The End of Economic Growth? Unintended Consequences of a Declining Population

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Abstract

In many models, economic growth is driven by people discovering new ideas. These models typically assume either a constant or a growing population. However, in high income countries today, fertility is already below its replacement level: women are having fewer than two children on average. It is a distinct possibility — highlighted in the recent book, *Empty Planet* — that global population will decline rather than stabilize in the long run. What happens to economic growth when population growth turns negative?

*I am grateful to Sebastian Di Tella for many helpful discussions about this project.*
1. Introduction

In many growth models based on the discovery of new ideas, the size of the population plays a crucial role. Other things equal, a larger population means more researchers which in turn leads to more new ideas and to higher living standards. This basic feature is shared by the original endogenous growth models of Romer (1990), Aghion and Howitt (1992), and Grossman and Helpman (1991) as well as by the semi-endogenous growth models of Jones (1995), Kortum (1997), and Segerstrom (1998). It is a feature of numerous other models.¹

In a recent book entitled *Empty Planet*, Bricker and Ibitson (2019) make the case based on a rich body of demographic research that global population growth in the future may not only fall to zero but may actually be negative. For example, the natural rate of population growth (i.e. births minus deaths, ignoring immigration) is already negative in Japan and in many European countries such as Germany, Italy, and Spain (United Nations, 2019).

Figure 1 shows historical data on the total fertility rate for various regions. This measure is the average number of live births a cohort of women would have over their reproductive life if they were subject to the fertility rates of a given five-year period. To sustain a constant population requires a total fertility rate slightly greater than 2 in order to compensate for mortality. The graph shows that high income countries as a whole, as well as the U.S. and China individually, have been substantially below 2 in recent years. According to the U.N.’s *World Population Prospects 2019*, the total fertility rate in the most recent data is 1.8 for the United States, 1.7 for China and for High Income Countries on average, 1.6 for Germany, 1.4 for Japan, and 1.3 for Italy and Spain. In other words, fertility rates in the rich countries of the world are already consistent with negative long-run population growth: women are having fewer than two children throughout much of the developed world.

A sharp downward trend in India and for the world as a whole is also evident in the figure. As countries get richer, fertility rates appear to decline to levels consistent, not with a constant population, but actually with a declining population.

Conventional wisdom holds that in the future, global population will stabilize at

Figure 1: The Total Fertility Rate (Live Births per Woman)

Note: The total fertility rate is the average number of live births a hypothetical cohort of women would have over their reproductive life if they were subject during their whole lives to the fertility rates of a given period and if they were not subject to mortality. Each data point corresponds to a five-year period. Source: United Nations (2019).

something like 8 or 10 billion people. But maybe that is not correct after all. We surely do not know for certain what will happen in the future. The fact that so many rich countries already have fertility below replacement indicates that a future with negative population growth is a possibility that deserves further consideration.

The models of economic growth cited above assume a constant or growing population, and for understanding economic growth historically, that is clearly the relevant case. The demographic evidence, however, suggests that this may not be the case in the future. Hence the focus of this paper: what happens to economic growth if population growth is negative?

We show below — first in models with exogenous population growth and then later in a model with endogenous fertility — that negative population growth can be particularly harmful. When population growth is negative, both endogenous and semi-endogenous growth models produce what we call an Empty Planet result: knowledge and living standards stagnate for a population that gradually vanishes. In a model with
endogenous fertility, even the social planner can get stuck in this trap if society delays implementing the optimal allocation and suffers from inefficient negative population growth for a sufficiently long period. In contrast, if the economy switches to the optimal allocation soon enough, it can converge to a balanced growth path with sustained exponential growth: an ever-increasing population benefits from ever-rising living standards. Policies related to fertility may therefore determine whether we converge to an “empty planet” or to an “expanding cosmos”; they may be much more important than we have appreciated.

1.1 Literature Review

Many models feature endogenous fertility, modeled in a variety of ways. Becker and Barro (1988) and Barro and Becker (1989) take an altruistic approach in which the utility of children enters the utility function for parents, giving rise to a dynastic utility function. Papers that follow this approach include Doepke and Zilibotti (2005) and Manuelli and Seshadri (2009). Other papers emphasize a “warm-glow” effect in which parents care about the number of their offspring; for example, see De La Croix and Doepke (2003) and Doepke and Tertilt (2016). Finally, many papers feature a quantity-quality tradeoff and assign a key role to education, often in the context of explaining the demographic transition and the emergence of modern economic growth. These include Becker, Murphy and Tamura (1990), Galor and Weil (1996, 2000), Greenwood and Seshadri (2002), and Kalemli-Ozcan (2002). On the empirical side, Jones and Tertilt (2008) provide a detailed account of the decline in U.S. fertility using Census data, while Delventhal, Fernandez-Villaverde and Guner (2019) study the demographic transition using data from 186 countries and 250 years. Doepke and Tertilt (2016) and Greenwood, Guner and Vandenbroucke (2017) provide general surveys of family macroeconomics, including fertility. In general, this literature sometimes recognizes the possibility that population growth could ultimately be negative, but that is not its emphasis.

The literature that explicitly considers negative population growth in a growth context is much smaller. Manuelli and Seshadri (2009) explain the heterogeneity in international fertility rates by emphasizing that taxes and transfers in Europe may in part be responsible for low fertility. Sasaki and Hoshida (2017) study negative population growth in a semi-endogenous growth setting. They show that the rate of technological
change falls to zero as people endogenously exit the research sector. More surprisingly, they provide a setting where negative population growth leads to positive steady-state growth in income per person because capital per person rises as the number of people declines. However, this result is incomplete in that they assume a zero depreciation rate for capital: if there is a fixed amount of capital but the population declines, then capital per person grows. One can easily generalize their result to positive depreciation rates using a Solow model. If the rate of population decline is \( \eta \) and capital depreciates at rate \( \delta \), then there are two possible regimes. If \( \eta > \delta \), i.e. the rate of population decline is faster than the depreciation rate of capital, then \( K/L \) rises asymptotically along a balanced growth path. But when \( \eta < \delta \), instead, you get the standard Solow result of constant \( K/L \) in steady state. Empirically, rates of population decline are perhaps 1% or smaller, whereas depreciation rates are 3% or 5% or more. The Sasaki and Hoshida (2017) case of exponential growth in capital per person from declining population therefore seems implausible as an empirical matter. Christiaans (2011) has results along these lines in a model with increasing returns that results from externalities to capital, showing the two possible regimes. This general result suggests that capital can be omitted from the model without much loss in generality, which is what we do below.

This motivates Sasaki (2019) to consider a model with non-renewable resources, where a zero depreciation rate is more natural. In that case, though, one might wonder about elasticities of substitution: if a single Robinson Crusoe populated an earth full of land and natural resources, would her income be extremely high? This would be interesting to consider in the future, but we omit natural resources as well in what follows.

