a simple path inspired by Meehl et al. (2007, Section 10.4), in particular by emission paths that lead to relatively low levels of stabilized concentrations of CO\(_2\) under the assumption of coupling between climate change and the carbon cycle.\(^{12}\) We choose the target stabilization level of 450 ppm (Meehl et al., 2007, Section 10.4, Figure 10.21(a)) and, conservatively, the path of coupled emissions for the Hadley model, as in C. D. Jones et al. (2006, Figure 10.21.(c)).

These paths involve increasing emissions in the near future, and drastically reduced emissions in the more distant future. We adopt this general pattern, but we simplify the path by postulating only three levels of emissions and stock, which average over each generation the abovementioned lifetime paths for emissions, while taking as stock values those dated at the end of the life of the generation. Hence, the Meehl et al. (2007) analysis justifies the feasibility of our paths given the initial values \((\bar{e}_{2000}^W, \bar{S}^m_{2000}) = (6.58, 177.1)\) at year 2000 (World Resources Institute, 2009), where \(e^W\) stands for annual world emission in GtC.\(^{13}\) Our postulated (emission, atmospheric stock) pairs are:

\[
(e^1_1, S^m_1) = (6.97, 303) \text{ for Generation 1,}
\]

\[
(e^2_2, S^m_2) = (4.43, 354) \text{ for Generation 2,}
\]

and \(e^W_t, S^m_t = (e^W_{t-1}, S^m_{t-1}) = (0.4, 363)\) for Generation \(t, t \geq 3\),

see Figure 2 in Section 6.1 below for a graphical representation. See also the first and last columns of Table 1.

---

\(^{12}\) The growth of the atmospheric CO\(_2\) induces a climate change that affects the carbon cycle. In their words (p. 789) “There is an unanimous agreement among the models that future climate change will reduce the efficiency of the land and ocean carbon cycle to absorb anthropogenic CO\(_2\), essentially owing to a reduction in the land carbon uptake.”

\(^{13}\) We take \(S^m_{2000} = 177.1\) GtC (or 83 ppm) as the year 2000 atmospheric CO\(_2\) concentration above pre-industrial level (of approximately 595.5 GtC in 1850) from the CAIT Indicator Framework Paper (World Resources Institute, WRI, 2009). Total annual world emissions from energy (fossil fuels and cement) are 6.58 GtC. Once we include CO\(_2\) emissions from land use change (7.62 GtCO\(_2\)) and from other Kyoto gases (9.72 GtCO\(_2\)e), total emissions (41.42 GtCO\(_2\)e) are consistent with the 42 GtCO\(_2\)e total GHG emissions in 2000 reported in the Stern Review (page 170).
### Table 1. Our postulated paths for the environmental variables

<table>
<thead>
<tr>
<th>Year</th>
<th>World CO₂ Emissions (GtC)</th>
<th>US CO₂ Emissions (GtC)</th>
<th>Scenario 1</th>
<th>Scenario 2</th>
<th>Stock of CO₂ in (World) Atmosphere (GtC)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2000</td>
<td>$e^w_{2000} = 6.58$</td>
<td>$e^US_{2000} = 1.6$</td>
<td></td>
<td></td>
<td>$S^m_{2000} = 177.1$</td>
</tr>
<tr>
<td>Gen. 1</td>
<td>$e^w_1 = 6.97$</td>
<td>$e^{US1}_{1} = 1.64$</td>
<td></td>
<td>$e^{US2}_{1} = 0.25$</td>
<td>$S^m_{1} = 303$</td>
</tr>
<tr>
<td>Gen. 2</td>
<td>$e^w_2 = 4.43$</td>
<td>$e^{US1}_{2} = 1.05$</td>
<td></td>
<td>$e^{US2}_{2} = 0.14$</td>
<td>$S^m_{2} = 354$</td>
</tr>
<tr>
<td>$t \geq 3$</td>
<td>$e^w_\infty = 0.4$</td>
<td>$e^{US1}_\infty = 0.09$</td>
<td></td>
<td>$e^{US2}_\infty = 0.01$</td>
<td>$S^m_\infty = 363$</td>
</tr>
</tbody>
</table>

Our choices for $(e^w_1, S^m_1)$, $(e^w_2, S^m_2)$ and $(e^w_\infty, S^m_\infty)$ imply that, in 2075, the concentration of CO₂ in the atmosphere will be of 450 ppm (this corresponds to our value of $S^m_\infty = 363$ GtC in the atmospheric stock of CO₂ beyond the preindustrial stock, see Footnote 6 above). The algorithm described in Section 3 below motivates our choice of a two-generation interval to reach the target stabilization level.

The path that we choose entails, relative to the 1990 world emission levels, a 24% reduction in 2025 and a 93% reduction in 2050. For comparison, the American Clean Energy and Security Act HR 2454 of 2009 (US House of Representatives, Waxman-Markey Bill) aims at a 1% reduction in 2020 and an 80% reduction in 2050. The United Nations Human Development Report 2007/2008 (Overview, p. 29) recommends, for developed countries, between a 20% and a 30% for 2020, and an 80% for 2050. And the OECD Environmental Outlook Policy Package (OECD 2008, Ch. 20) actually entails an increase in world GHG emissions for 2050.

As noted in the introduction, we have calibrated our economic model with US data due to the difficulty of obtaining reliable world data (see the following section for the numerical values). But the IPCC emissions paths refer to world emissions. We must therefore allocate emissions to

---

14 The act targets take 2005 as the reference year. From the CAIT, World Resources Institute, we take US emissions in 2005 to be 20% higher than in 1990.
the United States in line with the IPCC-inspired global emissions path that we adopt. To do so, we consider two alternative scenarios.

The first scenario maintains the share of US emissions at its year-2000 share. The US accounted for 1.6 GtC in that year, representing 24% of all energy (fuel and cement) emissions (World Resources Institute, 2009). Hence, our Scenario 1 levels of future US emissions are given by the 24% of \((e_{1}^{W}, e_{2}^{W}, e^{W*})\): they are displayed in the second column in Table 1.

The second scenario assumes that the US emits its per capita share of the global emissions \(e_{1}^{W}, e_{2}^{W}\) and \(e^{W*}\). We use the United Nations projections for world population, and compute the emissions per capita along for \(e_{1}^{W}, e_{2}^{W}\) and \(e^{W*}\) as \((e_{1}^{Wpc}, e_{2}^{Wpc}, e^{Wpc}) = (0.87, 0.48, 0.04)\) (in tC per capita). Keeping the US population constant at year 2000 level (284,257 thousands), we obtain the Scenario 2 values of total US displayed in the third column in Table 1.

These two scenarios represent upper and lower bounds for the welfare of the US representative agent: we conjecture that, even if emission permits were distributed on a per capita basis to the various countries, the US would end up purchasing rights permits from other countries. Hence, Scenario 2 provides a lower bound on the welfare of the representative US citizen.

2.6. The calibration of parameters and initial values.

As noted, we draw on US data in order to calibrate the parameters of the utility function, output and education production functions and the laws of motion for physical capital and knowledge, as well as the benchmark, year-2000 values of economic stocks and flows. Appendix 1 below details our calibration procedures, which yield the values displayed in tables 2 and 3. The values in Table 2, as well as those for \(k_{2000}, n_{2000}\) and \(x_{2000}\) from Table 3, will enter the computational algorithm described in the following section.

3. Computational strategy and algorithm

Our computational strategy is based on the Ray Optimization Theorem below, in the spirit of turnpike theory: see our companion paper Llavador et al. (2009) for a turnpike theorem in a simpler

---

15 We take the value \(e_{2000} = 1.6\) GtC as the US annual CO2 emissions from energy (fossil fuels and cement).

16 World population for 2000 is 6,124,123 thousand persons. Projections establish a population of 8,010,509 thousands for 2025, stabilized at 9,191,287 thousands in 2050.
Parameter | Value
---|---
$\alpha_c$ | 0.32
$\alpha_l$ | 0.65
$\alpha_n$ | 0.02
$\alpha_m$ | 0.01
$k_1$ | 16.328
$k_2$ | 13.118
$k_3$ | 649.34
$k_4$ | 35.45
$\theta_c$ | 0.67
$\theta_k$ | 0.28
$\theta_n$ | 0.06
$\theta_m$ | -0.0075
$\theta_e$ | 0.091
$\delta^k$ | 0.787
$\delta^n$ | 0.787
$\hat{S}^m$ | 781.55

Table 2. Calibrated values for functional parameters

<table>
<thead>
<tr>
<th>Variable</th>
<th>Value</th>
<th>Units</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\bar{S}^k_{2000}$</td>
<td>73.65</td>
<td>Thousands of 2000 dollars per capita.</td>
</tr>
<tr>
<td>$\bar{S}^n_{2000}$</td>
<td>15.64</td>
<td>Thousands of 2000 dollars per capita.</td>
</tr>
<tr>
<td>$\bar{x}_{2000}$</td>
<td>1.396</td>
<td>1950-efficiency units per capita.</td>
</tr>
<tr>
<td>$\bar{x}^c_{2000}$</td>
<td>0.047</td>
<td>1950-efficiency units per capita.</td>
</tr>
<tr>
<td>$\bar{x}_{2000}$</td>
<td>0.396</td>
<td>1950-efficiency units per capita.</td>
</tr>
<tr>
<td>$\bar{x}^n_{2000}$</td>
<td>0.023</td>
<td>1950-efficiency units per capita.</td>
</tr>
<tr>
<td>$\bar{x}^l_{2000}$</td>
<td>0.931</td>
<td>1950-efficiency units per capita.</td>
</tr>
<tr>
<td>$\bar{c}_{2000}$</td>
<td>27.78</td>
<td>Thousands of 2000 dollars per capita.</td>
</tr>
<tr>
<td>$\bar{I}_{2000}$</td>
<td>6.83</td>
<td>Thousands of 2000 dollars per capita.</td>
</tr>
<tr>
<td>$\bar{Y}_{2000}$</td>
<td>34.61</td>
<td>Thousands of 2000 dollars per capita.</td>
</tr>
</tbody>
</table>

Table 3. Benchmark year-2000 magnitudes
model. Consider a pair \((e^*, S^{m*})\) such that the constant sequence \((e^*, S^{m*})\) is an environmentally feasible flow-stock path, and the following optimization program.

**Program** \(E[\rho, e^*, S^{m*}]\). Given \((\rho, e^*, S^{m*})\), Max \(\Lambda_1\) subject to

\[
\begin{align*}
c_i^{\alpha_i} (x_i')^{\alpha_i} (S_i^m)^{\alpha_i} (\hat{S}^m - S^{m*})^{\alpha_i} & \geq \Lambda_1 (1 + \rho)^{t_i - 1}, \quad t_i \geq 1, \\
k_i (x_i')^{\theta_i} (S_i^e)^{\theta_i} (S_i^m)^{\theta_i} & \geq c_i + i_i, \quad t_i \geq 1, \\
(1 - \delta_i)S_{i-1} + k_i x_i' & \geq S_i^e, \quad t_i \geq 1, \\
(1 - \delta_n)S_{i-1} + k_n x_i' & \geq S_i^n, \quad t_i \geq 1, \\
x_i + x_i' + x_i'' & \equiv x_i, \quad t_i \geq 1, \\
k_n x_i' & \geq x_i, \quad t_i \geq 1,
\end{align*}
\]

with initial conditions \((x_0^e, S_0^e, S_0^n)\).

Recall that \(\rho\) is the rate of growth of the utility per generation. It will be convenient to denote by \(g\) the rate of growth of the economic variables, again per generation.

**Theorem 1: Ray Optimization Theorem.** Assume constant returns to scale in production in the sense that \(\theta_c + \theta_k + \theta_n = 1\). Given \((g, e^*, S^{m*}) \in [0, k - 1] \times \mathbb{R}_+ \times (0, \hat{S}^m)\), there is a ray

\[
\Gamma(g, e^*, S^{m*}) \equiv \{(x^e, S^e, S^n) \in \mathbb{R}^3 : (S^e, S^n) = x^e (q^k (g, e^*, S^{m*}), q^n (g))\},
\]

such that if

\[
(x_0^e, S_0^e, S_0^n) \in \Gamma(g, e^*, S^{m*}), \quad (x_0^e, S_0^e, S_0^n) \neq 0,
\]

then the solution path to Program \(E[\rho, e^*, S^{m*}]\) satisfies:

(i) \((x^e, S^e, S^n) = (1 + g) (x_0^e, S_0^e, S_0^n)\), \(t_i \geq 1\), and hence \((x^e, S^e, S^n) \in \Gamma(g, e^*, S^{m*})\), \(t_i \geq 0\);

\[
\begin{align*}
c_i & = p^e (g) q^k (g, e^*, S^{m*}) x_0^e, \\
\end{align*}
\]

(ii) \(x_i' = v^e (g) q^n(g) x_0^e\),

\[
\begin{align*}
x_i'' & = v^n (g) q^n(g) x_0^e, \\
x_i' & = v^e (g) q^n(g) x_0^e;
\end{align*}
\]

(iii) \((c_i, i_i, x_i', x_i'', x_i^e) = (1 + g)^{-1} (c_0, i_0, x_0', x_0'', x_0^e)\), \(t_i \geq 1\).

Utility grows at rate \(\rho\), were \(1 + \rho = (1 + g)^{1 - \alpha_i}\), and all other variables grow at rate \(g\), except for emissions and concentrations, which remain constant at \((e^*, S^{m*})\).

**Proof.** Appendix 2, where the various proportionality factors \((q, p, v)\) are computed in terms of the parameters of the model.
In particular, it is important to observe that, for \( g = \rho = 0 \), whenever the initial endowments \((x_0^*, S_0^*, S_0^m)\) lie in \( \Gamma(0, e^*, S^m^*) \), the solution to Program \( E[0, e^*, S^m^*] \) is stationary over time.

We conjecture that a turnpike theorem, analogous to the one in Llavador et al. (2009), is true for Program \( E[\rho, e^*, S^m^*] \) for any \( g \), and so, if we begin with an endowment vector off the ray \( \Gamma(g, e^*, S^m^*) \), then the optimal solution will converge to the ray \( \Gamma(g, e^*, S^m^*) \). Hence, in the long run, the solution will be almost a steady-state path. Motivated by this conjecture, we now construct feasible paths which begin at the actual year-2000 endowment values \((\bar{x}_{2000}^e, \bar{S}_{2000}^k, \bar{S}_{2000}^m)\) and reach the ray \( \Gamma(g, e^*, S^m^*) \) in two generations, taking as given the values \((e_1^{USj}, S_1^m)\), \((e_2^{USj}, S_2^m)\) and \((e^{USj^*}, S^m^*)\), \( j = 1, 2 \), reported in Table 1.\(^{17}\)

More precisely, for various rates of growth \( \rho \geq 0 \) of the utility (or associated rates of growth \( g \) of the variables), we construct feasible paths \((\Lambda_1, \Lambda_2, \ldots)\) such that the ratio \( \frac{\Lambda_t}{\Lambda_{t-1}} \) of utility growth experienced by the later generations \( t \geq 2 \) is \( 1 + \rho \), and analyze the implications of these sustained growth factors for the utility \( \Lambda_1 \) of Generation 1. A reference level of utility is the one determined by the year-2000 values of the relevant variables, to be denoted \( \Lambda_0 \).

We proceed in two steps. First, we solve the optimization problem for (endogenous) initial conditions guaranteeing that the optimal solution is a steady state (i.e., all economic variables, not including the environmental ones, grow at the same, predetermined rate.) Second, we go from the historical initial conditions to the steady state path in two generations, while keeping the rate of growth of the utility for all generations after the first one at the predetermined rate.

The utility of Generation \( t \) is given by \( c_t^{\alpha_t} (x_t^j)^{\alpha_t} (S_t^m)^{\alpha_m} (S_t^m - S_t^m)^{\alpha_n} \). If all variables (except biospheric quality) grow at a rate \( g \), then the utility will grow at rate \( \rho \) where \( 1 + \rho = (1 + g)^{1-\alpha_n} \). A balanced growth solution relative to our choice requires three growth rates:

\(^{17}\) Inspired by IPCC AR4 (2007), we have computed paths in which carbon concentrations converge to the stabilized level in two generations. However, our optimization program could be run for slower convergence paths, with convergence in three or more generations, at the cost of additional complexity in computation. We have also tried to maximize the utility of Generation 1 subject to its reaching the ray. Note that Generation 1’s investment in knowledge (which affects the utility of Generation 1 both directly and indirectly through production) and Generation 1’s investment in physical capital (which affects the utility of Generation 1 only indirectly through production) create intergenerational public goods. It turns out that, even for a zero-growth target, when Generation 1 maximizes its own utility subject to the stock proportionality dictated by the ray, it invests so heavily as to make the utility of the future generations higher than its own, a feature formally similar to the one discussed in Silvestre (2002). The resulting path yields therefore an unnecessarily low value. It is for this reason that we choose Generation 2 as the first one that has stocks on the ray.
$g$ for the variables $(S^n, x^n, x^e, x^c, x^f)$,

$\gamma$ for the variables $i, c$ and $S^k$,

$\rho$ for the utility.

But $\rho$ and $\gamma$ are functions of $g$: so there is one independently chosen growth rate. For $\theta_c + \theta_k + \theta_n = 1$, we have that $g = \gamma$.

We apply the following two-step algorithm for the chosen $(\rho, e_1, S^m_1, e_2, S^m_2, e^*, S^{m*})$.

**Step 1.** For an arbitrary $x_2^c$, solve the following program.

**Program $G[x_2^c]$**. Given $(\rho, e_1, S^m_1, e_2, S^m_2, e^*, S^{m*})$ and $x_2^c$, Max $\Lambda_1$ subject to

\[
c_1 a^c (x_1^c)^{\alpha_n} (S^m_1)^{\alpha_n} (S^n - S^m_1)^{\alpha_n} \geq \Lambda_1,
\]

\[
c_2 a^c (x_2^c)^{\alpha_n} (S^m_2)^{\alpha_n} (S^n - S^m_2)^{\alpha_n} \geq (1 + \rho)\Lambda_1,
\]

\[
(x_2^c, S^k_2, S^n_2) \in \Gamma(g, e^*, S^{m*}),
\]

\[
k_1 (x_1^e)^{\theta_0} (S^k_1)^{\theta_0} (S^n_1)^{\theta_0} (S^m_1)^{\theta_0} \geq c_1 + i_1,
\]

\[
k_1 (x_2^e)^{\theta_0} (S^k_2)^{\theta_0} (S^n_2)^{\theta_0} (S^m_2)^{\theta_0} \geq c_2 + i_2,
\]

\[
(1 - \delta^k) S^k_0 + k_2 i_1 \geq S^k_1,
\]

\[
(1 - \delta^k) S^k_1 + k_2 i_2 \geq S^k_2,
\]

\[
(1 - \delta^a) S^n_0 + k_3 x_1^n \geq S^n_1,
\]

\[
(1 - \delta^a) S^n_1 + k_3 x_2^n \geq S^n_2,
\]

\[
k_4 x_0^c \geq x_1^c + x_2^n + x_1^n + x_1^c,
\]

\[
k_4 x_1^c \geq x_2^c + x_2^n + x_1^n + x_2^c,
\]

for the initial conditions $(x_0^c, S_0^k, S^n_0) = (\bar{x}_0^c, \bar{S}_0^k, \bar{S}_0^n)$ as given in Table 3.

**Step 2.** Note that the utility of Generation 3, and of all subsequent generations, is determined by $x_2^c$. By trial and error, we locate the value of $x_2^c$ with the property that, at the solution to Program $G[x_2^c]$, the utility of Generation 3 equals $(1 + \rho)^2 \Lambda_1$. Note that then the utility of Generation $t$, $t \geq 4$, is $(1 + \rho)^{t-3}$ times the utility of Generation 3 (by Theorem 1), and that, by the second constraint of Program $G[x_2^c]$, the utility of Generation 2 is $(1 + \rho)\Lambda_1$. Hence, the utility of Generation $t$ is $(1 + \rho)^{t-1} \Lambda_1$, for all $t \geq 1$. 
Appendix 3 writes the solution to Program $G(x^r)$ as a system of 14 equations in the 14 endogenous variables $(\Lambda_1, c_1, x^1_1, x^1_2, x^2_1, c_2, x^2_2, x^n_1, i_1, i_2, S^k, S^n)$, which is then reduced to a system of seven equations in seven unknowns. Then, using Mathematica, we compute the numerical solution paths to Program $G(x^r)$ for our calibrated parameter values, and adjust $x^r$ so that the utility of Generation 3 equals $(1 + \rho)^2 \Lambda_1$, implying, as noted above, that the utility of Generation $t$ is $(1 + \rho)^{t-1} \Lambda_1$, for all $t \geq 1$. We perform this calculation for three sustained growth rates of the utility, namely $\hat{\rho} = 0.00$ (no growth), $\hat{\rho} = 0.01$ and $\hat{\rho} = 0.02$, where $\hat{\rho}$ is the rate of growth of the utility expressed in \textit{per annum} terms, with corresponding rates of growth per generation (defined by $\rho = (1 + \hat{\rho})^{25} \equiv \hat{\rho}$) equal to $\rho = 0.00$, $\rho = 0.28$ and $\rho = 0.64$, respectively.

4. Results

Tables 4-6 describe the obtained paths of utility and of the economic variables in the two scenarios for US emissions. The first rows in the tables display the year-2000 values (see Table 3 above) repeated in each table to facilitate comparison. Some of the information in these tables is summarized in Table 7 and depicted in Figure 1. Recall (see Section 2.5 above) that we postulate a rather conservative path of total GHG emissions aimed at stabilizing GHG concentrations at 450 ppm, and consider two scenarios: Scenario 1 $\left( e^{US} = 0.24 \times e^{World} \right)$, in which the US is responsible for 24% of all emissions (its share of total emissions in 2000); and Scenario 2 $\left( e^{US\ per\ capita} = \frac{e^{World\ per\ capita}}{100} \right)$, in which total emissions are allocated on a per capita basis.

Our computations yield the following results.

\textbf{Result 1. Utility can be sustained forever at a level substantially higher than the year-2000 reference level (26\% higher in Scenario 1, or 17\% higher in Scenario 2).}

See the first column (of the top half) of Table 4. The utility of the first generation jumps to 25.82\% (resp. 16.87\%) above that of the year-2000 reference level in the first (resp. second) scenario on US emissions, and stays there forever. This fact is illustrated by the two horizontal lines in both graphs in Figure 1: the lower, dotted line, with ordinate equal to 1, corresponds to the year-2000 reference level, while the continuous horizontal line with circular dots gives the sustained level of utility for all generations $t \geq 1$. As expected, the lower US emissions of Scenario 2 yield smaller increases in the utility of the US representative agents.
\[
\begin{array}{cccccccccc}
\Lambda_t & \Lambda_{t-1} & c_t & \frac{c_t}{c_0} & \frac{c_t}{c_{t-1}} & i_t & S^k_t & S^n_t \\
\hline
\hline
\text{Scenario 1} \left( e^{US} = 0.24 \times e^{World} \right) \\
2000 & 1.00 & 1.00 & 27.78 & 1 & - & 6.83 & 73.65 & 15.64 \\
1 & 1.2582 & 1.2582 & 41.0709 & 1.4784 & 1.4784 & 14.46 & 205.32 & 48.93 \\
2 & 1.2582 & 1.0000 & 38.2724 & 1.3777 & 0.9319 & 7.76 & 145.55 & 54.84 \\
3 & 1.2582 & 1.0000 & 30.6556 & 1.1035 & 0.8010 & 8.73 & 145.55 & 54.84 \\
4 & 1.2582 & 1.0000 & 30.6556 & 1.1035 & 1.0000 & 8.73 & 145.55 & 54.84 \\
\hline
\text{Scenario 2} \left( \frac{e^{US}}{\text{per capita}} = \frac{e^{World}}{\text{per capita}} \right) \\
2000 & 1.00 & 1.00 & 27.78 & 1 & - & 6.83 & 73.65 & 15.64 \\
1 & 1.1687 & 1.1687 & 32.4637 & 1.1686 & 1.1686 & 11.08 & 161.03 & 48.82 \\
2 & 1.1687 & 1.0000 & 29.8476 & 1.0744 & 0.9194 & 5.94 & 112.23 & 55.61 \\
3 & 1.1687 & 1.0000 & 23.6389 & 0.8509 & 0.7920 & 6.73 & 112.23 & 55.61 \\
4 & 1.1687 & 1.0000 & 23.6389 & 0.8509 & 1.0000 & 6.73 & 112.23 & 55.61 \\
\end{array}
\]

Table 4. \( \hat{\rho} = 0.00 \) (sustainable utility, no growth)
<table>
<thead>
<tr>
<th>Gen</th>
<th>$\Lambda_t$</th>
<th>$\Lambda_{t-1}$</th>
<th>$c_t$</th>
<th>$c_{t-1}$</th>
<th>$i_t$</th>
<th>$S^k_t$</th>
<th>$S^n_t$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2000</td>
<td>1.00</td>
<td>1.00</td>
<td>27.78</td>
<td>1</td>
<td>-</td>
<td>6.83</td>
<td>73.65</td>
</tr>
<tr>
<td>1</td>
<td>1.2476</td>
<td>1.2476</td>
<td>40.717</td>
<td>1.4657</td>
<td>1.4657</td>
<td>14.32</td>
<td>203.50</td>
</tr>
<tr>
<td>2</td>
<td>1.6000</td>
<td>1.2824</td>
<td>48.8354</td>
<td>1.7579</td>
<td>1.1994</td>
<td>11.11</td>
<td>189.09</td>
</tr>
<tr>
<td>3</td>
<td>2.0518</td>
<td>1.2824</td>
<td>50.3244</td>
<td>1.8115</td>
<td>1.0305</td>
<td>15.46</td>
<td>243.09</td>
</tr>
<tr>
<td>4</td>
<td>2.6313</td>
<td>1.2824</td>
<td>64.7000</td>
<td>2.3290</td>
<td>1.2857</td>
<td>19.88</td>
<td>312.54</td>
</tr>
</tbody>
</table>

Scenario 1 ($e^{US} = 0.24 \times e^{World}$)

<table>
<thead>
<tr>
<th>Gen</th>
<th>$\Lambda_t$</th>
<th>$\Lambda_{t-1}$</th>
<th>$c_t$</th>
<th>$c_{t-1}$</th>
<th>$i_t$</th>
<th>$S^k_t$</th>
<th>$S^n_t$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2000</td>
<td>1.00</td>
<td>1.00</td>
<td>27.78</td>
<td>1</td>
<td>-</td>
<td>6.83</td>
<td>73.65</td>
</tr>
<tr>
<td>1</td>
<td>1.1588</td>
<td>1.1588</td>
<td>32.1824</td>
<td>1.1585</td>
<td>1.1585</td>
<td>10.97</td>
<td>159.58</td>
</tr>
<tr>
<td>2</td>
<td>1.4860</td>
<td>1.2824</td>
<td>38.0923</td>
<td>1.3789</td>
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Scenario 2 ($e^{US}$ per capita = $e^{World}$ per capita)

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<tr>
<th>Gen</th>
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<th>$\Lambda_{t-1}$</th>
<th>$c_t$</th>
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Table 5. $\rho = 0.01$ (1% annual growth or 28% generational growth)
<table>
<thead>
<tr>
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<th>$\Lambda_t / \Lambda_0$</th>
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<th>$c_t / c_0$</th>
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**Scenario 1** \( e^{US} = 0.24 \times e^{World} \)

<table>
<thead>
<tr>
<th>Gen</th>
<th>$\hat{\rho}$ = 0.02 ((2% \text{ annual growth or } 64% \text{ generational growth}))</th>
<th>$\Lambda_t / \Lambda_0$</th>
<th>$\Lambda_t / \Lambda_{t-1}$</th>
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**Scenario 2** \( e^{US}_{\text{per capita}} = e^{World}_{\text{per capita}} \)

<table>
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<tr>
<th>Gen</th>
<th>$\hat{\rho}$ = 0.02 ((2% \text{ annual growth or } 64% \text{ generational growth}))</th>
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**Scenario 2** \( e^{US}_{\text{per capita}} = e^{World}_{\text{per capita}} \)

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**Scenario 2** \( e^{US}_{\text{per capita}} = e^{World}_{\text{per capita}} \)

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<th>Gen</th>
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Table 6. $\hat{\rho} = 0.02$ \((2\% \text{ annual growth or } 64\% \text{ generational growth})\)
<table>
<thead>
<tr>
<th>Scenario 1 ( (e_{US}^{US} = 0.24 \times e_{World}^{World}) )</th>
<th>( \frac{\tilde{\Lambda}_1(\hat{\rho})}{\Lambda_0} )</th>
<th>( \frac{\tilde{\Lambda}_1(0) - \tilde{\Lambda}_1(\hat{\rho})}{\tilde{\Lambda}_1(0)} )</th>
</tr>
</thead>
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<tr>
<td>( \hat{\rho} = 0.00 ) (Sustainable, No growth)</td>
<td>1.258</td>
<td>0.000</td>
</tr>
<tr>
<td>( \hat{\rho} = 0.01 ) ( \rho = 0.28 )</td>
<td>1.248</td>
<td>0.0085 = 0.85%</td>
</tr>
<tr>
<td>( \hat{\rho} = 0.02 ) ( \rho = 0.64 )</td>
<td>1.234</td>
<td>0.0192 = 1.92%</td>
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</table>

<table>
<thead>
<tr>
<th>Scenario 2 ( (e_{US}^{US} = e_{World}^{World}) )</th>
<th>( \frac{\tilde{\Lambda}_1(\hat{\rho})}{\Lambda_0} )</th>
<th>( \frac{\tilde{\Lambda}_1(0) - \tilde{\Lambda}_1(\hat{\rho})}{\tilde{\Lambda}_1(0)} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \hat{\rho} = 0.00 ) (Sustainable, No growth)</td>
<td>1.169</td>
<td>0.000</td>
</tr>
<tr>
<td>( \hat{\rho} = 0.01 ) ( \rho = 0.28 )</td>
<td>1.159</td>
<td>0.0085 = 0.85%</td>
</tr>
<tr>
<td>( \hat{\rho} = 0.02 ) ( \rho = 0.64 )</td>
<td>1.146</td>
<td>0.0193 = 1.93%</td>
</tr>
</tbody>
</table>

*Table 7.* The utility of the first generation (first column) relative to the year-2000 reference level \( \Lambda_0 \), and the sacrifice of the first generation to sustain subsequent positive growth rates (second column). The tildes denote the solution for the corresponding variable as a function of \( \hat{\rho} \).
Utility of generations $t = 1$ to $5$ for alternative rates $\hat{\rho}$ of per annum growth in utility. All variables grow at a rate slightly higher than $\hat{\rho}$, with the exception of emissions and the stock of the biosphere, which follow the path described in Table 1 above.
**Result 2.** For the low US emissions of Scenario 2, the sustainable steady state consumption is lower than the year-2000 reference consumption level; the increase in utility stated in Result 1 is then due to the increases in the stock of knowledge and in the quality of leisure.