2. The Empty Planet Result

How do idea-based growth models behave when population declines? We begin by introducing exogenous, negative population growth into a simplified version of the Romer (1990), Aghion and Howitt (1992), and Grossman and Helpman (1991) endogenous growth models. This case turns out to be especially easy to analyze. Then we consider semi-endogenous growth models.
2.1 Fully Endogenous Growth as in Romer/AH/GH

Consider the following simplified version of idea-driven endogenous growth models:

\[ Y_t = A_t^\sigma N_t \]  
\[ \frac{\dot{A}_t}{A_t} = \alpha N_t \]  
\[ N_t = N_0 e^{-\eta t}, \quad \eta > 0 \]

According to equation (1), a single consumption-output good is produced using people \( N_t \) and the stock of ideas (“knowledge”) \( A_t \). Crucially, as in Romer (1990), there is constant returns in this production function to rival inputs — here just people — and therefore increasing returns to people and ideas together. The degree of increasing returns is parameterized by \( \sigma \).

Equation (2) is the endogenous growth equation. It says that the growth rate of knowledge is proportional to the population. The literature often distinguishes between researchers and workers who produce the consumption good, but not always. Here, we make the simplifying assumption that is closer in spirit to learning by doing: people can work to make consumption goods and get new ideas at the same time. An exogenous split of people into workers and researchers would deliver identical results to those below. One can also allow this allocation to be determined endogenously.

Finally, equation (3) specifies that the population declines exogenously at the rate \( \eta \). For example, \( \eta = .005 \) corresponds to a population that declines exponentially at a half a percent per year. We write the model here and throughout the paper so that all parameter values (Greek letters) are positive.

Combining (2) and (3) gives the following differential equation, in which the growth rate of knowledge declines exponentially:

\[ \frac{\dot{A}_t}{A_t} = \alpha N_0 e^{-\eta t}. \]

This differential equation is easy to solve, yielding the following result (detailed derivations for this and other results are available in Appendix A):

**Result 1** (Romer/AH/GH with Negative Population Growth): In the Romer/AH/GH model
with negative population growth, the stock of knowledge $A_t$ is given by

$$\log A_t = \log A_0 + \frac{gA_0}{\eta} (1 - e^{-\eta t})$$

Both $A_t$ and income per person $y_t \equiv Y_t/N_t$ converge to constant values $A^*$ and $y^*$ as $t$ goes to infinity, where

$$A^* = A_0 \exp \left( \frac{gA_0}{\eta} \right)$$

$$y^* = y_0 e^{g_0 \eta} = \exp \left( \frac{g_0}{\eta} \right)$$

where variables indexed by 0 denote initial values.

We refer to this as the **Empty Planet result**. Economic growth stagnates as the stock of knowledge and living standards settle down to constant values. Meanwhile, the population itself falls at a constant rate, gradually emptying the planet of people. This outcome stands in stark contrast to the conventional result in growth models in which knowledge, living standards, and even population grow exponentially: not only do we get richer over time, but these higher living standards apply to an ever rising number of people.

The last equation in Result 1 is amenable to calibration. For example, if $g_0 = 2\%$ and $\eta = 1\%$, so that the population is falling at 1% per year, the long-run level of GDP per person will be $e^2 \approx 7.4$ times higher than current income. Obviously, slower declines in population would make this factor even higher.

In what follows, we explore the robustness of this finding. First, we see that it occurs in semi-endogenous growth models as well, and then we consider what happens when the population growth rate itself is an endogenous outcome.

### 2.2 Semi-Endogenous Growth with Declining Population

With positive rates of population growth, semi-endogenous growth models in the tradition of Jones (1995), Kortum (1997), and Segerstrom (1998) give very different results from the fully endogenous growth models. We see next that with negative population growth, the **Empty Planet** result still emerges.

A simplified semi-endogenous growth model is obtained by changing the idea pro-
production function:

\[
Y_t = A_t^{\sigma} N_t \quad (4)
\]

\[
\frac{\dot{A}_t}{A_t} = \alpha N_t^\lambda A_t^{-\beta} \quad (5)
\]

\[
N_t = N_0 e^{-\eta t}. \quad \eta > 0 \quad (6)
\]

In particular, we introduce the parameter \( \beta > 0 \), capturing the extent to which new ideas (proportional improvements in productivity) are getting harder to find.\(^2\)

Combining (5) and (6) gives the following differential equation:

\[
\frac{\dot{A}_t}{A_t} = \alpha N_0^\lambda e^{-\lambda \eta t} A_t^{-\beta}. 
\]

Solving this differential equation gives the next result.

**Result 2** (Semi-Endogenous Growth with Negative Population Growth): In the semi-endogenous growth model with negative population growth, the stock of knowledge \( A_t \) is given by

\[
A_t = A_0 \left( 1 + \frac{\beta g A_0}{\lambda \eta} \left( 1 - e^{-\lambda \eta t} \right) \right)^{1/\beta}. 
\]

Defining \( \gamma = \lambda \sigma / \beta \) to capture the overall degree of increasing returns to scale in this economy, both \( A_t \) and income per person \( y_t = Y_t/N_t \) converge to constant values \( A^* \) and \( y^* \) as \( t \) goes to infinity, where

\[
A^* = A_0 \left( 1 + \frac{\beta g A_0}{\lambda \eta} \right)^{1/\beta}. 
\]

\[
y^* = y_0 \left( 1 + \frac{g y_0}{\gamma \eta} \right)^{\gamma/\lambda}. \quad (7)
\]

Along the transition path, the growth rate satisfies

\[
\frac{\dot{y}_t}{y_t} = g y_0 \cdot \left( \frac{y_t}{y_0} \right)^{-\lambda} e^{-\lambda \eta t} = \frac{g y_0 e^{-\lambda \eta t}}{1 + \frac{2g y_0}{\gamma \eta} \left( 1 - e^{-\lambda \eta t} \right)}
\]

In other words, the growth rate falls to zero slightly faster than \( e^{-\lambda \eta t} \).

\(^2\)An alternative in the literature is to write the idea production function as \( \dot{A}_t = \alpha N_t^\lambda A_t^\phi \) with \( \phi < 1 \). These are equivalent, with \( \beta = 1 - \phi \).
This result confirms that both endogenous growth and semi-endogenous growth lead to the *Empty Planet* outcome. Rather than sustained exponential growth in living standards and population, living standards stabilize for a vanishing number of people.

Quantitatively, however, the level at which lower living standards stagnate can be much lower with semi-endogenous growth. To illustrate, we need to calibrate one additional parameter relative to what we had before. Across a range of different case studies, Bloom, Jones, Van Reenen and Webb (2019) find estimates of $\beta \approx 3$ when $\sigma = 1$ (a normalization in the general case where we do not observe ideas directly) and $\lambda = 1$ (a standard value in the literature), which gives $\gamma = 1/3$. Plugging these values into equation (7), the long-run level of GDP per person would be $(1 + 3 \cdot 2)^{1/3} \approx 1.9$ times higher than current income. The difference versus the endogenous growth case is striking: with $\beta = 0$ (so that $\gamma = \infty$), long run income was 7.4 times higher than current income for the same parameter values.

### 3. Endogenous Fertility and the Equilibrium Allocation

We now endogenize the population growth rate itself. There are many related ways to accomplish this, but the literature has not converged on a single approach; see the literature review at the start of this paper for references. Almost all approaches assume that having offspring is a time intensive activity, and this is at the center of the approach we take below.