The fourth column (of the top half) of Table 4 shows that steady-state consumption falls to a 85% of its year-2000 value. This decrease is more than compensated by the improvements in knowledge, which augments by a factor of 3.56 (see 8th column of the top half of Figure 4), and in the quality of leisure, which improves by a 35% (see first column in the bottom half of Table 4: the fraction of time devoted to leisure in fact slightly decreases, see the 1st, 5th and 9th columns of the bottom half of Table 4). In other words, the Pure Sustainability Optimization path requires a degree of sacrifice in material consumption, offset by improvements in other sources of the quality of life.

**Result 3.** Moderate growth rates can be achieved at the cost of a small reduction in the utility of the first generation, which stays well above the year 2000 reference level.

A tradeoff between the utility of the first generation and the subsequent growth rates must indeed be expected. But our analysis shows that its magnitude is quite small: Generation 1’s sacrifice for the sake of a higher growth rate is tiny for reasonable growth rates.

Table 7 (obtained from Tables 5 and 6) displays the relevant ratios. As stated in Result 1, utility can be sustained forever while the utility of the first generation is 1.26 (Scenario 1) and 1.17 (Scenario 2) times the year-2000 reference level. The second and fourth rows of Table 7 show that, in order to subsequently maintain a 1% growth rate per year (28% per generation), the utility of the first generation must be about 0.85% lower than the no-growth value. In other words, a maintained growth rate of 28% per generation can be reached at the cost of a less than 1% reduction of the utility of the first generation relative to the sustainable (no growth) path.

Similarly, the third and fifth rows of Table 7 show that, in order to subsequently maintain a 2% utility growth rate per year (64% per generation), the utility of the first generation would be about 1.9% lower than the sustainable, no-growth value. In other words, a maintained growth rate of 64 % per generation can be reached at the cost of a less than 2% reduction of the utility of the first generation relative to the no growth path.

Figure 1 shows the utility paths computed under the different growth targets. Note that they stay well above the year-2000 reference level. It is not possible at the scale of the graph to
distinguish among the three values of the utility of the first generation (for annual growth rates of 0, 1% and 2%, respectively), all clustered close to the 1.25 value.

How are these utility paths implemented? Labor time is, in the reference year 2000, allocated to the various ends as follows (see, e. g., the last four columns of Table 4, and the disaggregation of output into consumption and investment in columns 3 and 6 therein):

| Fraction allocated to education: | 0.0333 |
| Fraction allocated to the production of consumption: | 0.2150 |
| Fraction allocated to investment in physical capital: | 0.0683 |
| Fraction allocated to the creation of knowledge: | 0.0167 |
| Fraction allocated to leisure: | 0.6667 |

Table 8 indicates how these fractions should be modified in the proposed solutions. We observe the following features.

<table>
<thead>
<tr>
<th>Education</th>
<th>Knowledge</th>
<th>Investment in Phys. Capital</th>
<th>Consumption</th>
<th>Leisure</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \tilde{x}_E^r(\hat{\rho}) / \tilde{x}_E^u(\hat{\rho}) )</td>
<td>( \tilde{x}_K^r(\hat{\rho}) / \tilde{x}_K^u(\hat{\rho}) )</td>
<td>( \tilde{\ell}_E^r(\hat{\rho}) + \tilde{\ell}_E^u(\hat{\rho}) )</td>
<td>( \tilde{c}_E^r(\hat{\rho}) + \tilde{c}_E^u(\hat{\rho}) )</td>
<td>( \tilde{x}_L^r(\hat{\rho}) / \tilde{x}_L^u(\hat{\rho}) )</td>
</tr>
</tbody>
</table>

### Scenario 1 \( e^{US} = 0.24 \times e^{World} \)

| \( \hat{\rho} = 0 \) (No growth) | 0.848 | 2.552 | 1.334 | 0.932 | 0.964 |
| \( \hat{\rho} = 0.01 \) | \( \rho = 0.282 \) | 1.099 | 2.529 | 1.322 | 0.924 | 0.956 |
| \( \hat{\rho} = 0.02 \) | \( \rho = 0.64 \) | 1.418 | 2.498 | 1.306 | 0.914 | 0.945 |

### Scenario 2 \( e^{US\ per\ capita} = e^{World\ per\ capita} \)

| \( \hat{\rho} = 0 \) (No growth) | 0.855 | 2.546 | 1.297 | 0.934 | 0.966 |
| \( \hat{\rho} = 0.01 \) | \( \rho = 0.282 \) | 1.108 | 2.522 | 1.284 | 0.926 | 0.958 |
| \( \hat{\rho} = 0.02 \) | \( \rho = 0.64 \) | 1.430 | 2.492 | 1.268 | 0.916 | 0.947 |

**Table 8.** Comparison between steady state and year-2000 values of the allocation of labor for the various growth rates. Again, the tildes denote the solution for the corresponding variable as a function of \( \hat{\rho} \).
**Result 4.** The most important change required by the implementation of the proposed utility paths is the (more than) doubling of the reference fraction of labor devoted to the creation of knowledge, whereas the fractions of labor allocated to consumption and to leisure are similar to those of the reference year 2000.

The largest change displayed in Table 8 occurs in the fraction allocated to knowledge, which must be about twice (2.14, 2.26 or 2.34) the year-2000 reference level. The fraction of labor time devoted to the production of the consumption good is slightly lower. We also observe a slight decrease in leisure time relative to the year 2000 reference values.

As might be expected, higher growth rates require higher fractions of labor dedicated to the various forms of investment (education, knowledge and physical capital), and lower fractions dedicated to consumption and leisure. But as just noted, the fractions of labor dedicated to consumption and leisure are not very sensitive to the growth rate, whereas it turns out that the fraction of labor devoted to education increases rapidly with the target growth rate. Table 9, obtained by dividing the values in rows 2 and 3 of Table 8 by those of the first row (and subtracting 1), illustrates. This yields the following result.

**Result 5.** Higher growth rates require substantial increases in the fraction of labor devoted to education (of the order of a 30% increase for each additional 1% of annual growth), together with moderate increases in the fractions of labor devoted to knowledge and the investment in physical capital (of the order of a 5% increase for each additional 1% of annual growth). Higher growth rates also require minor decreases in the amount of labor-time devoted to the production of consumption goods and to leisure (of the order of a 2% decrease for each additional 1% of annual growth).

We have tested for the robustness of our results in several ways. In addition to our calibrated values, we also considered a lower $\hat{S}^m = 650$ ppm and a higher $\hat{S}^m = 750$ ppm, as well as lower and higher values for $\frac{\alpha_m}{\alpha_c}$, $\theta_c$, and $\theta_m$: we have obtained qualitatively similar results. Unsurprisingly, the sustainable level of utility increases with the catastrophic level of carbon concentration in the atmosphere ($\hat{S}^m$) and with the elasticities of output to emissions ($\theta_c$) and to concentration ($\theta_m$), and decreases with the relative weight of the environment in utility ($\alpha_m$). Yet our qualitative conclusions continue to hold under these changes. Finally, we have also considered different values of
parameters associated with the educational technology \( (k_e) \), and we have found that we can sustain forever levels of life-quality above the 2000 reference value, even for much lower values of \( k_e \).\(^{18}\)

\[
\frac{\ddot{x}_e(\hat{\rho})}{\ddot{x}_e(0)} / \frac{\ddot{x}_e(\hat{\rho})}{\ddot{x}_e(0)} - 1
\]

<table>
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<tr>
<th>Education</th>
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<th>Investment in Phys. Capital</th>
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<th>Leisure</th>
</tr>
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<tbody>
<tr>
<td>( \frac{\ddot{x}_e(\hat{\rho})}{\ddot{x}_e(0)} / \frac{\ddot{x}_e(\hat{\rho})}{\ddot{x}_e(0)} - 1 )</td>
<td>( \frac{\ddot{x}_k(\hat{\rho})}{\ddot{x}_k(0)} / \frac{\ddot{x}_k(\hat{\rho})}{\ddot{x}_k(0)} - 1 )</td>
<td>( \frac{\ddot{c}_i(\hat{\rho})}{\ddot{c}_i(0)} / \frac{\ddot{c}_i(\hat{\rho})}{\ddot{c}_i(0)} - 1 )</td>
<td>( \frac{\ddot{c}_c(\hat{\rho})}{\ddot{c}_c(0)} / \frac{\ddot{c}_c(\hat{\rho})}{\ddot{c}_c(0)} - 1 )</td>
<td>( \frac{\ddot{x}_l(\hat{\rho})}{\ddot{x}_l(0)} / \frac{\ddot{x}_l(\hat{\rho})}{\ddot{x}_l(0)} - 1 )</td>
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<table>
<thead>
<tr>
<th>Scenario 1 ( (e^{US} = 0.24 \times e^{World}) )</th>
<th>Scenario 2 ( (e^{US} = e^{World}) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \hat{\rho} = 0.01 ) ( \rho = 0.282 )</td>
<td>( \hat{\rho} = 0.02 ) ( \rho = 0.64 )</td>
</tr>
<tr>
<td>( 29.631% ) above ( ) no growth</td>
<td>( 67.325% ) above ( ) no growth</td>
</tr>
<tr>
<td>( 0.928% ) below ( ) no growth</td>
<td>( 2.108% ) below ( ) no growth</td>
</tr>
<tr>
<td>( 0.951% ) below ( ) no growth</td>
<td>( 2.160% ) below ( ) no growth</td>
</tr>
<tr>
<td>( 0.850% ) below ( ) no growth</td>
<td>( 1.931% ) below ( ) no growth</td>
</tr>
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<td>( 29.633% ) above ( ) no growth</td>
<td>( 67.329% ) above ( ) no growth</td>
</tr>
<tr>
<td>( 0.938% ) below ( ) no growth</td>
<td>( 2.131% ) below ( ) no growth</td>
</tr>
<tr>
<td>( 0.987% ) below ( ) no growth</td>
<td>( 2.242% ) below ( ) no growth</td>
</tr>
<tr>
<td>( 0.855% ) below ( ) no growth</td>
<td>( 1.942% ) below ( ) no growth</td>
</tr>
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**Table 9.** The sensitivity of the fractions of labor resources devoted to each activity with respect to the target growth rate.

5. Relation to the literature

5.1. Discounted utilitarianism

A large fraction of the literature on climate change adopts the discounted-utilitarian normative criterion. But we find discounted utilitarianism ethically unacceptable, at least for the low pure-time discount factors typically used, which put a weight on the utility of future generations much lower than that of the present generation. The only ethical justification for putting a lower weight on the welfare of future generations in the utilitarian calculus should be based on a positive probability of extinction of mankind. As argued in the Stern Review, this rationale would perhaps support a discount rate of \( \hat{\delta} = 0.001 = 0.1\% \) per annum, associated with a

\(^{18}\) For example, even for an unrealistically low value of \( k_e \) equal to 31, utility can be sustained forever at a level 9,8% higher than the year 2000 reference level under scenario 1, and at a level 2% higher under Scenario 2.
Of course, a rigorous development of this idea requires an explicit model of uncertainty: see Appendix 4 below and Llavador et al. (2009).

The low time discount factors frequent in the literature are mathematically expedient, because they make the sum of discounted utilities finite. But only with extremely low factors this would be the case for the economy modeled and calibrated here, as it can be argued as follows.

Denote by $\hat{P}[e^*, S^{m*}]$ the set of feasible paths according to the constraints of Program $E[\rho, e^*, S^{m*}]$ of Section 3 above, for some fixed endowment vector $(x_0^e, S_0^k, S_0^n)$. (This set is independent of the value of $\rho$.) The associated Discounted-Utilitarian Program, with a discount factor of $\varphi$, is:

$$
\text{Program } DU[\varphi, e^*, S^{m*}]: \max \sum_{t=1}^{\infty} \varphi^{t-1} \Lambda_t(\pi) \quad \text{s.t. } \pi \in \hat{P}[e^*, S^{m*}],
$$

where $\Lambda_t(\pi)$ is the utility at date $t$ along the path $\pi$. We have:

**Corollary to Theorem 1.** Program $DU[\varphi, e^*, S^{m*}]$ diverges if $\varphi k_{4-\alpha}^{1-\alpha} > 1$.

**Proof.** By Theorem 1, for any $g < k_4 - 1$ there is a ray $\Gamma(g, e^*, S^{m*})$ such that, from any initial endowment vector on this ray, the balanced growth path where the economic variables grow at rate $g$ is feasible. For any $g < k_4 - 1$, we can construct a path which, in a finite number of dates, moves from the given endowment vector $(x_0^e, S_0^k, S_0^n)$ to some point on this ray. We then complete the path by appending the balanced growth path just referred to. Again by Theorem 1, the utility grows by a factor of $1 + \rho$ at each date, after the initial section of the path, where $1 + \rho = (1 + g)^{1-\alpha}$. But $g$ may be chosen so that $1 + g$ is arbitrarily close to $k_4$. Hence, the terms of the discounted-utilitarian objective will grow by a factor arbitrarily close to $\varphi k_{4-\alpha}^{1-\alpha}$: in particular, $g$ can be chosen so that this factor is greater than one, by the premise, which proves the corollary. ■

---

19 A discount rate $\delta$ defines a discount factor $\varphi$ by $\varphi = \frac{1}{1 + \delta}$. Hence, a discount rate of 0.1% yields a discount factor of 0.999001 per annum. See sections 5.2 and 5.3 below for the discussion of other discount factors used in the literature.

20 See Llavador et al. (2009) for a discussion of how a discounted utilitarian would choose paths when the discounted-utilitarian program diverges,
If we take \(1 - p = 0.975\) per generation of 25 years, as does the Stern Review, then the discounted utilitarian program will diverge as long as \((1 - p)k_4^{1-\alpha_\omega} > 1\). But this inequality surely holds with our calibration of the parameters.

It is notable that the ‘power’ of the technology, in the sense of whether or not Program \(DU[\varphi, e^*, S^*]\) diverges, depends only on the technological parameter \(k_4\), associated with the educational technology, not on any parameters associated with the other two production functions. In a simpler model than the one here, studied in Llavador et al. (2009), we attempt to explain in an intuitive way why this is the case, and we shall not repeat that argument here. The fact depends upon the constant-returns technology, that labor is the single input in the production of skilled labor, and upon the constant-returns utility function. In particular, the last fact requires that leisure be measured in quality units, an assumption we strongly defend. As long as the assumption that the educational technology uses only educated labor as an input is approximately true, we believe this result is robust. We are reminded of Goldin and Katz (2008), who argue that the power of the American growth performance in the twentieth century was fundamentally due to universal education.