In models of endogenous fertility, population growth in a decentralized equilibrium can be optimal or may be either above or below the optimal rate. In fact, because the number of people is endogenous, the definition of “optimal” is itself not obvious; for example, see Golosov, Jones and Tertilt (2007). The most natural case of interest here is one in which parents do not fully internalize the fact that their offspring create nonrival ideas that benefit the entire economy, so that equilibrium fertility is too low.

But there are also other possible nuances. For example, Farhi and Werning (2007) note that the social planner may care about future generations both because individuals care about their own children and because the social planner puts weight on each generation. This means that social welfare will generally put more weight on future generations than individuals do, also leading optimal fertility to be higher than equilib-
Table 1: Economic Environment: Endogenous Fertility Model

<table>
<thead>
<tr>
<th>Category</th>
<th>Equation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Final output</td>
<td>$Y_t = A_t^\sigma (1 - \ell_t) N_t$</td>
</tr>
<tr>
<td>Population growth</td>
<td>$\frac{\dot{N}_t}{N_t} = n_t \equiv \tilde{n}_t - \delta$</td>
</tr>
<tr>
<td>Fertility</td>
<td>$\tilde{n}_t = b(\ell_t) \equiv \bar{a} + \bar{b} \log \ell_t$</td>
</tr>
<tr>
<td>Ideas</td>
<td>$\frac{\dot{A}_t}{A_t} = N_t^{\lambda} A_t^{-\beta}$</td>
</tr>
<tr>
<td>Generation 0 utility</td>
<td>$U_0 = \int_0^\infty e^{-\rho t} u(c_t, \tilde{N}_t) dt, \quad \tilde{N}_t \equiv N_t/N_0, \rho \equiv \bar{\rho} + \delta$</td>
</tr>
<tr>
<td>Flow utility</td>
<td>$u(c_t, \tilde{N}_t) = \log c_t + \epsilon \log \tilde{N}_t$</td>
</tr>
<tr>
<td>Consumption</td>
<td>$c_t = Y_t/N_t$</td>
</tr>
</tbody>
</table>

rium fertility. Externalities to human capital in models with a quality-quantity tradeoff can also give rise to inefficiently low fertility. Alternatively, one can construct idea-based models in which optimal fertility is below equilibrium fertility; see Jones (2003) and Futagami and Hori (2010) for some discussion. Here, we do not attempt to draw any firm conclusion about the range of possible externalities that may exist. Instead, we focus on one interesting case in which the equilibrium features negative population growth while the optimal allocation has positive population growth.

3.1 Environment

The economic environment for the setup with endogenous fertility is in Table 1. It builds on our earlier model, with one enhancement. There is now a single allocative decision that has to be made at each date: each person is endowed with one unit of time that can be used to produce either consumption or offspring. Devoting $\ell_t$ units of time to producing children leads to a fertility rate of $\tilde{n}_t = b(\ell_t) = \bar{a} + \bar{b} \log \ell_t$ where $b(\cdot)$ can be thought of as the “birth” function. The log function is convenient, as when $\ell_t$ is sufficiently small, the net replacement rate can be negative, which we think of as a couple choosing to have fewer than two children. There is a constant death rate, $\delta$, and the population growth rate is $n_t \equiv \tilde{n}_t - \delta$.

This setup excludes many other considerations such as human capital, physical
capital, and a quantity-quality tradeoff that would be interesting to explore in the future. We instead focus on the simplest model that allows us to make our points.

People obtain utility from consumption and from having descendents. The expected lifetime utility of a member of the generation born at date 0 is

$$U_0 = \int_0^\infty e^{-(\bar{\rho} + \delta)t} u(c_t, \tilde{N}_t) dt$$

where $\bar{\rho}$ is the pure rate of time preference, $c_t$ is consumption, and $\tilde{N}_t \equiv N_t/N_0$ is the number of descendents of generation 0 at date $t$. Discounting also occurs because of the death rate, and we define $\rho \equiv \bar{\rho} + \delta$ as the overall discount rate. The flow of utility at date $t$ is

$$u(c_t, \tilde{N}_t) = \log c_t + \epsilon \log \tilde{N}_t.$$

The log functional forms throughout are convenient in several ways. First, $\log c_t$ leads income and substitution effects to cancel, so that a constant fertility rate is consistent with balanced growth. The $\log \tilde{N}_t$ is helpful for time consistency and for simplifying the value function in solving for transition dynamics later.

### 3.2 A Competitive Equilibrium with Externalities

As in Romer (1990), the nonrivalry of ideas leads to increasing returns. Some departure from pure perfect competition is necessary, and the equilibrium in general will not be efficient. We consider a simple equilibrium here in which the production of ideas is purely external. Also, we start with the equilibrium allocation because it is designed to be simple. Section 4 below considers an optimal allocation.

Firms produce final output in perfectly competitive markets, taking the stock of ideas $A_t$ as exogenous. Each person chooses time spent raising children versus working in the market sector, $\ell_t$ versus $1 - \ell_t$, in order to maximize utility, also taking the time path of $A_t$ as exogenous. Hence ideas evolve according to the idea production function entirely as an externality: people do not recognize that by having children, their kids may produce new knowledge in the future that makes the entire economy more productive. Markets are perfectly competitive, subject to the idea externality, and the only price is the wage per unit of work, given by $w_t = A_t^\sigma$ in equilibrium.
Taking \( \{ w_t \} \) as given, people in each generation solve

\[
\max_{\{ \ell_t \}} \int_0^\infty e^{-\rho t} u(c_t, \tilde{N}_t) dt
\]

subject to

\[
\dot{\tilde{N}}_t = (b(\ell_t) - \delta) N_t
\]

\[
c_t = w_t (1 - \ell_t)
\]

and given the function forms assumed in Table 1.

The Hamiltonian for this problem is

\[
\mathcal{H} = u(c_t, \tilde{N}_t) + v_t (b(\ell_t) - \delta) N_t
\]

where \( v_t \) is the shadow price (in utils) of another person.

The first-order condition for this problem with respect to \( \ell_t \) is

\[
\frac{v_t N_t b' (\ell_t)}{u_c(c_t, \tilde{N}_t) w_t} = \frac{1}{1 - \ell_t},
\]

(8)

The first-order condition for \( N_t \) is

\[
\rho = \frac{\dot{v}_t}{v_t} + \frac{1}{v_t} \left( \frac{\epsilon}{\tilde{N}_t} + v_t n_t \right).
\]

(9)

To solve further, define \( V_t = v_t N_t \) to be the shadow value in utils of the entire population. Then (9) can be written as

\[
\dot{V}_t = \rho V_t - \epsilon.
\]

This differential equation obviously has a steady state at \( V_{eq}^* = \epsilon / \rho \). Simple inspection of the differential equation reveals that this steady state is unstable: if \( V_t \) differs from \( V_{eq}^* \), then \( V_t \) goes off to \( \pm \infty \). So the only path that does not violate the transversality condition is

\[
V_t = V_{eq}^* = \frac{\epsilon}{\rho}.
\]

That is, \( V_t = V_{eq}^* \) at all points in time.
Substituting this into (8) and using our function form assumptions gives the equilibrium amount of time spent on fertility. This is stated in our next result:

**Result 3 (The Equilibrium with Endogenous Fertility):** In the equilibrium with endogenous fertility, the allocation of time devoted to offspring is constant and given by

\[
\ell_t = \frac{bV_t}{1+bV_t} = \frac{bV_{eq}^*}{1+bV_{eq}^*} = \frac{1}{1 + \rho b\epsilon} \equiv \ell_{eq}
\]  

(10)

Equilibrium population growth is then also constant at the rate \( n_{eq} = \bar{a} - \bar{b}\ell_{eq} - \delta \). Depending on parameter values, equilibrium population growth can be positive or negative.

Fertility is constant at each point in time, even along the transition path. A higher rate of time preference (\( \rho \)) lowers fertility, while a higher preference for offspring (\( \epsilon \)) or a better fertility technology (\( \bar{b} \)) raises fertility. This setup, therefore, does not explain the demographic transition or how we got to the situation we are in today; that’s not its purpose.

Instead, it allows us to study the current situation: depending on parameter values, equilibrium population growth can be positive or negative. The negative case is the one of interest here. In that case, we have an equilibrium setup with endogenous fertility that feeds naturally into the results from Section 2. The negative population growth combined with the idea production function implies that the equilibrium with endogenous fertility features a growth rate that falls to zero so that output per person converges to a steady state, as in equation (7). Therefore, the *Empty Planet* result can be supported as an equilibrium outcome with endogenous fertility.

### 4. The Optimal Allocation

Now instead consider the optimal allocation in this economic environment. With endogenous fertility, there is no unique criterion for social welfare, as discussed by Golosov, Jones and Tertilt (2007). Instead, we consider the allocation that maximizes the utility of a representative generation. The key reason this differs from the equilibrium allocation considered above is that the optimal allocation takes into account the fact that a larger population generates more nonrival ideas, raising everyone’s income. This will lead
optimal fertility to be higher than its equilibrium rate.

The Hamiltonian for the optimal allocation is

\[ H = u(c_t, \tilde{N}_t) + \mu_t \tilde{N}_t^\lambda A_t^{1-\beta} + v_t (b(\ell_t) - \delta) N_t \]

where \( \mu_t \) is the shadow price (in utils) of an idea and \( v_t \) is the shadow price (in utils) of another person. The first-order condition for this problem with respect to \( \ell_t \) is

\[ \frac{v_t N_t b'(\ell_t)}{1 - \ell_t} = \frac{u_c(c_t, \tilde{N}_t) y_t}{1 - \ell_t} = \frac{1}{1 - \ell_t}. \]  

(11)

Notice that this equation has the same form as the equilibrium first-order condition; however, the shadow value of people, \( v_t \), will be different. (We abuse notation for now by not using a different letter for the equilibrium versus optimal \( v_t \).)

The first-order condition with respect to \( A_t \) can be expressed as an arbitrage equation:

\[ \rho = \frac{\dot{\mu}_t}{\mu_t} + \frac{1}{\mu_t} \left( u_c \sigma y_t A_t + \mu_t (1 - \beta) \frac{\dot{A}_t}{A_t} \right). \]  

(12)

The required rate of return from investments is \( \rho \), and the investments in ideas yield both a capital gain and a dividend. Continuing this analogy, this equation can be solved to yield the shadow price of an idea along a balanced growth path as the initial dividend divided by “r-g”:

\[ \mu_t^* = \frac{1}{\rho - g_\mu - (1 - \beta) g_A} = \frac{\sigma / A_t^*}{\rho + \beta g_A} \]  

(13)

where the second equality using the fact that the growth rate of the dividend (and hence the growth rate of \( \mu \)) equals \(-g_A\) along a BGP.

Similarly, the first-order condition for \( N_t \) in arbitrage form is

\[ \rho = \frac{\dot{v}_t}{v_t} + \frac{1}{v_t} \left( \frac{\epsilon}{N_t} + \mu_t \lambda \frac{\dot{A}_t}{N_t} + v_t n_t \right). \]  

(14)
4.1 Steady State Balanced Growth

Using (13), the social value of people along a balanced growth path is given by

\[ V_{sp}^* = v_t^* N_t^* = \frac{\epsilon + \lambda z^*}{\rho} \]  

where

\[ z^* \equiv \mu_t^* \dot{A}_t^* = \frac{\sigma g_A^*}{\rho + \beta g_A^*} \]  

(16)

denotes the social value of the new ideas produced at a point in time, and we’ve used the fact that \( g_v = -n \) along a BGP.

Finally, if the planner solution features positive population growth in the steady state, i.e. if \( n_{sp}^* > 0 \), then

\[ g_A^* = \frac{\lambda n_{sp}^*}{\beta} \]  

(17)

\[ g_y^* = \gamma n_{sp}^* \], where \( \gamma \equiv \frac{\lambda \sigma}{\beta} \).  

(18)

As usual in semi-endogenous growth models, the long-run growth rate is the product of the overall degree of increasing returns to scale, \( \gamma \equiv \lambda \sigma / \beta \) and the rate at which scale is growing, \( n_{sp}^* \). Alternatively, if the planner solution features zero or negative population growth in the steady state, then \( g_A^* = g_y^* = 0 \).

Equations (15) and (16) make clear the difference between the equilibrium and the optimal allocations in steady state. The equilibrium value of people is \( V_{eq}^* = \frac{\xi}{\rho} \), while the social value includes an additional term reflecting the ideas that are produced:

\[ V_{sp}^* = \frac{\epsilon + \lambda z^*}{\rho} \Rightarrow V(n) = \begin{cases} \frac{1}{\rho} \left( \epsilon + \frac{\gamma}{1 + \frac{\lambda n}{\sigma \rho}} \right) & \text{if } n > 0 \\ \frac{\xi}{\rho} & \text{if } n \leq 0 \end{cases} \]  

(19)

Both allocations value an extra person because of the additional utility that descendents provide, via \( \epsilon \). Only the optimal allocation values the extra ideas produced at the margin by the extra person. However, if \( n^* \leq 0 \), then the growth rate \( g_A^* \) is also zero, implying from (16) that \( z^* = 0 \). That is, the value of the additional ideas produced by people falls to zero (we discuss this more below), so that additional people are only valued because of the utility benefit, just as in the equilibrium.
Figure 2: A Unique Steady State for the Optimal Allocation when $n_{eq}^* > 0$

Note: When equilibrium fertility is positive, there is a unique solution to equations (19) and (20).
That is, the optimal allocation features a unique steady state with $n_{sp}^* > n_{eq}^*$. Solving (11) gives the optimal allocation of time along a BGP as well as the implied population growth rate:

$$\ell_{sp}^* = \frac{bV_{sp}^*}{1 + bV_{sp}^*} \Rightarrow n(V) = \bar{a} - \delta + b \log \left( \frac{bV}{1 + bV} \right)$$ (20)

Because the planner values the ideas produced by more people, the optimal allocation generally features higher fertility.

Equations (19) and (20) give two nonlinear equations in two unknowns, $n$ and $V$ that characterize the optimal allocation in steady state. These two equations cannot be solved analytically but instead are characterized graphically in Figures 2 and 3.