5.2. Nordhaus’s optimization

Nordhaus (2008a,b) proposes particular paths for CO₂ emissions, CO₂ concentrations and consumption per capita based on an optimization program with objective function

\[
\sum_{t=1}^{T} L_t \frac{1}{1-\eta} (c_t)^{1-\eta} \frac{1}{(1+\delta)^t},
\]

where \(L_t\) is the number of people in generation \(t\).\(^{21}\) He calls the \(\delta\) and \(\eta\) of (2) “central” and “unobserved normative parameters,” reflecting “the relative importance of the different generations.” (Nordhaus 2008a, p. 33, 60).\(^{22}\) Note that the Pure Sustainability Optimization objective function of our Section 2.2 above could be viewed as a limit case of (2) for \(L_t = 1, \delta = 0\) and \(\eta \to \infty\). Nordhaus

\(^{21}\)The objective function is given in Nordhaus (2008a, p. 205), with each period \(t = 1, 2, \ldots\) understood as a decade (instead of our 25-year generations). His notation is different. The optimization is numerically solved by the General Algebraic Modeling System (GAMS) program, see Nordhaus (2008b).

\(^{22}\)The parameter \(\eta\) could also be interpreted, following the classical utilitarians and the discounted utilitarian approach discussed in the previous section, as an index of the concavity of a common, cardinal and interpersonally unit-comparable utility function displaying decreasing marginal utility, see Roemer (1998) for definitions. But Nordhaus (2008a,b) does not adopt this interpretation.
(2008a) chooses $\eta = 2$ and $\delta = (0.015)^{10}$, corresponding to a per year rate of $\hat{\delta} = 0.015$.\(^{23}\) Appendix 5 below comments on Nordhaus’s (2008a) objective function and on his calibration of its parameters.

The paths for emissions and concentrations proposed as optimal by Nordhaus differ markedly from the ones that we postulate: figures 2(a) (emissions) and 2(b) (concentrations) illustrate. Recall that we take as given a conservative path that drops emissions to very low levels by 2050 and stabilizes atmospheric CO$_2$ concentration at about 450 ppm by 2050. In striking contrast, Nordhaus (2008a, b) proposes as “optimal” a path where emissions and concentrations keep increasing past the end of the 21$^{\text{st}}$ century. Nordhaus (2008a, b) proposed values for 2100 are about 11 GtC in emissions, with concentrations at 586.4 ppm at 2100 and at a peak of about 680 ppm in 2180.

In light of the recent climate science research, we view Nordhaus’s (2008a, b) “optimal” emission and concentration figures as excessively high, likely to bring about irreversible changes in temperature and unavoidable negative impacts in the welfare of future generations.

A striking feature of Nordhaus (2008a) is that the path for per capita consumption (his only variable in the individual utility function) is virtually identical (at least for the 21$^{\text{st}}$ century) in the “optimal” and in the “baseline” (laissez faire) paths, see his Figure 5.9. Yet he claims (p. 82) that the value of the objective function at the “optimal” solution is 3.37 trillions of 2005 US$ higher than at the baseline solution. We conjecture that this puzzle may be partially explained by population growth, which increases the value of the objective function for a given level of consumption per capita, together with minute differences in consumption per capita. Because of the little difference between the optimal and nonoptimal paths of consumption per capita, we conjecture that his rate of growth in consumption per capita is basically driven by his postulated exogenous growth in total factor productivity.

### 5.3. Cost-Benefit analysis: The Stern Review

Cost-Benefit analysis underpins the recommendations of the Stern Review, in turn based on the Third Assessment Report of the United Nation’s Intergovernmental Panel on Climate Change (TAR IPCC, 2001) and on Christopher Hope (2006). The Stern Review does not attempt to solve an optimization program: it is rather a cost-benefit analysis arguing that the “costs of inaction are larger than costs of action.” Assuming a path of growth for the GDP, and starting from a Business as Usual (laissez-faire) hypothesis on the path of GHG emissions, it considers alternative policies that reduce

\(^{23}\) The latter is half the value adopted in Nordhaus and Boyer (2000), see Nordhaus (2008a, p. 50).
emissions in the present, and eventually stabilize GHG in the atmosphere. The review argues that, properly discounted, the benefits of strong, early action on climate change outweigh the costs. It should be noted that discount rates have different roles in Cost-Benefit Analysis and in discounted-utilitarianism optimization. Discounted utilitarianism (see Section 5.1) uses the pure time discount rate $\delta$ to weight the utilities of the various generations in the utilitarian maximand, whereas Cost-Benefit Analysis uses the consumption discount rate $\delta + \eta \tilde{g}$ to evaluate the changes in future consumption streams due to a particular (marginal) investment project, relative to a reference consumption path that exogenously grows at a rate $\tilde{g}$. The project passes the Cost Benefit test if the discounted sum of the consumption streams is positive. As noted above, the Stern Review uses a pure time discount rate of $\hat{\delta} = 0.001$ (based on the survival justification), together with $\eta = 1$ and $\tilde{g} = 0.013$ (1.3 % per annum), yielding a consumption discount rate of 0.014. Its commentators suggest higher consumption discount rates (Arrow, 2007, Nordhaus, 2007, Martin Weitzman, 2007: see the debate in the Postscripts to the Stern Review available at www.sternreview.org.uk, as well as the issue of World Economics 7(4), October-December 2006, and the subsequent Simon Dietz et al., 2007).²⁴

Because the Stern Review does not solve an optimization program, its recommendations are in principle open to the objection, voiced by the critics of the Review, that the consumption discount rate should reflect the rates of return of the available investment alternatives: even if, using a consumption discount rate of 0.014, carbon emission reductions pass the Cost-Benefit test, future generations could conceivably be better off if the current generation avoided incurring the costs of GHG reductions and invested instead in other intergenerational public goods. In defense of the Review, Dietz et al. (2007, p. 137) argue that “it is hard to know why we should be confident that social rates of return would be, say, 3% or 4% into the future. In particular, if there are strong climate...

²⁴ Nordhaus discounts the utility of future generations by the time-rate of discount that he deduces for today’s market consumer, from the Ramsey equation, which he takes to be $\delta = .015$ per annum. This leads to a discount factor applied to the utility of those alive a century from now of $(\frac{1}{1+\delta})^{100} = (\frac{1}{1.015})^{100} = 0.225$. Stern discounts the utility of those a century from now (who may not exist) according to the probability of extinction of the human species; he applies a discount factor of $(1 - p)^4 = (0.975)^4 = 0.904$. If we adopt Stern’s probability-of-extinction, we do not discount the utility (utility) of those a century from now at all: that is, our discount factor applied to the utility of those a century from now is unity.
Comparison of paths for the environmental variables proposed by Nordhaus (2008a,b) with the ones postulated in the present paper. The paths for Nordhaus “Optimal” are computed by running the program GAMS with data provided in Nordhaus (2008b). The curve labeled “Optimal” of Figure 5-6 in Nordhaus (2008a) displays emissions only for the period 2005-2105, where they coincide with those of Figure 2(a) here (except that there the emissions are per decade, and here per year). Similarly, the curve labeled “Optimal” of Figure 5-7 in Nordhaus (2008a) displays concentrations only for the period 2005-2205, where they coincide with those of Figure 2(b) here.
change externalities, then social rates of return on investment may be much lower than the observed
private returns on capital over the last century, on which suggestions of a benchmark of 3% or 4%
appear to be based.”

As we have shown, the discounted utilitarian program with the Stern Review’s discount
factor diverges on the set of feasible paths that we have proposed in this article. Because the Stern
Review only calculates discounted utility for a small number of generations, it need not address this
issue. This again shows the limitations of the cost-benefit method.

Our approach is in a sense dual to Cost-Benefit analysis. The latter takes as given a path for
the economic variables, and recommends a path for the environmental variables (based on a cost-
benefit criterion in the spirit of discounted utilitarianism). We, on the contrary, take as given a path
for the environmental variables, and recommend paths for the economic variables (based on the
criteria of sustainable utility and sustainable growth).

6. Summary and conclusions

Our analysis departs from the literature in three dimensions: (a) the concept of the utility, (b)
social welfare criteria, and (c) method.

For (a), we adopt a comprehensive notion of utility, in the spirit of the Human Development
Index, that emphasizes the following three factors in addition to the conventional consumption and
leisure.

(i) Education, which modifies the value of leisure time to the individual, besides
enhancing her productivity;

(ii) Knowledge, in the form of culture and science, which directly improves the living
experience, besides raising total factor productivity; and

(iii) The quality of the environment, which, because of the importance of climate change,
we interpret as depending on the concentration of greenhouse gases in the atmosphere.

For (b), we consider two criteria. First, Pure Sustainability Optimization, which aims
maximizing the utility that can be sustained for all generations. Second, Sustainable Growth
Optimization, where we fix positive rates of growth, with the justification that growth has the
character of a public good, and maximize the utility of the first generation subject to achieving the
given, constant rate of utility growth for all subsequent generations. These objectives stand in sharp
contrast to the conventional criterion of maximizing the discounted sum of utilities, which we find ethically unjustifiable, at least for the discount factors typically used.

As for (c), our method is inspired by optimization, but, given the current uncertainties in climate science, we do not attempt to compute an optimal path for environmental variables: we take instead as given a conservative path for the environmental variables, and propose paths for the economic variables based on the criteria of Pure Sustainability Optimization and Sustainable Growth Optimization. Ideally, for the Pure Sustainability Optimization Program, we would like to approach paths where all variables are stationary, whereas for the Sustainable Growth Optimization Program we would like to approach balanced-growth paths, where all variables grow at the same rate. But we cannot confidently adopt a reasonably simple model of emission-stock interaction. In addition, our formulation does not allow the quality of the atmosphere to improve without limit. Accordingly, our computations fix emissions and concentrations at levels that allow for stabilization after two generations. The resulting dynamic optimization programs defy explicit analytical solutions, and our approach has been computational. We have devised computational algorithms inspired by the turnpike property for constructing feasible and desirable, although not necessarily optimal, paths in the more complex and interesting models.

In more detail, we have adopted a simplified path for world emissions and concentrations that is based on the more elaborate paths proposed in the IPCC 2007 report aiming at stabilizing the concentration of CO$_2$ in the atmosphere at 450 ppm. Our simplified version assumes that we jump to a steady state in two generations, after which emissions are maintained at a very low level and the concentration of CO$_2$ in the atmosphere is stabilized. We have calibrated our economic model with US data, and consider two scenarios for the path of future US CO$_2$ emissions, which imply upper and lower bounds on the utility of US representative generational agents. We have then computed solutions for the economic variables, by an algorithm that mimics the turnpike method.

Our main result is the feasibility of sustaining utility levels higher than the year 2000 reference value, even when maintaining a positive rate of growth for all successive generations. Not surprisingly, higher rates of sustained growth require a lower utility for the first generation, but the tradeoff is small, and the first generation reaches a utility higher than the reference value for reasonable rates of growth.

\[25\] In our units, 363 GtC above preindustrial levels.
Pure Sustainability Optimization maximizes the utility level sustained for all generations, and corresponds to a zero rate of growth. Interestingly, for the scenario of low US emissions its steady state consumption is *lower* than the year-2000 reference consumption level; the increase in utility relative to the year-2000 reference value is then due to the increases in the stock of knowledge and in the quality of leisure.

Achieving this kind of human sustainability under the postulated environmental path requires particular kinds of behavior for the economic variables. The most important change is doubling the fraction of labor resources devoted to the creation of knowledge, whereas the fractions of labor allocated to consumption and leisure are similar to those of the reference year 2000.

On the other hand, higher growth rates require substantial increases in the fraction of labor devoted to education, together with moderate increases in the fractions of labor devoted to knowledge and the investment in physical capital.
APPENDIX 1. CALIBRATIONS

We interpret that generations live for 25 years. In this appendix, flow variables are typically defined as per year averages, and it is understood that stocks are located in the last year of life of a generation. The calibrated values that we obtain are reported in Section 2.6 above.

A1.1. Variables

\( S^k_t \) = capital stock available to Generation \( t \) (in thousands of dollars per capita).

\( S^n_t \) = stock of knowledge available to Generation \( t \) (in thousands of dollars per capita).

\( S^m_t \) = CO\(_2\) concentration in the atmosphere above the equilibrium pre-industrial level at the end of Generation \( t \)’s life (in GtC).

\( x_t \) = average annual efficiency units of time (labor and leisure) available to Generation \( t \) (in efficiency units per capita).

\( x'^e_t \) = average annual labor devoted to education by Generation \( t \) (in efficiency units per capita).

\( x'^c_t \) = average annual labor devoted to the production of output by Generation \( t \) (efficiency units per capita).

\( x'^l_t \) = annual average leisure by Generation \( t \) (in efficiency units per capita).

\( x'^n_t \) = average annual labor devoted to the production of knowledge by Generation \( t \) (in efficiency units per capita).

\( c_t \) = annual average consumption by Generation \( t \) (in thousands of dollars per capita).

\( i_t \) = average annual investment by Generation \( t \) (in thousands of dollars per capita).

\( e_t \) = average annual U.S. emissions of CO\(_2\) from fuel and cement in GtC by Generation \( t \) (in GtC).

A1.2. Parameters

\( \alpha_j \) = exponents of the utility function for \( j \in \{c \text{ (consumption)}, l \text{ (leisure)}, n \text{ (stock of knowledge)}, \text{ and } m \text{ (quality of the biosphere)}\} \).

\( k_1 \) = parameter of the production function \( f \).

\( k_2 \) = parameter of the law of motion of capital.

\( k_3 \) = parameter of the law of motion of the stock of knowledge.

\( k_4 \) = parameter of the education production function.
\( \theta_j \) = exponents of the inputs in the production function \( f \) for \( j \in \{ c \) (labor), \( k \) (stock of capital), \( n \) (stock of knowledge), \( e \) (emissions of CO₂), \( m \) (atmospheric carbon concentration) \}. 

\( \delta^k \) = depreciation rate of the stock of capital (per generation). 

\( \delta^n \) = depreciation rate of the stock of knowledge (per generation). 

\( \hat{S}^m \) = catastrophic level of carbon concentration in the atmosphere above the equilibrium pre-industrial level (in GtC). 

\( \rho \) = generational rate of growth of the utility. 

\( \hat{\rho} \) = annual rate of growth of the utility (\( \rho = (1 + \hat{\rho})^{25} \)).

### A1.3. Functions

Utility function (utility): 

\[ \hat{\Lambda}(c_t, x_t^j, S_t^n, S_t^m) \equiv (c_t)^{\alpha_c} (x_t^j)^{\alpha_j} (S_t^n)^{\alpha_n} (\hat{S}_t^m - S_t^m)^{\alpha_m}. \]

Production function: 

\[ f(x_t^c, S_t^k, S_t^n, e_t, S_t^m) \equiv k_t (x_t^c)^{0_1} (S_t^k)^{0_k} (S_t^n)^{0_n} (e_t)^{0_e} (S_t^m)^{0_m}, \quad \theta_c + \theta_k + \theta_n = 1. \]

Law of motion of physical capital: 

\[ S_t^k \leq (1 - \delta^k) S_{t-1}^k + k_2 l_t. \]

Law of motion of the stock of knowledge: 

\[ S_t^n \leq (1 - \delta^n) S_{t-1}^n + k_3 x_t^n. \]

Education production function: 

\[ x_t \leq k_4 x_{t-1}^e. \]

### A1.4. The calibration of the utility function

We take the exponent of leisure to be twice that of consumption (\( \alpha_l = 2 \alpha_c \)) and calibrate \( \alpha_n/\alpha_c = 0.05 \) as the average ratio of expenditure in knowledge (R&D expenditure plus investment in computer components and software) over expenditure in consumption during the period 1953-2000.\(^{26}\)

Next, we calibrate the ratio \( \alpha_m/\alpha_c \) by the Stern Review (2007) statement that a 5°C increase in the global temperature over the pre-industrial level would imply a health related damage equivalent

\(^{26}\) The data on R&D are derived from Research and Development in Industry, Academic Research and Development Expenditures, Federal Funds for Research and Development, and the Survey of Research and Development Funding and Performance by Nonprofit Organizations (National Science Foundation, 2003). Data on public investment in software are constructed taking the value of public investment in equipment and software (U.S. Bureau of Economic Analysis 2007) and assuming the same share of software in private and public investment.
to a 5% loss of global GDP (page x). We can read the statement of the Stern Review as saying that a 5% decrease in consumption is equivalent to suffering an atmospheric CO₂ concentration of \( \hat{S}^m \), yielding

\[
(0.95)\alpha_c (x^c)^{\alpha_c} (S^n)^{\alpha_n} (\hat{S}^m - S^n)^{\alpha_n} = (c)^{\alpha_c} (x^c)^{\alpha_c} (S^n)^{\alpha_n} (\hat{S}^m - \hat{S}^m)^{\alpha_n},
\]

that is,

\[
(0.95)\alpha_c (\hat{S}^m - S^n)^{\alpha_n} = (\hat{S}^m - \hat{S}^m)^{\alpha_n}.
\]

Taking logs,

\[
\alpha_c \ln(0.95) = \alpha_m \left( \ln\left(\hat{S}^m - \hat{S}^m\right) - \ln\left(\hat{S}^m - S^n\right) \right).
\]

We take a 5ºC increase in temperature to be associated with CO₂ equivalent (CO₂e) concentrations of 1470 GtC (Stern 2007, Figure 2 in page v). Because we only consider CO₂ emissions (which account for 84% of all GHG) and we compute values above pre-industrial level (595.5 GtC), we adopt the value

\[
\hat{S}^m = \frac{1470}{1.16} - 595.5 = 671.64 \text{ GtC}.
\]

We consider that an increase in temperature of 6º-8ºC (relative to pre-industrial level) would have catastrophic impacts. We take this temperature increases to be associated with CO₂ equivalent concentrations of 750 ppm (or 1597.5 GtC), the lower bound of the studies reported in the Stern Review (2007, p.12). As before, adjusting for all gases and subtracting pre-industrial levels, we obtain

\[
\hat{S}^m = \frac{1597.5}{1.16} - 595.5 = 781.55.
\]

It follows that

\[
\frac{\alpha_m}{\alpha_c} = \frac{\ln 0.95}{\ln(781.55 - 671.64) - \ln(781.55 - 177.1)} = 0.03.\]

Finally, we normalize \( \alpha_c + \alpha_f + \alpha_m + \alpha_n = 1 \), and obtain the values reported in Table 1 of the main text.

---

27 This is also in line with Nordhaus and Boyer (2000) who estimate a total cost (market and non-market) of between 9% and 11% of global GDP for a 6ºC warming (as quoted in Stern, 2007, p. 148).

28 The Stern Review consistently associates catastrophic consequences to temperature increases of 6-8ºC, like, for example, sea level rise threatening major world cities (including London, Shanghai, New York, Tokyo and Hong Kong), entire regions experiencing major declines in crop yields and high risk of abrupt, large scale shifts in the climate system (Figure 2 in page v), and catastrophic major disruptions and large-scale movements of population (Table 3.1 in p. 57).

29 As a reference, the US currently devotes approximately 2% of its gross domestic product to all forms of environmental protection.
A1.5. The calibration of the production function

We calibrate the production function

\[ f(x^c, S^k_t, S^n_t, e_t, S^m_t) \equiv k_t(x^c)\theta^0 (S^k_t)^\theta (S^n_t)^\theta (e_t)^\theta (S^m_t)^\theta, \]

in the following inputs: first the more usual labor, physical capital and knowledge, to which we add the environmental stock and emissions. We assume constant returns to scale in the first three inputs, i.e., \( \theta_c + \theta_k + \theta_n = 1 \). We construct time series for the stocks of physical capital, knowledge, and human capital, see sections A3.6-8 below. We take the labor income share to be 2/3, and compute the average share of physical capital and knowledge in the total stock of capital for the period 1960-2000, corresponding to 5/6 and 1/6, respectively. Hence, \( \theta_c = 2/3 \, , \, \theta_k = 5/18 \, \, \text{and} \, \theta_n = 1/18 \), representing the income share of each input.

We calibrate \( \theta_e = 0.091 \) as the “elasticity of output with respect to carbon services” from RICE99 in Nordhaus and Boyer (2000).

For the calibration of \( \theta_m \), the elasticity of output to the CO2 concentration in the atmosphere, we assume that doubling the CO2 concentration from pre-industrial levels would increase temperature by 2.5ºC (Stern, 2007, p.7),\(^{30}\) and that a 2.5ºC increase in temperature is associated with a 1.5% loss of total GDP (Nordhaus and Boyer, 2000, p.91). Hence,

\[ \theta_m = \frac{\% \Delta y}{\% \Delta T} = \frac{\% \Delta y}{\% \Delta S^m} = \frac{\% \Delta y}{\% \Delta T} \frac{\% \Delta S^m}{\% \Delta S^m} = -\frac{0.015}{2} = -0.0075, \]

where \( y \) is GDP per capita and \( T \) is global temperature.

We compute \( k_t \) as the TFP of the US economy calibrated to year 2000 values.\(^{31}\)

\[ k_t = \frac{y^\text{US}_t}{(S^c_{2000})^{\theta_c} (S^k_{2000})^{\theta_k} (S^n_{2000})^{\theta_n} (e_{2000})^{\theta_e} (S^m_{2000})^{\theta_m}} \]

\[ = \frac{34.61}{0.3955^{0.67} \times 73.62^{0.28} \times 15.54^{0.06} \times 1.6^{0.091} \times 177.1^{-0.0075}} = 16.328 \]

---

\(^{30}\) The Stern Review asserts that temperature would increase 1.5ºC-4.5ºC (if we consider feedback effects) and 1ºC as direct effects.

\(^{31}\) GDP is denoted in thousands of constant 2000 dollars per capita. USA emissions are obtained from the World Resources Institute (2009). See Section A3.9 below for the values of the other stocks and flows in the year 2000.
A1.6. The calibration of the law of motion of the stock of physical capital

Physical capital investment is equal to private plus public investment less investment in software. We take $\hat{\delta}^k = 0.06$ as the annual rate of depreciation (Thomas Cooley and Edward Prescott, 1995). In generational terms, $\delta^k = 0.787$.

To approximate the year-to-year discounting, we take $i =$ average investment in physical capital of Generation $t$ per year, and compute that, at the end of Generation $t$’s life, the accumulated investment amounts are

$$i + i \times (1 - \hat{\delta}^k) + i \times (1 - \hat{\delta}^k)^2 + \cdots + i \times (1 - \hat{\delta}^k)^{24} = \frac{1 - (1 - \hat{\delta}^k)^{25}}{1 - (1 - \hat{\delta}^k)} i.$$ 

Thus, since $1 - \hat{\delta}^k = 0.94$, the parameter $k_2 = \frac{1 - (1 - \hat{\delta}^k)^{25}}{1 - (1 - \hat{\delta}^k)} = 13.118$.

The time series of the stock of capital is constructed by the perpetual inventory method, using US data for 1960-2000 and taking 1960 as initial value. For the initial year, 1960

$$S^k_{1960} = \frac{i^k_{1960}}{\hat{\delta}^k + g^k} = \frac{2.51}{0.06 + 0.038} = 25.63 \text{ thousands of constant 2000 dollars per capita},$$

where $i^k$ represents total (private and public) investment per capita minus expenditure in software, and $g^k$ represents the average yearly growth rate of investment between 1960-1970 (set at 0.038). The value for the stock of physical capital in the year 2000 is $\bar{S}^k_{2000} = 73.65$ (in thousands of 2000 dollars per capita).

A1.7. The calibration of the law of motion of the stock of knowledge

The yearly depreciation rate for knowledge commonly used is much lower than the one for capital (e. g., the Bank of Spain uses $\hat{\delta}^n = 0.15$, which would mean that knowledge dissipates almost entirely in one generation). We believe that the discount factor should be higher because of the intergenerational-public-good character of knowledge. A dollar invested in R&D by a firm may well generate no returns to the firm 25 years later, yet its impact to the accumulation of social knowledge capital may be substantial. Thus, as an approximation we take the depreciation rate of the stock of knowledge to be the same as that of physical capital, i. e., in generational terms, $\delta^n = \delta^k = 0.787$.

We approximate the year-to-year discounting with the same argument as in physical capital. If we denote by $i^n$ the average annual expenditure per capita in knowledge, then we could write
\[(1 - \delta^n)S_{i-1}^n + \frac{1 - (1 - \delta^n)^{25}}{1 - (1 - \delta^n)} i^n > S_i^n.\] But, because investment in knowledge is written in efficiency units of labor per capita, then \[\frac{1 - (1 - \delta^n)^{25}}{1 - (1 - \delta^n)} i^n = k_3 x_i^n,\] that is, \[k_3 = \frac{1 - (1 - \delta^n)^{25}}{1 - (1 - \delta^n)} \frac{i^n}{x_i^n},\] where \(i^n/x_i^n\) is the wage of an efficiency unit of labor.

Now, we estimate \[\frac{i^n}{x_i^n} = \frac{\hat{i}^n}{\epsilon^n(1/3)x_i},\] where \((1/3)x_i\) is the total efficient units of labor and \(\epsilon^n\) the share of labor devoted to the production of knowledge. We take \(\epsilon^n = 0.05\) (5% of total labor) and use the average values for the last generation (1976-2000) to obtain \[\frac{i^n}{x_i^n} = \frac{\hat{i}^n_{76-00}}{0.05(1/3)x_{76-2000}} = \frac{0.99}{0.02} = \frac{49.5}{0.02} \approx 49.5\] thousands of 2000 dollars.

Hence, \[k_3 = \frac{1 - (1 - \delta^n)^{25}}{1 - (1 - \delta^n)} \frac{i^n}{x_i^n} = 13.118 \times 49.5 = 649.34.\]

The time series of the stock of knowledge is constructed by the perpetual inventory method, using US data for 1960-2000 and taking 1960 as initial value. For the initial stock of knowledge, \[S_{1960}^n = \frac{i_{1960}^n}{\delta^n + g^n} = \frac{0.421}{0.06 + 0.041} = 4.21\] thousands of constant 2000 dollars per capita, where \(i^n\) represents total expenditure per capita in R&D plus public and private investment in software,\(^{32}\) and \(g^n\) represents the average yearly growth rate between 1960-1970. The value for the stock of knowledge in the year 2000 is \(\bar{S}_{2000}^n = 15.64\) (in thousands of 2000 dollars per capita).

### A1.8. The calibration of the education production function

The parameter \(k_4\), capturing the productivity of education, plays an important role in the model. By definition, \(k_4 = \frac{x_i}{x_{i-1}},\) where both the numerator and the denominator are measured in efficiency units. We can transform efficiency units into hours by the equality \[\frac{x_i}{x_{i-1}} = \frac{(1 + s)^{T} \hat{x}_i}{(1 + s)^{T-1} \hat{x}_{i-1}} = (1 + s) \frac{\hat{x}_i}{\hat{x}_{i-1}}\] (for some \(T\)), where \((1 + s)\) is the growth factor of human capital per generation, and where the “hats” represent annual data in hours. Hence, the calibration of \(k_4\) is based on two rates: \(s\) and the

\(^{32}\) See Footnote 28 above.
share \( \frac{\dot{X}_{t-1}}{\dot{X}_t} \) of time devoted to education out of total time. Note that \( k_4 \) is increasing in \( s \) and decreasing in the share \( \frac{\dot{X}_{t-1}}{\dot{X}_t} \).

We take the average yearly growth rate \( \hat{s} \) of the human capital stock a value \( \hat{s} = 0.67\% \), which yields the per-generation factor \( (1 + s) = (1 + \hat{s})^{25} = 1.0067^{25} \). This figure, supported by the 1960-85 average provided by de la Fuente and Domènech (2001), is lower than the 0.93\% average for 1960-2000 found in Robert Barro and Jong-Wa Lee (2001).

The rate \( \frac{\dot{X}_{t-1}}{\dot{X}_t} \) is the product of the rate of education in labor and the rate of labor in total time. From our time series, we infer that about 10\% of total labor is devoted to education, and that labor accounts for 1/3 of total time. These figures are conservative in the sense that lower values for them would yield a higher value for \( k_4 \). In summary, we take

\[
k_4 = (1.0067)^{25} \frac{1}{0.1} \approx 35.45.
\]

A1.9. Initial values in the benchmark year 2000

The values for the stock of physical capital, \( S_{2000}^k = 73.65 \), and knowledge, \( S_{2000}^n = 15.64 \) (in thousands of 2000 dollars per capita), are obtained by using the perpetual inventory method as reported in sections A1.6-7 above.

The series of human capital stock (in efficiency units) is constructed normalizing year 1950 equal to 1 and taking the average yearly growth rate of human capital stock equal to 0.67\% (de la Fuente and Domènech, 2001). Hence, \( x_t = 1.0067^{t-1950} \) in 1950-efficiency units, and therefore \( \bar{x}_{2000} = 1.0067^{100} = 1.396 \).

We take education to occupy 10\% of labor time. And consequently, \( \bar{x}_{2000} = 1.396 \times 1/3 \times 0.1 = 0.0465 \) in 1950-efficiency units.

Finally, see the calibration of the production functions in Section A3.5 above for total income, consumption and investment.
APPENDIX 2. PROOF OF THEOREM 1: BALANCED GROWTH PATHS IN PROGRAM E

Table A2.1 illustrates the theorem.

<table>
<thead>
<tr>
<th>Initial Cond.</th>
<th>STOCKS</th>
<th>FLOWS</th>
<th>QuoL</th>
</tr>
</thead>
<tbody>
<tr>
<td>$t=1$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$x_0^e$</td>
<td>$S_0^k$ = $q^k x_0^e$</td>
<td>$S_0^n$ = $q^n x_0^e$</td>
<td>$c_1$ = $p^e q^k x_0^e$</td>
</tr>
<tr>
<td>$= (1+g)x_0^e$</td>
<td>$= (1+g)S_0^k$</td>
<td>$= (1+g)S_0^n$</td>
<td></td>
</tr>
<tr>
<td>$t=2$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$x_2^e$</td>
<td>$S_2^k$ = $q^k x_2^e$</td>
<td>$S_2^n$ = $q^n x_2^e$</td>
<td>$c_2$ = $p^e q^k x_2^e$</td>
</tr>
<tr>
<td>$= (1+g)^2x_0^e$</td>
<td>$= (1+g)^2S_0^k$</td>
<td>$= (1+g)^2S_0^n$</td>
<td></td>
</tr>
<tr>
<td>$t$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$x_t^e$</td>
<td>$S_t^k$ = $q^k x_t^e$</td>
<td>$S_t^n$ = $q^n x_t^e$</td>
<td>$c_t$ = $p^e q^k x_{t-1}$</td>
</tr>
<tr>
<td>$= (1+g)^t x_0^e$</td>
<td>$= (1+g)^t S_0^k$</td>
<td>$= (1+g)^t S_0^n$</td>
<td></td>
</tr>
</tbody>
</table>

Table A2.1. Stocks and flows in Theorem 1.