Figure 2 considers the case in which equilibrium fertility is positive. In this case, there is a unique solution to equations (19) and (20). The optimal allocation features a unique steady state in which optimal population growth exceeds the equilibrium rate, i.e. $n_{sp}^* > n_{eq}^*$. The case of interest in this paper, however, is when equilibrium population growth
Figure 3: Multiple Steady States in the Optimal Allocation when $n_{eq}^* < 0$

![Graph showing multiple steady states](image)

Note: When equilibrium fertility is negative, there are generically three candidate solutions to equations (19) and (20) that characterize the optimal allocation in the steady state. We will see later that the middle steady state is unstable and can be ruled out.

is negative, and that case gives rise to a rich set of outcomes, as suggested by Figure 3. With $n_{eq}^* < 0$, the optimal allocation generically features three candidate steady states. The high steady state has positive population growth. The low steady state has the same population growth rate as the equilibrium allocation and features negative population growth; this is the Empty Planet outcome. Finally, there is a middle steady state in between. We will see shortly that this steady state is unstable and would never be reached along the optimal path. But even the presence of two stable steady states suggests that the dynamics of the optimal allocation are rich.

### 4.2 Stability and Transition Dynamics

The transition dynamics turn out to be economically interesting but must be solved for numerically. We begin by describing the baseline parameter values that we use in the analysis; the results are robust to a range of alternative values.
**Parameter values.** Table 2 summarizes our parameter choices for the numerical examples. They are chosen to be somewhat realistic, but the main point is to show qualitatively what the transition dynamics of the model look like.

Because we do not observe ideas directly, it is convenient to normalize \( \sigma = 1 \) so that \( \lambda \) has the units of total factor productivity. The extensive evidence on idea production functions in Bloom, Jones, Van Reenen and Webb (2019) suggests that \( \beta/\lambda \approx 3 \) is a typical value. With little lost, we assume \( \lambda = 1 \) so that \( \beta = 3 \); \( \lambda = 3/4 \) gives similar results.

We assume a conventional rate of time preference of \( \rho = 1\% \). We assume a maximum population growth rate of \( \bar{a} = 10\% \) could hypothetically be achieved if 100% of time were devoted to fertility and the death rate were zero. We set \( \delta = 1\% \), implying an expected lifetime of 100 years.

Motivated by the recent fertility experience in the OECD, Japan, and the United States, we assume \( n_{eqm} = -0.5\% \), so that in equilibrium, the population will decline at half a percent per year. Finally, we assume that the typical person spends about 1/6 of his or her time producing and raising children.

Given these assumptions, the following four equations determine the values of \( \bar{b}, \epsilon, n^{sp}, \) and \( \ell^{sp} \):

\[
\begin{align*}
    n^{eq} &= \bar{a} + \bar{b} \log \ell^{eqm} - \delta = -0.5\% \quad (21) \\
    \ell^{eq} &= \frac{1}{1 + \frac{\rho}{b \epsilon}} = 1/6 \quad (22) \\
    n^{sp} &= \bar{a} + \bar{b} \log \ell^{sp} - \delta \quad (23) \\
    \ell^{sp} &= \frac{1}{1 + \frac{\rho}{b (\epsilon + \lambda z^*)}} \quad (24)
\end{align*}
\]

where \( z^* \) is given by (16) and (17).

**Implied Parameter Values and Steady-State Results.** The implied parameter values and steady-state outcomes (corresponding to the “high” steady state of the planner problem) are then shown in the bottom part of Table 2. It is worth noting that the optimal population growth rate given these values is substantially higher than the equilibrium rate: 1.86% versus -0.5%. Even with sharp dynamic diminishing returns in the idea production function (\( \beta = 3 \)), there is a large positive externality to offspring in this
Table 2: Parameter Values and Steady-State Results for the Numerical Examples

Key Assumed Values as Inputs to Numerical Examples

<table>
<thead>
<tr>
<th>Parameter/Moment</th>
<th>Value</th>
<th>Comment</th>
</tr>
</thead>
<tbody>
<tr>
<td>σ</td>
<td>1</td>
<td>Normalization</td>
</tr>
<tr>
<td>λ</td>
<td>1</td>
<td>Duplication effect of ideas; 0.75 possible</td>
</tr>
<tr>
<td>β</td>
<td>3</td>
<td>Bloom, Jones, Van Reenen and Webb (2019)</td>
</tr>
<tr>
<td>ρ</td>
<td>.01</td>
<td>Standard value</td>
</tr>
<tr>
<td>̄a</td>
<td>10%</td>
<td>Maximum population growth rate if ℓ = 1 and δ = 0</td>
</tr>
<tr>
<td>δ</td>
<td>1%</td>
<td>Death rate</td>
</tr>
<tr>
<td>neq</td>
<td>-0.5%</td>
<td>Suggested by fertility rates in Europe, Japan, U.S.</td>
</tr>
<tr>
<td>ℓeq</td>
<td>1/6</td>
<td>Time spent raising children</td>
</tr>
</tbody>
</table>

Implied Parameter Values and Steady-State Results

<table>
<thead>
<tr>
<th>Result</th>
<th>Value</th>
<th>Comment</th>
</tr>
</thead>
<tbody>
<tr>
<td>̄b</td>
<td>.025</td>
<td>̄n^{eq} = ̄a + ̄b \log ℓ^{eq} − δ = −0.5%</td>
</tr>
<tr>
<td>ε</td>
<td>.080</td>
<td>From equation (22) for ℓ^{eq}</td>
</tr>
<tr>
<td>n^{sp}</td>
<td>1.86%</td>
<td>From equations (23) and (24) for ℓ^{sp} and n^{sp}</td>
</tr>
<tr>
<td>ℓ^{sp}</td>
<td>0.43</td>
<td>From equations (23) and (24) for ℓ^{sp} and n^{sp}</td>
</tr>
<tr>
<td>g_y^{sp} = g_A^{sp}</td>
<td>0.62%</td>
<td>Equals γn^{sp} with σ = 1</td>
</tr>
<tr>
<td>...</td>
<td>73.1%</td>
<td>Idea share of social value of people: λz^<em>/(ε + λz^</em>)</td>
</tr>
</tbody>
</table>

Note: The first panel in the table shows key assumptions that are an input into the numerical examples. The second panel shows implied parameter values and steady-state results given these assumptions. The steady-state for the planner problem corresponds to the “high” steady state.
calibration. The associated steady-state growth rate is 0.62%, which can be compared to 1% or 2% in the U.S. economy in recent decades.

Equation (19) suggests an alternative way to make the point about the size of the population externality. The equilibrium social value of people is \( V_{eq}^* = \epsilon/\rho \) while the optimal value in steady state is \((\epsilon + \lambda z^*)/\rho\). The last row of Table 2 reports the fraction of the optimal social value of people that is associated with ideas: \( \lambda z^*/(\epsilon + \lambda z^*) \approx 73.1\% \). Of course, this is not a formal calibration that one should take too seriously, but the values give a sense of the magnitudes that may be relevant.

**Stability and Dynamics.** Given the presence of three candidate steady states in the optimal allocation, the transition dynamics are subtle. We derive the results in detail in Appendix B. Here, we summarize the key findings.