Recall that Program $E[\rho, e^*, S^m*]$ of Section 3 above assumes that emissions and concentrations are fixed at levels $e^*$ and $S^m*$, respectively. It can be written as:

$$\max \Lambda \text{ subject to }\begin{align*}
(\lambda_t) & \quad c_t^e (x_t^e)^{\alpha_t} (S_t^n)^{\alpha_t} (S_t^m - S_t^m*)^{\alpha_t} \geq \Lambda (1+p)^{t-1}, \text{ for } t \geq 1, \\
(y_t) & \quad k_t (S_t^m)^{\alpha_e} (e^*)^\theta_e (x_t^e)^{\theta_e} (S_t^n)^{\alpha_i} (S_t^m)^{\alpha_i} \geq c_t + i_t, \text{ for } t \geq 1, \\
(w_t) & \quad (1-\delta^k)S_{t-1}^k + k_t i_t \geq S_t^k, \text{ for } t \geq 1, \\
(n_t) & \quad (1-\delta^n)S_{t-1}^n + k_t x_t^n \geq S_t^n, \text{ for } t \geq 1, \\
(p_t) & \quad k_t x_{t-1}^n \geq x_t^n + x_t^i + x_t^e, \text{ for } t \geq 1.
\end{align*}$$

The Lagrangian multipliers have been written to the left of the constraints. Our problem is to find the condition on the endowment vector $(x_0^e, S_0^k, S_0^n)$ such that the optimal solution to the
program is a path of steady growth. At steady-state growth there will be three different growth rates:

- the variables $S^e_t, x^e_t, x^c_t, x^l_t$ will grow at a rate $g$;
- the variables $S^k_t, i_t, c_t$ will grow at a rate $\gamma$;
- $\Lambda_t$ will grow at a rate $\rho$.

From the production function, we must have:

$$(1 + \gamma) = (1 + g)^0 (1 + \gamma)^0_0 (1 + g)^0^0.$$  

However, as we have chosen parameters so that $1 - \theta_k = 0 + \theta_n$, we have $\gamma = g$, and so there will be only two growth rates, namely $g$ and $\rho$. From the first constraint, we have $(1 + g)^{x^e_t + x^c_t + x^l_t} = 1 + \rho$; thus a chosen rate $g$ determines $\rho$.

Given the ordered triple $(g, e^*, S^{m*})$, there will be a ray $\Gamma(g, e^*, S^{m*}) \subset \mathbb{R}^3$ such that if the endowment vector $(x_0^e, S_0^k, S_0^c) \in \Gamma(g, e^*, S^{m*})$, then balanced growth at rates $g$ (and $\rho$) will occur at the optimal solution to the program. We proceed to determine this ray.

To do so, we first derive the Kuhn-Tucker conditions for the program, which are:

(a) $(\hat{e} \Lambda) \quad 1 - \sum_{i=1}^{\infty} \lambda_i (1 + \rho)^{i-1} = 0, t \geq 1,$  

(b) $(\hat{e} x^e_t) \quad k_i p_{i+1} - p_t = 0, i.e., p_t = (1/k_i)^{i-1} p_i, t \geq 1,$  

(c) $(\hat{e} x^c_t) \quad \frac{\lambda_i \alpha_i \Lambda(1 + \rho)^{i-1}}{x^c_t} - p_t = 0, t \geq 1,$  

(d) $(\hat{e} x^l_t) \quad k_j n_i = p_t, t \geq 1,$  

(e) $(\hat{e} c_t) \quad \frac{y_t \theta_t (c_t + i_t)}{x^c_t} - p_t = 0, t \geq 1,$  

(f) $(\hat{e} c_t) \quad \frac{\lambda_i \alpha_i \Lambda(1 + \rho)^{i-1}}{c_t} - y_t = 0, t \geq 1,$  

(g) $(\hat{e} i_t) \quad -y_t + k_i w_i = 0, t \geq 1,$  

(h) $(\hat{e} S^k_t) \quad \frac{y_t \theta_k (c_t + i_t)}{S^k_t} - w_t + (1 - \delta^k) w_{i+1} = 0, t \geq 1,$  

(i) $(\hat{e} S^c_t) \quad \frac{\lambda_i \alpha_c \Lambda(1 + \rho)^{i-1}}{S^c_t} + \frac{y_t \theta_c (c_t + i_t)}{S^c_t} + (1 - \delta^c) n_{i+1} - n_t = 0, t \geq 1.$  

We now substitute into these equations the variable values on a balanced growth path.

1. (b) and (c) imply that:
\[ \lambda_t = \left( \frac{p_i x_i^t}{\alpha_t \Lambda} \left( \frac{1 + g}{k_t^i (1 + \rho)} \right) \right)^{t-1}. \]

2. By (a), it follows that
\[ 1 = \left[ \sum_{i=1}^{\infty} \left( \frac{1 + g}{k_t^i} \right)^{-1} \right] \frac{p_i x_i^t}{\alpha_t \Lambda}. \]
This defines \( p_i \) at the solution, and hence \( p_t \).

Note that \( p_i \) will be defined as long as \( k_t^i > 1 + g \), so that the series converges. It follows that:
\[ 1 = \left( \frac{p_i x_i^t}{\alpha_t \Lambda} \right) \frac{k_t^i}{k_t^i - (1 + g)}. \]

3. (d) defines \( n_t = p_t / k_t = \frac{p_t}{k_t} (1 / k_t^i)^{-1} \), whereas (e) defines \( y_t \geq 0 \), and (g) defines \( w_t = y_t / k_t \). Thus all the dual variables are defined and non-negative.

This leaves equations (h), (f) and (i) which we now analyze.

4. Analysis of (h)
(e) implies \( y_t = \frac{p_t x_t^c}{\theta_c (c_t + i_t)} \), so (h) says
\[ \frac{p_t x_t^c}{\theta_c} \frac{\theta_k}{S_t^k} = \frac{y_t - (1 - \delta^k) y_{t+1}}{k_2}. \]
Substituting for \( y_t \), and multiplying by \( \frac{\theta_k}{p_t} \) gives:
\[ \frac{k_t x_t^c}{S_t^k} \frac{\theta_k}{S_t^k} = \frac{x_t^c}{c_t + i_t} - \frac{1 - \delta^k}{k_t^i} \frac{x_{t+1}^c}{c_{t+1}^i + i_{t+1}}, \]
which, using the balanced growth property of the path means:
\[ \frac{k_t x_t^c}{S_t^k} \frac{\theta_k}{S_t^k} = \frac{x_t^c}{c_t + i_t} - \frac{1 - \delta^k}{k_t^i} \frac{x_{t+1}^c}{c_{t+1}^i}. \]

Multiplying by \( \frac{1 + g}{x_t^c} \), we have:
\[ \frac{k_t x_t^c \theta_k}{S_t^k} = \frac{(1 + g) (1 - \delta^k)}{c_t + i_t}. \]

(A) \[ \frac{k_t x_t^c \theta_k}{S_t^k} = \frac{(1 + g) (1 - \delta^k)}{c_t + i_t}. \]

5. Analysis of (f)
(f) implies \( \frac{\lambda_t \alpha_t \Lambda (1 + \rho)^{-1}}{c_t} = \frac{p_t x_t^c}{\theta_c (c_t + i_t)} \), which may be reduced to the equation:
\[ \frac{x_i^j}{\alpha_i} = \frac{c_j x_i^c}{\alpha c \theta_i (c_i + i_i)}. \]

6. Analysis of (i)

We express \( \lambda, y, n \) in terms of \( p_i \); after some algebraic manipulation (i) reduces to:

\[
\frac{\alpha_n x_i^l}{\alpha_i (1 + g) S_0^n} + \frac{\theta_n x_i^c}{\theta_i (1 + g) S_0^n} + \frac{1}{k_3} \left( \frac{1 - \delta^n}{k_4} - 1 \right) = 0.
\]

In sum, we have the three equations (A), (B), and (C). From the primal constraints we have:

\[
k_i (1 + g)^{0_i} (S_0^k)^{0_i} (S_0^n)^{0_i} (x_i^c)^{0_i} (S^{m*})^{0_i} (e^*)^{0_i} = c_i + i_i,
\]

\[
x_0^c (k_4 - (1 + g)) = x_i^c + x_i^l + x_i^l,
\]

\[
k_i i_i = S_0^k (g + \delta^k),
\]

\[
k_i x_i^n = (g + \delta^n) S_0^n.
\]

Define the following expressions:

\[
v^n(g) = \frac{\delta^n + g}{k_3},
\]

\[
p^i(g) = \frac{\delta^k + g}{k_2},
\]

\[
p^c(g) = \left( 1 - \frac{1 - \delta^k}{k_4} \right) \frac{(1 + g) \delta^k}{k_2 \theta_k} - \frac{\delta^k + g}{k_2},
\]

\[
v^i(g) = \left( 1 - \frac{1 - \delta^n}{k_4} \right) \alpha_i \theta_n \frac{p^c(g)}{k_3 \alpha_c \theta_n \left( p^c(g) + p^l(g) \right)} + \alpha_n p^l(g) (1 + g),
\]

\[
v^c(g) = \left( 1 - \frac{1 - \delta^n}{k_4} \right) \alpha_c \theta_c \frac{p^i(g) + p^l(g)}{k_3 \alpha_c \theta_c \left( p^i(g) + p^l(g) \right)} + \alpha_n p^l(g) (1 + g),
\]

\[
q^n(g) = \frac{k_4 - (1 + g)}{v^n(g) + v^i(g) + v^c(g)},
\]

and

\[
q^k \left( g, e^*, S^{m*} \right) = \left( k_1 \left( v^i(g) \right)^{0_i} \right)^{0_i} \left( p^c(g) + p^l(g) \right)^{0_{i+0_c}} (1 + g)^{0_{i+0_c}} \cdot q^n(g) \cdot (e^*)^{0_i+0_c} \cdot \left( S^{m*} \right)^{0_i+0_c}.
\]
Note that these seven functions are all positive. In particular, it is easily checked that
\( p^e(g) > 0 \) for any \( g \geq 0 \), since \( k_4 > 1 \).

Now from (F) we solve for \( i \):
\[
i_i = p^i(g)S_0^k.
\]
From (G), we have
\[
x_i^n = v^n(g)S_0^n.
\]
From (A) and the above expression for \( i_1 \), we have:
\[
c_1 = p^e(g)S_0^k.
\]
Now view (B) and (C) as a pair of simultaneous linear equations in \((x_i^n, x_i^l)\). Solving them gives
\[
(x_i^n, x_i^l) = (v^n(g), v^l(g))S_0^n.
\]
Substituting these values into (E) gives
\[
S_0^n = q^n(g)x_0^n.
\]
Finally, we obtain
\[
S_0^k = q^k(g, e^*, S^{m*})x_0^n
\]
by substituting \( S_0^n = q^n(g)x_0^n \) and \( x^e = v^e(g)q^n(g)x_0^n \) into equation (D) and solving for \( S_0^k \).

Statement (ii) of Theorem 1 is immediately derived from the above equations. Statement (i) asserts that the endowments grow along the ray \( \Gamma(g, e^*, S^{m*}) \) at rate \( 1 + g \), and statement (iii) says that all flow variables exhibit balanced growth.
### APPENDIX 3. REACHING THE RAY $\Gamma(g, e^*, S^{m^*})$ IN TWO GENERATIONS FROM DATE-2000 ENDOWMENTS

Table A3.1 illustrates Step 1 in our computation procedure.

<table>
<thead>
<tr>
<th>Year 2000</th>
<th>STOCKS</th>
<th>FLOWS</th>
<th>utility</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$\bar{x}_k^{e}$</td>
<td>$\bar{S}_k^{e}$</td>
<td>$\bar{S}_n^{e}$</td>
</tr>
<tr>
<td>$t = 1$</td>
<td>$x_1^{e}$</td>
<td>$S_1^k$</td>
<td>$S_1^n$</td>
</tr>
<tr>
<td>$t = 2$</td>
<td>$x_2^{e}$</td>
<td>$S_2^k = q^k x_1^{e}$</td>
<td>$S_2^n = q^n x_2^{e}$</td>
</tr>
<tr>
<td>$t = 3$</td>
<td>$x_3^{e}$</td>
<td>$S_3^k = q^k x_2^{e}$</td>
<td>$S_3^n = q^n x_3^{e}$</td>
</tr>
<tr>
<td>$t = 4$</td>
<td>$x_4^{e}$</td>
<td>$S_4^k = q^k x_4^{e}$</td>
<td>$S_4^n = q^n x_4^{e}$</td>
</tr>
<tr>
<td>$t \geq 4$</td>
<td>$x_t^{e}$</td>
<td>$S_t^k = q^k x_t^{e}$</td>
<td>$S_t^n = q^n x_t^{e}$</td>
</tr>
</tbody>
</table>

**Table A3.1.** Step 1 in our computation procedure, where $q^k = q^k(g, e^*, S^{m^*}), q^n = q^n(g), v^j = v^j(g)(j = l, n, c), p^i = p^i(g)(j = c, i)$.

The ray $\Gamma(g, e^*, S^{m^*})$ is defined by

$$\Gamma(g, e^*, S^{m^*}) = \{(x^e, S^k, S^n) \in \mathbb{R}_+^3 : S^k = q^k(g, e^*, S^{m^*})x^e, S^n = q^n(g)x^e\},$$

where the coefficients $q^k$ and $q^n$ have been computed in Appendix 2 above. Program $G[x^e]$ of Section 3 above can now be written as follows.

**Program $G[x^e]$:** Given $(\rho, e_1, S_1^m, e_2, S_2^n, e^*, S^{m^*})$ and $x^e$, Max $\Lambda_1$ subject to
(A3.1) \((\mu_i) (c_i) \alpha_i (x_i^f)^\alpha (S_i^m)\alpha (\hat{S}^m - \hat{S}_i^m)^\alpha \geq \Lambda_i,\)

(A3.2) \((\mu_2) (c_2) \alpha_i (x_2^f)^\alpha (q^n(g)x_2^e)^\alpha (\hat{S}^m - \hat{S}_2^m)^\alpha \geq (1 + p)\Lambda_1,\)

(A3.3) \((r_1) k_i (x_i^e)^\alpha (S_i^0)^\alpha (e_i)^\alpha (\hat{S}_i^m)^\alpha \geq c_1 + i_1,\)

(A3.4) \((r_2) k_i (x_i^e)^\alpha (q^k(g,e^*,S^{m*})x_i^e)^\alpha (e_i)^\alpha (S_i^m)^\alpha \geq c_2 + i_2,\)

(A3.5) \((z_1) (1 - \delta^k)S_0^k + k_2 \geq S_1^k,\)

(A3.6) \((z_2) (1 - \delta^k)S_0^k + k_2 \geq q^k(g,e^*,S^{m*})x_2^e,\)

(A3.7) \((\beta_1) (1 - \delta^n)S_0^n + k_3 x_1^n \geq S_1^n,\)

(A3.8) \((\beta_2) (1 - \delta^n)S_0^n + k_3 x_2^n \geq q^n(g)x_2^e,\)

(A3.9) \((\zeta_i) k_4 x_i^n \geq x_i^e + x_i^n + x_i^l + x_i^c,\)

(A3.10) \((\zeta_2) k_4 x_i^n \geq x_i^e + x_i^n + x_i^l + x_i^c,\)

For the year-2000 initial conditions \((x_0^e, S^k, S_0^n) = (\bar{x}_0^e, \bar{S}_0^k, \bar{S}_0^n, \bar{S}_0^n)\).

This is a concave program, and therefore the first-order conditions will be sufficient. We have 10 constraints and hence 10 Lagrangian multipliers, shown to the left of each constraint. There are 14 endogenous variables \((\Lambda_1, c_1, x_1^f, x_1^e, x_1^n, c_2, x_2^f, x_2^e, x_2^n, i_1, i_2, S_1^k, S_1^n, \), and hence 14 Kuhn-Tucker conditions, as follows.

\[ KT1: (\partial x_1) (1 - \mu_1 - (1 + p)\mu_2 = 0; \]

\[ KT2: (\partial c_1) \mu_1 \frac{\alpha_i^\alpha_i \Lambda_1}{c_1} - r_1 = 0; \]

\[ KT3: (\partial c_2) \mu_2 \frac{\alpha_i^\alpha_i (1 + p)\Lambda_1}{c_2} - r_2 = 0; \]

\[ KT4: (\partial x_1^f) \mu_1 \frac{\alpha_i^\alpha_i \Lambda_1}{x_1^f} - \zeta_1 = 0; \]

\[ KT5: (\partial x_1^g) \mu_2 \frac{\alpha_i^\alpha_i (1 + p)\Lambda_1}{x_1^g} - \zeta_2 = 0; \]

\[ KT6: (\partial S_1^k) \frac{r_1 \theta_i (c_1 + i_1)}{S_1^k} - z_1 + z_2 (1 - \delta^k) = 0; \]

\[ KT7: (\partial S_1^n) \mu_1 \frac{\alpha_i^\alpha_i \Lambda_1}{S_1^n} + r_1 \frac{\theta_n (c_1 + i_1)}{S_1^n} - \beta_1 + (1 - \delta^n)\beta_2 = 0; \]

\[ KT8: (\partial x_2^e) - \zeta_1 + \zeta_2 k_4 = 0; \]

\[ KT9: (\partial x_2^n) \beta_1 k_4 - \zeta_1 = 0; \]
\[ KT10: (\partial x^i) \quad \beta_2 k_3 - \zeta_2 = 0; \]
\[ KT11: (\partial x^i) \quad \frac{\theta_c (c_1 + i)}{x_1^c} - \zeta_1 = 0; \]
\[ KT12: (\partial x^i) \quad \frac{\theta_c (c_2 + i_2)}{x_2^c} - \zeta_2 = 0; \]
\[ KT13: (\partial x) \quad -r_1 + z_1 k_2 = 0; \]
\[ KT14: (\partial x) \quad -r_2 + z_2 k_2 = 0. \]

(a) From KT11, \( \frac{\zeta_1}{r_1} = 0 \). From KT4 and KT2, \( \frac{\zeta_1}{r_1} = \frac{\mu_i \alpha_i \Lambda_i}{\mu_i \alpha_c \Lambda_i} \frac{1}{c_1} = \frac{\alpha_c c_1}{\alpha_c x_1^c} \). It follows that

\[ c_1 = \frac{\theta_c \alpha_c x_1^c}{\alpha_c x_1^c - \theta_c ^\prime \alpha_c x_1^c} i_1. \]  

Similarly, from KT12, \( \frac{\zeta_2}{r_2} = 0 \). From KT5 and KT3,

\[ \frac{\zeta_2}{r_2} = \frac{\mu_2 \alpha_i (1+\rho) \Lambda_i}{\mu_2 \alpha_c (1+\rho) \Lambda_i} \frac{1}{c_2} = \frac{\alpha_c c_2}{\alpha_c x_2^c}, \text{ yielding} \]

\[ c_2 = \frac{\theta_c \alpha_c x_2^c}{\alpha_c x_2^c - \theta_c x_2^c \alpha_c} i_2. \]  

(b) From KT8

\[ \frac{\zeta_2}{\zeta_1} = \frac{1}{k_4}, \]  

whereas from KT9 and KT10,

\[ \frac{\beta_2}{\beta_1} = \frac{\zeta_2}{\zeta_1}, \]  

yielding

\[ \frac{\beta_2}{\beta_1} = \frac{1}{k_4}. \]  

From KT4 and KT9

\[ \mu_i \frac{\alpha_i \Lambda_i}{x_i^c} = \beta_i k_3, \]  

(b.4)
and from KT5 and KT10

\[
\mu_2 \frac{\alpha_i (1 + \rho) \Lambda_i}{x'_2} = \beta_2 k_3.
\]  \hspace{1cm} (b.5)

Dividing (b.5) by (b.4)

\[
\frac{\mu_2 x'_2}{\mu_1 x'_2} (1 + \rho) = \frac{\beta_2}{\beta_1},
\]  \hspace{1cm} (b.6)

which together with (b.3) yields

\[
k_4 \mu_2 \frac{x'_1}{x'_2} (1 + \rho) = \mu_1.
\]  \hspace{1cm} (b.7)

Substituting (b.7) into KT1 gives

\[
1 = k_4 \mu_2 \frac{x'_1}{x'_2} (1 + \rho) + (1 + \rho) \mu_2,
\]  
i. e.,

\[
\mu_2 (1 + \rho) \left[ k_4 \frac{x'_1}{x'_2} + 1 \right] = 1,
\]  
or:

\[
\mu_2 = \frac{x'_2}{(1 + \rho)(k_4 x'_1 + x'_2)},
\]  \hspace{1cm} (b.8)

which together with (b.7) yields

\[
\mu_1 = \frac{k_4 x'_1}{k_4 x'_1 + x'_2}.
\]  \hspace{1cm} (b.9)

From (b.4) and (b.9),

\[
\beta_1 = \frac{\mu_1}{\beta_3} \frac{\alpha_i \Lambda_i}{x'_1} = \frac{k_4 \alpha_i x'_1 \Lambda_i}{k_3 (k_4 x'_1 + x'_2) x'_1},
\]  
i. e.,

\[
\beta_1 = \frac{k_4 \alpha_i \Lambda_i}{k_3 (k_4 x'_1 + x'_2)},
\]  \hspace{1cm} (b.10)

and from (b.5) and (b.8),

\[
\beta_2 = \frac{\mu_2}{\beta_3} \frac{\alpha_i (1 + \rho) \Lambda_i}{x'_2} = \frac{\alpha_i x'_2 (1 + \rho) \Lambda_i}{k_3 (k_4 x'_1 + x'_2) (1 + \rho) x'_2},
\]  
i. e.,

\[
\beta_2 = \frac{\alpha_i \Lambda_i}{k_3 (k_4 x'_1 + x'_2)}.
\]  \hspace{1cm} (b.11)

From KT9, \( \zeta_1 = \beta_1 k_3 \), i. e., using (b.10),
\[ \zeta_1 = \frac{k_4 \alpha_c \Lambda_1}{k_4 x_1^j + x_2^j}, \]  
(b.12)

and similarly, from KT10 and (b.11).

\[ \zeta_2 = \frac{\alpha_c \Lambda_1}{k_4 x_1^j + x_2^j}. \]  
(b.13)

Finally, from KT2 and (b.9),

\[ r_1 = \frac{k_4 \alpha_c x_1^j \Lambda_1}{(k_4 x_1^j + x_2^j)c_1}, \]  
(b.14)

and from KT3 and (b.8)

\[ r_2 = \frac{x_2^j}{(1+\rho)(k_4 x_1^j + x_2^j)c_2} \alpha_c (1+\rho) \Lambda_1, \]  

i.e.,

\[ r_2 = \frac{x_2^j \alpha_c \Lambda_1}{(k_4 x_1^j + x_2^j)c_2}. \]  
(b.15)

From KT13 and (b.14)

\[ z_1 = \frac{r_1}{k_2} = \frac{k_4 \alpha_c x_1^j \Lambda_1}{k_2 (k_4 x_1^j + x_2^j)c_1}, \]  
(b.16)

and from KT14 and (b.15)

\[ z_2 = \frac{r_2}{k_2} = \frac{x_2^j \alpha_c \Lambda_1}{k_2 (k_4 x_1^j + x_2^j)c_2}. \]  
(b.17)

(c) Inserting (b.14), (b.9), (b.10) and (b.11) into KT7:

\[ \mu_i \frac{\alpha_c \Lambda_1}{S_i^n} + r_1 \frac{\theta_n (c_i + i_j)}{S_i^n} - \beta_1 + (1-\delta^s) \beta_2 = 0, \]

we obtain

\[ \frac{k_4 x_1^j}{k_4 x_1^j + x_2^j} \frac{\alpha_c \Lambda_1}{S_i^n} + \frac{k_4 x_1^j \alpha_c \Lambda_1}{(k_4 x_1^j + x_2^j)c_1} \frac{\theta_n (c_i + i_j)}{S_i^n} - \frac{k_4 \alpha_c \Lambda_1}{k_3 (k_4 x_1^j + x_2^j)} + (1-\delta^s) \frac{\alpha_c \Lambda_1}{k_3 (k_4 x_1^j + x_2^j)} = 0, \]

i.e.,

\[ \frac{k_4 \alpha_c x_1^j}{S_i^n} + \frac{k_4 x_1^j \theta_n (c_i + i_j)}{c_i S_i^n} - \frac{k_4 \alpha_c}{k_3} + (1-\delta^s) \frac{\alpha_c}{k_3} = 0. \]  
(c.1)

Inserting (b.14), (b.16), and (b.17) into KT6:

\[ r_1 \frac{\theta_n (c_i + i_j)}{S_i^k} - z_1 + z_2 (1-\delta^s) = 0, \]
we obtain
\[ \frac{k_i x_i^l \alpha_i \Lambda_i}{(k_i x_i^l + x_i^l)c_i} - \frac{k_i x_i^l \alpha_i \Lambda_i}{k_2(k_i x_i^l + x_i^l)c_1} + \frac{x_i^l \alpha_i \Lambda_i}{k_2(k_i x_i^l + x_i^l)c_2} - \frac{x_i^l(1 - \delta^k)}{k_3} = 0, \]

i.e.,
\[ \frac{k_i x_i^l \theta^k_i(c_i + i_l)}{c_i S_i^k} - \frac{k_i x_i^l}{k_2 c_1} + \frac{x_i^l(1 - \delta^k)}{k_2 c_2} = 0. \]  

(c.2)

(d) In summary, the Kuhn-Tucker conditions yield the following four equations involving only primal variables, which added to the 10 constraints, written as equalities, constitute a system of 14 equations in the 14 primal variables. The four equations are:

\[ c_1 = \frac{\theta^k \alpha_i x_i^l}{\alpha_i x_i^l - \theta^c \alpha_c x_i^l} i_l, \]  

(a.1)

\[ c_2 = \frac{\theta^k \alpha_i x_i^l}{\alpha_i x_i^l - \theta^c \alpha_c x_i^l} i_2, \]  

(a.2)

\[ \frac{k_i x_i^l \theta^k_i(c_i + i_l)}{c_i S_i^k} - \frac{k_i x_i^l}{k_3} + \frac{x_i^l(1 - \delta^k)}{k_3} = 0, \]  

(c.1)

\[ \text{and} \quad \frac{k_i x_i^l \theta^k_i(c_i + i_l)}{c_i S_i^k} - \frac{k_i x_i^l}{k_2 c_1} + \frac{x_i^l(1 - \delta^k)}{k_2 c_2} = 0. \]  

(c.2)

(e) From (A3.5), \[ i_l = \frac{S_i^k - (1 - \delta^k)S_0^k}{k_2}, \]  

(e.1)

which substituted into (a.1) yields
\[ c_1 = \frac{\theta^k \alpha_i x_i^l}{\alpha_i x_i^l - \theta^c \alpha_c x_i^l} \frac{S_i^k - (1 - \delta^k)S_0^k}{k_2}, \]  

(e.2)

and
\[ c_1 + i_l = \left[ \frac{\theta^k \alpha_i x_i^l}{\alpha_i x_i^l - \theta^c \alpha_c x_i^l} + 1 \right] \frac{S_i^k - (1 - \delta^k)S_0^k}{k_2}, \]  

i.e.,
\[ c_1 + i_l = \frac{\alpha_i x_i^c}{\alpha_i x_i^l - \theta^c \alpha_c x_i^l} \frac{S_i^k - (1 - \delta^k)S_0^k}{k_2}, \]  

(e.3)

which in turn gives
\[ \frac{c_1 + i_l}{c_1} = \frac{\alpha_i x_i^c}{\theta^c \alpha_c x_i^l}. \]  

(e.4)

Similarly, from (A3.6), \[ i_2 = \frac{d_i^k x_i^c - (1 - \delta^k)S_0^k}{k_2}, \]  

(e.5)

which substituted into (a.2) yields
\[ c_2 = \frac{\theta_c \alpha_c x_1^l}{\alpha_c x_2 - \theta_c \alpha_c x_1^l} \frac{q^k x_2^e}{k_2} - (1 - \delta^k) S_1^k \]  
\text{(e.6)}

and

\[ c_2 + i_2 = \frac{\alpha_c x_2^e}{k_2} \frac{q^k x_2^e}{\alpha_c x_2 - \theta_c \alpha_c x_1^l} - (1 - \delta^k) S_1^k \]  
\text{(e.7)}

which in turn gives

\[ \frac{c_2 + i_2}{c_2} = \frac{\alpha_c x_2^e}{\theta_c \alpha_c x_1^l}. \]  
\text{(e.8)}

From (A3.7)

\[ x_1^n = \frac{S_1^n - (1 - \delta^n) S_0^n}{k_3}, \]  
\text{(e.9)}

and from (A3.8)

\[ x_2^n = \frac{q^x x_2^c - (1 - \delta^n) S_1^n}{k_3}. \]  
\text{(e.10)}

\[ k_1 (x_1^c)^0 (S_1^k)^0 (S_1^n)^0 (e_1)^0 (S_1^m)^0 - \alpha_i x_1^c \frac{S_1^k - (1 - \delta^n) S_0^k}{k_2 (\alpha_c x_1^c - \theta_c \alpha_c x_2^c)} = 0, \]  
\text{(f.1)}

an equation of the form \( \varphi_1 (x_1^c, x_1^l, S_1^k, S_1^n) = 0, \) while inserting (e.7) into (A3.4) yields

\[ k_1 (x_2^c)^0 (q^x x_2^c)^0 (q^x x_2^c)^0 (e_2)^0 (S_2^m)^0 - \alpha_i x_2^c \frac{q^x x_2^c - (1 - \delta^n) S_1^k}{k_2 (\alpha_c x_2^c - \theta_c \alpha_c x_2^c)} = 0, \]  
\text{(f.2)}

an equation of the form \( \varphi_2 (x_2^c, x_2^l, S_1^k) = 0. \)

Inserting (e.4) into (c.1), we obtain

\[ \frac{k_4 x_1^l \alpha_n}{S_1^n} + \frac{k_4 x_1^l \alpha_\theta}{S_1^n} + \frac{\alpha_i x_1^c}{\theta_c \alpha_c x_1^l} - \frac{k_4 \alpha_i}{k_3} + (1 - \delta^n) \frac{\alpha_i}{k_3} = 0, \]  
\text{or: \[ \theta_c k_4 \alpha_n x_1^l + k_4 \alpha_c \theta_c x_1^l + \alpha_i [1 - \delta^n - k_4] S_1^n = 0, \]  
\text{(f.3)}

a linear equation of the form \( \varphi_3 (x_1^c, x_1^l, S_1^n) = 0. \)

Inserting (e.4), (e.2) and (e.6) into (c.2) yields
\[
\frac{k_4x_i^e}{S_i^k} \theta_c x_i^e - \frac{\alpha}{\theta_c x_i^e} = \frac{k_4x_i^f}{S_i^k} \theta_c x_i^f - \frac{\alpha}{\theta_c x_i^f} + \frac{x_i^e(1-\delta^k)}{k_2} \theta_c x_i^e - \frac{\alpha}{\theta_c x_i^e} + \frac{x_i^f(1-\delta^k)}{k_2} \theta_c x_i^f - \frac{\alpha}{\theta_c x_i^f} = 0,
\]

i. e.,
\[
\frac{k_4x_i^e}{S_i^k} \theta_c x_i^e - \frac{k_4x_i^f}{S_i^k} \theta_c x_i^f + \frac{(1-\delta^k)(\alpha x_i^e - \theta_c x_i^f)}{\theta_c x_i^e - \frac{\alpha}{\theta_c x_i^e} + \frac{(1-\delta^k)\alpha x_i^f - \theta_c x_i^f}{\theta_c x_i^f - \frac{\alpha}{\theta_c x_i^f}} = 0,
\]

an equation of the form \( \varphi_4(x_i^e, x_i^f, x_i^e, x_i^f, S_i^k) = 0 \).