While there are two fundamental state variables in our environment \((N_t, A_t)\), it is convenient for studying transition dynamics to consider the state variables as \(N_t\) and \(x_t = A_t^{\beta}/N_t^\lambda\). We will refer to \(x_t\) as “knowledge per person,” which is a slight abuse of language in that it ignores the exponents, but this is convenient. As shown in Appendix B, the optimal allocation of labor can be expressed solely as a function of \(x_t\), which means that population growth can as well. This allows us to study the transition dynamics in a simple two-dimensional plane, as in Figure 4.

The high steady state is stable. There is a wide range of potential starting points for knowledge per person, \(x\), such that the economy ultimately converges to the high steady state. This is what one would generally expect in a problem like this.

The middle steady state is unstable: linearizing the first-order conditions from the Hamiltonian of the optimal problem reveals that it is unstable with imaginary eigenvalues, generating spiral dynamics. In other words, for a given value of knowledge per person, \(x\), close to the middle steady state, there are multiple candidate paths that satisfy the first-order conditions from the Hamiltonian. For the parameter values in Table 2, the path that delivers the highest welfare is shown in Figure 4: it involves staying on the “upper” arm of the spiral for values of \(x\) slightly higher than \(x_{middle}^*\) and then a jump down to the “lower” outer arm of the spiral for even higher values of \(x\); see Figure 5, discussed next, for more details. This jump does not violate any first-order condition because the dynamics are unstable around the jump: the economy never
Figure 4: Transition Dynamics for the Optimal Allocation

Note: This figure shows the transition dynamics in the optimal allocation. The state variable on the horizontal axis is $x \equiv \frac{A^{\beta}}{N^{\lambda}}$, which we somewhat loosely refer to as “knowledge per person.” Arrows indicate transition dynamics. If the economy begins with a stock of knowledge per person that is not too high, it converges to the stable “high” steady state. Alternatively, if knowledge per person is sufficiently high, the economy converges to the low steady state with negative population growth, which equals the equilibrium rate. The candidate “middle steady state” is unstable with imaginary eigenvalues that generate a spiral path locally. The dynamics shown correspond to the path that maximizes welfare for each value of the state variable and correspond to the upper and lower arms of the spiral. See the appendix for details.
jumps along an optimal path.

Figure 5 zooms in on the middle candidate steady state and gives the details underlying the jump. The top panel shows the first spiral around the middle steady state. The blue line is the upper arm and the green line is the lower arm. Solid lines are optimal paths, while dashed lines are points that are dominated. The numbers in red report welfare for \( N = 1 \). The jump between the upper arm and the lower arm occurs where the welfares are equal. The bottom panel zooms in closer to the middle steady state to show the “second” spiral: the blue line is a continuation of the “upper” arm while the green line is a continuation of the “lower” arm, but their positions are reversed since we are now on an inner spiral (all of the points on this second spiral are dominated by points on the outer arms).

If the starting value for \( x \) is below the jump point in Figure 4, the economy converges to the high steady state. Conversely, if the starting value for \( x \) is above the jump point, then the economy asymptotically converges to the “low” steady state, the Empty Planet outcome. This features negative population growth at a rate that asymptotically equals the equilibrium rate \( n^{eq} \) and occurs as knowledge per person, \( x = A^{\beta}/N^\lambda \), goes off to infinity. This happens because \( N_t \) falls due to the negative population growth while the stock of knowledge \( A_t \) stabilizes at some positive value, as in Result 2.

### 4.3 The Economics of the Transition Dynamics

The transition dynamics lead to an important economic point. Consider an economy that is governed by the equilibrium allocation. It features negative population growth at rate \( n^{eq} \), and suppose the economy is initially endowed with a certain population and stock of knowledge such that knowledge per person, \( x_0 \), equals 50. Now look back at Figure 4. The social planner would like the economy to have a much higher fertility rate and converge to the high steady state, which would feature positive population growth and positive economic growth: both the number of people and the amount of income per person would rise exponentially forever. In contrast, the equilibrium allocation will simply move the economy steadily to the right (to higher values of \( x \)) along the lower dashed line: there will be a constant negative rate of population growth, so knowledge
Figure 5: Population Growth Near the Middle Steady State

(a) The first spiral

(b) Zooming in: the second spiral

Note: The top panel shows the first spiral around the middle steady state. The blue line is the upper arm and the green line is the lower arm. Solid lines are optimal paths, while dashed lines are points that are dominated. The numbers in red report welfare for $N = 1$. The jump between the upper arm and the lower arm is the point where the welfares are equal. The bottom panel zooms in closer to the middle steady state to show the “second” spiral: the blue line is a continuation of the “upper” arm while the green line is a continuation of the “lower” arm, but their positions are reversed since we are now on an inner spiral.
per person, \( x \), will rise as the number of people declines.\(^3\)

At any point in time, society may adopt better institutions, such as a fertility subsidy, that move the economy to the optimal allocation. If this occurs at \( x = 100 \) or \( x = 400 \) or \( x = 1600 \), then the economy will eventually transition to the high steady state and exhibit exponential growth forever. However — and this is the surprising point — if the economy delays adopting good institutions for too long, eventually knowledge per person \( x \) will rise above the jump point (around 4200 in the figure). Once this happens, the optimal regime changes. Adopting good institutions that deliver the optimal allocation will involve negative population growth, albeit at a higher rate than the equilibrium. This means that knowledge per person will continue to grow — not because of large increases in knowledge but rather because of declines in the number of people — and the economy will converge to the low steady state. Population will decline, knowledge will remain below an upper bound, and income per person will stagnate. This is the Empty Planet outcome. The surprise is that if society waits too long to adopt good institutions, the optimal allocation switches from one of sustained exponential growth in population, knowledge, and living standards to one of stagnation and an empty planet.

This is summarized in our last main result:

**Result 4 (The Optimal Allocation with Endogenous Fertility):** The allocation that maximizes the welfare of each generation converges to one of two steady states. If the economy adopts the optimal allocation while knowledge per person, \( x \), is sufficiently low, it leads to sustained exponential growth in population, knowledge, and living standards. Alternatively, if the economy waits too long to switch to the optimal path, it converges to the Empty Planet outcome: living standards stagnate as the population gradually declines toward zero.

The intuition for the two steady states goes as follows. An increase in knowledge per person \( x \) causes optimal fertility to decline because the extra ideas produced by offspring have a diminishing marginal benefit; this explains the negative slope of \( n(x) \) in Figure 4. If equilibrium fertility is positive, then optimal fertility will also remain positive — the planner values people at least as much as the equilibrium. But if equilibrium

\(^3\)As shown in Appendix B, the law of motion for \( x \) is \( \dot{x}_t = \beta - \lambda n(x_t)x_t \). When population growth is negative, \( \dot{x} \) is therefore positive.
fertility is negative, then for $x$ high enough, optimal fertility becomes negative as well. This is because as $x$ goes to infinity, the stock of knowledge divided by the number of people is so high that the “knowledge value” of additional offspring falls to zero. But once population growth is negative, $x$ increases over time rather than decreasing since the denominator of $x \equiv A^\beta / N^\lambda$ is falling. That causes $x$ to increase, reinforcing the change. That is the intuition for the bifurcation point in Figure 4.

5. Conclusion

Growth models based on the discovery of new ideas rely on either a constant or a growing population. But current fertility rates in high-income countries are already below replacement. Global population projections by the United Nations suggest that negative population growth for the world as a whole may set in after the year 2100; Bricker and Ibbitson (2019) suggests this may happen as soon as 2050. These two facts motivate the present paper: what happens to economic growth if future population growth rates are negative instead of zero or positive?