Inserting (e.9) into (A3.9), we obtain
\[
x_i^e + \frac{S_i^e - (1-\delta^k)S_i^0}{k_3} + x_i^f - k_4x_i^e = 0,
\]
a linear equation of the form \( \varphi_5(x_i^e, x_i^f, x_i^e, x_i^f, S_i^e) = 0 \), whereas the insertion of (e.10) into (A3.10) yields
\[
x_i^e + \frac{q^e x_i^e - (1-\delta^k)S_i^e}{k_3} + x_i^f - k_4x_i^e = 0,
\]
a linear equation of the form \( \varphi_6(x_i^e, x_i^f, x_i^e, x_i^f, S_i^e) = 0 \).

Finally, from (A3.1) and (A3.2), we have
\[
(c_e)^a (x_i^e) = (1+p)(c_i)^a (x_i^f)^a (S_i^e - S_i^0)^a.
\]

Inserting (e.6) and (e.2) into this equation yields
\[
\left( \frac{x_i^e}{\alpha x_i^e - \theta_c x_i^f} \right)^a \left( \frac{x_i^f}{\alpha x_i^f - \theta_c x_i^e} \right)^a = (1+p) \left( \frac{x_i^e}{\alpha x_i^e - \theta_c x_i^f} \right)^a \left( \frac{x_i^f}{\alpha x_i^f - \theta_c x_i^e} \right)^a,
\]
an equation of the form \( \varphi_7(x_i^e, x_i^f, x_i^e, x_i^f, S_i^e, S_i^0) = 0 \).

The seven equations (f.1) to (f.7) form a system in the seven unknowns \((x_i^e, x_i^f, x_i^e, x_i^f, S_i^k, S_i^0)\). We numerically solve these seven equations using Mathematica, and then compute all the other values (including \(\Lambda_1\), which can be obtained from (A3.1)). We check that all values and Lagrangian multipliers are non-negative to be assured that we have found a solution.

**APPENDIX 4. UNCERTAINTY**

This appendix refers to results from our companion paper (Llavador et al., 2009) that have a bearing on the analysis in this one. We suppose, following the Stern Review, that there is an exogenous
probability $p$ that mankind becomes extinct at any generation, and that there is an (independent) draw from this random variable at the end of each generation. To model the intergenerational welfare objective, we suppose that there is an Ethical Observer (EO) whose preferences satisfy the expected utility hypothesis. An outcome (or prize) is the event that mankind lasts exactly $T$ generations, with a vector of qualities of life $(\Lambda_1, \ldots, \Lambda_T)$. The EO’s von Neumann-Morgenstern utility at a given outcome is denoted $W^T(\Lambda_1, \ldots, \Lambda_T)$. This function, together with the $p(1 - p)^{t-1}$ exogenous probability of extinction at (the end of) date $t$, define the expected utility of the EO when she chooses an infinite path $(\Lambda_1, \Lambda_2, \ldots)$ as $\sum_{t=1}^{\infty} p(1 - p)^{t-1} W^t(\Lambda_1, \ldots, \Lambda_t)$.

A purely Rawlsian EO would only be concerned with the utility of the worst-off person who ever lived, and hence her vNM utility at the outcome $(\Lambda_1, \ldots, \Lambda_T)$ would be $\min \{\Lambda_1, \Lambda_2, \ldots, \Lambda_T\}$. More generally, we consider the family of vNM utility functions $[1 + (T - 1)\beta] \min \{\Lambda_1, \Lambda_2, \ldots, \Lambda_T\}$ parameterized by $\beta \in [0,1]$, and we call an EO with such a vNM function an extended Rawlsian EO. Note that, for $\beta > 0$, such an EO takes into account both the utility of the worst-off generation and the future time span $T$ of the human species.\(^{33}\)

We prove in Llavador et al. (2009) that, for a simpler economy with education and physical capital as only intergenerational links, if the discounted-utilitarian program with discount factor $\varphi = 1 - p$, diverges, then the solution to the optimization program of the Extended Rawlsian EO under uncertainty is exactly the solution to the Pure Sustainability Optimization program. In particular, $\Lambda_t$ is constant with respect to $t$. The EO can then ignore the uncertainty!

We conjecture that an analogous result holds for the economy with the constraints of Program $E[\rho, e^*, S^m]$ of Section 3 above. Then, as noted in Section 5.1, if we take $1 - p = 0.975$ per generation of 25 years, then the discounted utilitarian program will diverge with our calibration of the parameters. Therefore we conjecture that the solution to the program of the Extended Rawlsian EO is just the solution to the Pure Sustainability Optimization program, for the discount factor $\varphi = 1 - p = 0.975$. The very rough intuition is that the possibilities for growth inherent in a large value of $k_4$ more than counteract the discount on the resources allocated to future generations that the EO might contemplate placing, due the possibility that they may not exist, if $(1 - p)k_4 > 1$.

\(^{33}\) We are indebted to Klaus Nehring for suggesting that we extend the pure Rawlsian EO to the “$\beta = 1$” case.
However, it must be remarked that a more important kind of uncertainty to introduce would be our uncertainty with regard to the physics of global warming. This is a more difficult undertaking.

APPENDIX 5. NORDHAUS’S SOCIAL WELFARE FUNCTION AND THE CALIBRATION OF ITS PARAMETERS

A5.1. A long-lived consumer

The traditional theory of economic growth considers the accumulation of physical capital, in particular the tradeoff between present consumption and the enhanced consumption possibilities of future generations offered by saving. It often postulates a long-lived representative consumer, whose preferences are representable in an additively separable manner as the discounted sum of future single-date subutilities, one for each future date. If only the consumption \( c_t \) at each date enters the single-date subutility function, and if such function is of the form \( \frac{1}{1-\eta} c^{1-\eta} \), then the consumer’s preferences are represented by the utility function

\[
\sum_{t=1}^{T} \frac{1}{1-\eta} \left( \frac{1}{1+\delta} \right) c_t^{1-\eta},
\]

where \( T < \infty, \eta > 0 \) (for \( \eta = 1 \), \( \ln c \) replaces \( \frac{1}{1-\eta} c^{1-\eta} \)), and where the discount factor \( \frac{1}{1+\delta} \) (or the discount rate \( \delta \)) reflects the consumer’s marginal rate of intertemporal substitution: a more impatient consumer has a larger \( \delta \), and attaches little value to a unit of consumption made available to him far into the future.

A5.2. Nordhaus’s social welfare function

The social welfare function in Nordhaus (1991, 1994, 2008a,b), and Nordhaus and Boyer (2000), see (2) in Section 5.2 above, is similar to (A5.1), but with a quite different meaning. Now \( t = 1, 2, \ldots \) represent generations, and \( c_t \) is the consumption per capita of Generation \( t \). As noted in Section 6.1 above, Nordhaus’s (2008a) calls \( \delta \) and \( \eta \) “central” and “unobserved normative parameters,” affecting “the relative importance of the different generations.” The parameter \( \delta \) is a “pure social time discount rate:” a high \( \delta \) means that the welfare of a generation born far into the future counts very little in the social welfare function. The second one represents “the aversion to inequality of
different generations.” Informally speaking, if the rates of growth turn out to be negative, then $\delta$ and $\eta$ push in opposite directions, a high $\delta$ favoring the earlier generations and a high $\eta$ favoring the later, less well off, generations. But for positive rates of growth, when the latter generations are better off, high values of either $\delta$ or $\eta$ favor the earlier generations. This is the case in the paths proposed by Nordhaus (2008a, b).

A5.3. Nordhaus’s calibration of the parameters

Nordhaus (2008a,b) calibrates $\eta$ and $\hat{\delta}$ as follows. First, he adopts the “Ramsey equation”

$$\hat{r} = \hat{\delta} + \eta \hat{g},$$

(A5.2)

where $\hat{r}$ is the real per year rate of interest on capital and $\hat{g}$ is the per year rate of growth of consumption. Nordhaus (2008a, p. 60-61) justifies equation (A5.2) by the maximization of the function (A5.1) subject to some constraints. In his words, and noting that his symbol $\rho$ (resp. $\alpha$) corresponds to the $\delta$ (resp. $\eta$) of the present paper:

“The basic economics can be described briefly. Assume a time discount rate of $\rho$ and a consumption elasticity of $\alpha$. Next, maximize the social welfare function described earlier and in the Appendix with a constant population and a constant rate of growth per generation $g^*$. This yields the standard equation for the equilibrium real return on capital, $r^*$, given by $r^* = \rho + \alpha g^*$.”

Second, he infers $\hat{r}$ and $\hat{g}$ from “observed economic outcomes as reflected by interest rates and rates of return on capital” (p. 33-34).

Third, he chooses $\hat{\delta}$ and $\eta$ subject to the Ramsey equation, which gives one degree of freedom. In particular, Nordhaus (2008a, p. 178) takes the values ($\hat{r}, \hat{g}$) = (0.055, 0.02). Equation (A4.2) then holds for any ($\hat{\delta}, \eta$) pair satisfying $\hat{\delta} = 0.055 - 0.02\eta$, in particular by the values ($\hat{\delta}, \eta$) = (0.015, 2) chosen by Nordhaus (2008a).34

Summarizing, equation (A5.2) is obtained by the constrained maximization of (A5.1), whereas $\hat{r}$ and $\hat{g}$ are deduced from observed behavior. Inserting $\hat{r}$ and $\hat{g}$ into (A5.2) could make sense if, as in Section A5.1 above, observed behavior was generated by a single long-lived consumer who solves the optimization program. But in this case the parameters ($\hat{\delta}, \eta$) would be “positive,”

34 Elsewhere in the book he refers to a $\hat{r}$ of 0.04 (pp. 9-11) and to a $\hat{g}$ of 0.013 (p. 108).
rather than “normative,” whereas Nordhaus’ analysis concerns a world of many distinct generations, with parameters \((\hat{\delta}, \eta)\) which are “normative.” It is peculiar to think of rates of return observed in the market as depending on these “normative” parameters, in particular on the aversion, by past and current market participants, to inequality among generations.

In addition, because Nordhaus (2008a) gives little detail on the constraints of the optimization program leading to (A5.2), it is hard to evaluate the assumption that \(r^*\) and \(g^*\) are constant at the solution. In any event, the solution paths will depend on the initial conditions on the stocks, so that the constancy of rates can typically be justified only asymptotically.\(^{35}\)

\(^{35}\) Consider, for instance, the traditional Ramsey problem, with capital but without environmental stocks: An infinitely lived consumer maximizes 

\[
\int_0^\infty \frac{1}{1-\eta} c(t)^{1-\eta} e^{-\delta t} \, dt
\]

subject to the law of motion of capital \(k_R\) and the initial condition

\(k_R(0) = k_0\). Let capital depreciate at the rate \(\delta_R\), and let the production function be \(A k_R \psi e^{nt}\), where \(\psi \in (0,1)\) and \(n \geq 0\) is the rate of exogenous technological change. The constraint is then

\[
\dot{k}_R(t) \leq A k_R(t)^{1-\eta} + \lambda [Ak_R \psi e^{nt} - c(t) - \delta_R k_R(t)]
\]

Writing the Hamiltonian as

\[
H(c, k_R, \lambda) = (1-\eta)^{-1} c^{1-\eta} e^{-\delta t} + \lambda [Ak_R \psi e^{nt} - c - \delta_R k_R],
\]

at the solution path one must have (see, e. g., George Hadley and Murray Kemp, 1971, Th. 4.3.1) (a) \(\frac{\partial H}{\partial c} = 0\), i.e., \(c^{-\eta} e^{-\delta t} - \lambda = 0\), and (b) \(-\frac{\partial H}{\partial k_R} = \dot{\lambda}\), i.e., \(-\lambda (A \psi k_R^{1-\eta} e^{nt} - \delta_R) = \dot{\lambda}\). From (a), \(-\eta c^{-\eta-1} \cdot \dot{c} \cdot e^{-\delta t} + c^{-\eta} \cdot e^{-\delta t} \cdot (\delta) = \dot{\lambda}\), which together with (b), using (a) again and dividing through by \(c^{-\eta} \cdot e^{-\delta t}\), gives \(A \psi k_R^{1-\eta} e^{nt} - \delta_R = \delta + \eta \frac{\dot{c}}{c}\), a time-dependent form of (A5.2). Assume now that \(\frac{\dot{c}}{c} = \bar{g}\), a constant, i.e., \(c(t) = c_0 e^{\bar{g}t}\) for some \(c_0 > 0\). The last equation then reads

\[
A \psi k_R(t)^{1-\eta} e^{nt} = \delta_R + \dot{\bar{g}} + \eta \bar{g},
\]

i.e., \(k_R(t) = (\delta_R + \delta + \eta \bar{g})^{1/(1-\psi)} (A \psi)^{1/(1-\psi)} e^{nt/(1-\psi)}\), and the initial condition \(k_R(0) = k_0\) requires \(k_0 = (\delta_R + \delta + \eta \bar{g})^{1/(1-\psi)} (A \psi)^{1/(1-\psi)}\). Writing \(k_R(t) = k_0 e^{nt/(1-\psi)}\) and dividing through by \(k_R\), the law of motion becomes

\[
\frac{n}{1-\psi} = A k_0^{1-\psi} - \frac{c}{k} \delta_R, \text{ i.e., } A[k_0 e^{nt/(1-\psi)}]^{1-\psi} e^{nt/(1-\psi)} - \frac{c_0}{k_0} e^{\bar{g}t} e^{nt/(1-\psi)} - \delta_R, \text{ or }
\]

\[
\frac{n}{1-\psi} = A k_0^{1-\psi} - \frac{c_0}{k_0} e^{\bar{g}t} e^{nt/(1-\psi)} - \delta_R, \forall t, \text{ which implies that } \bar{g} = \frac{n}{1-\psi}. \text{ But then the parameters }
\]

\((\eta, \delta, A, \psi, n, \delta_R, k_0)\) must belong to the set of measure zero defined by the equality

\[
k_0 = (\delta_R + \delta + \eta \frac{n}{1-\psi}) \frac{1}{1-\psi} (A \psi)^{1/(1-\psi)}.
\]
REFERENCES


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