Both fully endogenous and semi-endogenous growth models lead to qualitatively similar results: with negative population growth, knowledge and living standards stagnate while the population shrinks toward zero, an outcome we called the *Empty Planet* result.

Endogenizing fertility leads to an additional possibility. When the equilibrium fertility rate is negative, the optimal allocation features two stable steady states. If the economy adopts the optimal allocation soon enough, it converges to a “good” outcome with exponential growth in population, knowledge, and living standards. But if the economy waits too long to switch, even the optimal allocation converges to the “bad” *Empty Planet* outcome.

This is not a forecast, of course, and there are many ways in which this model could be misleading. Automation could enhance our ability to produce ideas sufficiently that growth in living standards continues even with a declining population, for example. Or perhaps automation will create new technologies for raising offspring. Nevertheless, the recent evidence on fertility and the fact that it may have such profound implications for the future of economic growth make this a topic worthy of further exploration.
References


Appendix to “The End of Economic Growth?
Unintended Consequences of a Declining Population”

A. Derivation of Results

Derivation of Result 1. Romer/AH/GH with Negative Population Growth

Integrate the differential equation:

\[
\int \frac{dA_t}{A_t} = \alpha N_0 \int e^{-\eta t} dt
\]

which gives

\[
\log A_t = C_0 - \frac{\alpha N_0}{\eta} e^{-\eta t}
\]

Setting \(t = 0\) to solve for the constant gives

\[
C_0 = \log A_0 + \frac{\alpha N_0}{\eta}
\]

Next, note that \(g_{A0} = \alpha N_0\). Then the time path for the stock of ideas over time:

\[
\log A_t = \log A_0 + \frac{g_{A0}}{\eta} (1 - e^{-\eta t})
\]

So that as \(t \to \infty\),

\[
\log A_t \to \log A^* \equiv \log A_0 + \frac{g_{A0}}{\eta}
\]

In other words, an exponentially declining growth rate leads to a steady state level of technology and income per person.

\[
y_t \to y^* \equiv \left(A_0 e^{g_{A0}/\eta}\right)^\sigma
\]

Finally, converting fully into output terms using \(g_y = \sigma g_A\):

\[
\frac{y^*}{y_0} = e^{g_{yo}/\eta} = \exp \left(\frac{g_{yo}}{\eta}\right).
\]
Derivation of Result 2. Semi-Endogenous Growth with Negative Population Growth

Integrate the differential equation:

\[ \int A_t^{\beta-1} dA_t = \alpha N_0^\lambda \int e^{-\lambda \eta t} dt \]

which gives

\[ \frac{1}{\beta} A_t^{\beta} = C_0 - \frac{\alpha N_0^\lambda}{\lambda \eta} e^{-\lambda \eta t} \]

Setting \( t = 0 \) to solve for the constant gives

\[ C_0 = \frac{1}{\beta} A_0^{\beta} + \frac{\alpha N_0^\lambda}{\lambda \eta} \]

Then the time path for the stock of ideas over time:

\[ A_t^{\beta} = A_0^{\beta} + \frac{\beta \alpha N_0^\lambda}{\lambda \eta} \left( 1 - e^{-\lambda \eta t} \right) \]

Dividing by \( A_0^{\beta} \) and noting that \( g_{A0} = \alpha N_0^\lambda A_0^{-\beta} \) gives

\[ \frac{A_t}{A_0} = \left( 1 + \frac{\beta g_{A0}}{\lambda \eta} \left( 1 - e^{-\lambda \eta t} \right) \right)^{1/\beta} \]

Converting to output using \( y = A^\sigma \) and defining \( \gamma \equiv \lambda \sigma / \beta \) to measure the overall degree of increasing returns to scale:

\[ \frac{y_t}{y_0} = \left( 1 + \frac{g_{y0}}{\gamma \eta} \left( 1 - e^{-\lambda \eta t} \right) \right)^{\gamma / \lambda} \] (25)

Taking the limit as \( t \to \infty \),

\[ \frac{y^*}{y_0} = \left( 1 + \frac{g_{y0}}{\gamma \eta} \right)^{\gamma / \lambda} \]

Taking logs and derivatives of equation (25) gives the growth rate over time:

\[ \frac{\dot{y}_t}{y_t} = g_{y0} \cdot \left( \frac{y_t}{y_0} \right)^{-\frac{\gamma}{\lambda}} e^{-\lambda \eta t} = \frac{g_{y0} e^{-\lambda \eta t}}{1 + \frac{g_{y0}}{\gamma \eta} \left( 1 - e^{-\lambda \eta t} \right)} \]
Derivation of Result 3. The Equilibrium with Endogenous Fertility

These results are all derived in the main text.

Derivation of Result 4. The Optimal Allocation with Endogenous Fertility

These results are derived in the main text and in Appendix B on transition dynamics.

B. Solving Numerically for Transition Dynamics

The numerical solution of the transition dynamics for the optimal allocation proceeds in four broad steps, first listed here and then described in more detail below:

1. We linearize the transition dynamics from the FONC associated with Hamiltonian to see that the high steady state is stable while the middle steady state is unstable with spiral dynamics.

2. We state the optimal allocation problem in terms of a value function and a Hamilton-Jacobi-Bellman (HJB) equation, which can be specialized so that the policy function depends on a single state variable ($x_t$).

3. We solve the HJB equation numerically; see the Matlab program HJB.m.

4. We solve the original Hamiltonian system numerically; see the Matlab program HamiltonianDynamics.m.

Solving the transition dynamics in these two ways helps in understanding the dynamics (and allows for an independent check that the solution is correct).

B.1 Linearizing around the Candidate Steady States

Using the idea production function $\dot{A}_t/A_t = N_t^{\lambda}A_t^{1-\beta}$ as well equations (12) and (14), the FOCN conditions from the Maximum Principle for the optimal allocation can be expressed in terms of three key variables, a state-like variable $x_t \equiv A_t^{\beta}/N_t^{\lambda}$ and the two
costate-like jump variables, $V_t \equiv v_t N_t$ and $z_t \equiv \mu_t \dot{A}_t$:

$$
\dot{x}_t = \beta - \lambda f(V_t)x_t
$$

$$
\dot{V}_t = \rho V_t - (\epsilon + \lambda z_t)
$$

$$
\dot{z}_t = (\rho + \lambda f(V_t))z_t - \frac{\sigma}{x_t}
$$

where $n_t = f(V_t) \equiv b(\ell(V_t)) - \delta$.

In steady state, each of these key variables is constant. Linearizing this differential system around the steady state gives

$$
\dot{x}_t = -\lambda n^*(x_t - x^*) - \lambda x^* f'(V^*)(V_t - V^*)
$$

$$
\dot{V}_t = \rho(V_t - V^*) - \lambda(z_t - z^*)
$$

$$
\dot{z}_t = \frac{\sigma}{(x^*)^2}(x_t - x^*) + \lambda z^* f'(V^*)(V_t - V^*) + (\rho + \lambda n^*)(z_t - z^*)
$$

where $f'(V^*) = \frac{\bar{b}}{V^*(1 + bV^*)}$. Expressing the linearized system in matrix form with $X \equiv [x \ V \ z]'$ allows us to write it as $\dot{X}_t = B(X_t - X^*)$ where $B$ is the matrix of coefficients, which in turn depends on various steady-state values.

We can now evaluate this linearized system around the candidate steady states using the parameter values in Table 2. First, consider the “high” steady state. The matrix $B_{high}$ has one negative eigenvalue and two positive eigenvalues, all real, indicating a saddle-path stable steady state. In contrast, the “middle” steady state is unstable: all its eigenvalues have positive real parts, and two are imaginary. These results are computed in the Matlab program HamiltonianDynamics.m.

This general characterization is broadly robust to the parameter values and shows our first point: the high steady state is stable, while the middle steady state has unstable spiral dynamics.

### B.2 The Value Function Approach

We begin by setting up the value function using the “natural” state variables, $A$ and $N$. The Hamilton-Jacobi-Bellman equation for the optimal allocation is

$$
\rho V(A, N) = \max_{\ell} \sigma \log A + \log(1 - \ell) + \epsilon \log N + V_{A}N^\lambda A^{1-\beta} + V_{N}Nn(\ell)
$$

(26)
Taking the FOC with respect to \( \ell \) gives

\[ \ell = \frac{bV_N N}{1 + bV_N N}. \quad (27) \]

This expression for the planner takes the same form as the equilibrium solution in equation (10), where \( V_N N \) replaces \( V_t \): time invested in children depends on the value of people in both cases.

Taking the derivative of the value function with respect to \( N \) (and using the envelope theorem), we obtain

\[ V_N N = \frac{\epsilon + \lambda V_A \dot{A}}{\rho + (1 + \eta(A, N))n(\ell)}, \quad (28) \]

where \( \eta(A, N) \) is the elasticity of \( V_N \) with respect to \( N \): \( \eta(A, N) \equiv \frac{V_{NN} N}{V_N N} \).

This expression says that the social planner values people for two reasons. First is the direct effect from the utility captured by \( \epsilon \); this is the only effect that the equilibrium allocation considers. Second, is the value of the additional ideas produced by the population, captured by the \( \lambda V_A \dot{A} \) term. This is the effect ignored by individuals in the equilibrium allocation. The denominator of (28) converts these effects into a present discounted value.

**The Value Function Approach, Revisited.** We now show how to redefine the state variables to simplify the problem. Define new state variables to be \( p_t \equiv \log N_t \) and \( x_t \equiv \frac{A_t \beta}{N_t} \). We will refer to \( x_t \) as “knowledge per person”; because of the exponents, this is not strictly accurate, but it is convenient and helps with intuition. Notice that in terms of these states,

\[ \frac{\dot{A}_t}{A_t} = \frac{1}{x_t} \]

and

\[ c_t = A_t' (1 - \ell_t) = x_t^\gamma N_t^\gamma (1 - \ell_t), \quad \gamma \equiv \lambda \sigma / \beta \]

and therefore flow utility is

\[ u = \log c + \epsilon \log N \]

\[ = \frac{\gamma}{\lambda} \log x + (\epsilon + \gamma) \log N + \log(1 - \ell) \]

\[ = \frac{\gamma}{\lambda} \log x + (\epsilon + \gamma) p + \log(1 - \ell). \]
Also, the law of motion for $x$ is
\[ \dot{x}_t = \beta - \lambda n(\ell)x_t. \] (29)

With this setup, the Hamilton-Jacobi-Bellman equation for the optimal allocation is
\[ \rho \tilde{W}(x, p) = \max_{\ell} \left( \epsilon + \gamma \right) p + \frac{\gamma}{\lambda} \log x + \log(1 - \ell) + \tilde{W}_p n(\ell) + \tilde{W}_x[\beta - \lambda n(\ell)x] \] (30)

A key simplification is obtained by guessing a partial form of this value function. In particular, we guess that
\[ \tilde{W}(x, p) = \alpha p + W(x) \] (31)
for some constant $\alpha$. With this guess $\tilde{W}_p = \alpha$, $\tilde{W}_{pp} = 0$, and $\tilde{W}_{xp} = 0$. Computing these derivatives of the value function in equation (32) directly, using the envelope theorem, gives
\[ \rho \tilde{W}_p = (\epsilon + \gamma) + \tilde{W}_{pp} n(\ell) + \tilde{W}_{xp}[\beta - \lambda n(\ell)x] \]

Applying our guess for the form of the value function means that
\[ \alpha = \frac{1}{\rho} (\epsilon + \gamma), \]
and verifies that our guess is correct: the value function can be written as a separable function, linear in the log of population.

Because $\tilde{W}_x = W_x$, we can rewrite the key part of the HJB equation as
\[ \rho W(x) = \max_{\ell} \frac{\gamma}{\lambda} \log x + \log(1 - \ell) + \alpha n(\ell) + W_x[\beta - \lambda n(\ell)x]. \] (32)

This reduces our two-state problem to a one-state problem that is much easier to analyze both graphically and numerically. Notice that under the initial condition $N_0 = 1$ (a free normalization given that we can choose the units of population), $W(x) = \tilde{W}(x, 0)$ and the value functions are equal; otherwise, welfare just shifts up based on the size of the population.

Taking the FOC with respect to $\ell$ gives
\[ \ell = \frac{bV}{1 + bV} \] (33)
where $\hat{V} \equiv \alpha - \lambda W_x$. Comparing this to (27) earlier reveals that $\hat{V}$ is the “social” value of people.

**B.3 Numerical Transition Dynamics using the HJB Value Function**

We solve for the value function numerically using the finite difference approach discussed by Moll (2018). Given that we have reduced the problem to a single state variable, the method discussed in those notes applies directly.

The only subtlety is the presence of the unstable middle candidate steady state. However, the numerical solution using the FONC from the Hamiltonian, discussed next in Section B.4 delivers values for the outer arms of the spirals, providing good guesses for the place where the transition path become vertical so that $\dot{x} = 0$. These points can be used as endpoints for the grid for $x$ in the numerical solution.

Figure 6 shows the phase diagram in $(x, \dot{x})$ space. From equation (29), as $x \to \infty$, $\dot{x}/x \to -\lambda n_{eq} > 0$. The phase diagram makes the nature of the transition dynamics clear and justifies the dynamics shown in the main text in Figure 4.

Figure 7 shows the numerical solution of the value function, which is nicely concave and appears continuous despite the jump in the phase diagram. This is because the jump point occurs precisely the point where welfare is the same on either side of the jump. This will be clear in the next subsection.

**B.4 Numerical Transition Dynamics using the Original Hamiltonian**

Earlier in the appendix in Section B.1, we stated the differential system from the FONC for the Hamiltonian approach:

\[
\begin{align*}
\dot{x}_t &= \beta - \lambda f(V_t)x_t \\
\dot{V}_t &= \rho V_t - (\epsilon + \lambda z_t) \\
\dot{z}_t &= (\rho + \lambda f(V_t))z_t - \frac{\sigma}{x_t}
\end{align*}
\]

where $n_t = f(V_t) \equiv b(\ell(V_t)) - \delta$.

We solve this system numerically using a “reverse shooting” approach. That is, we start from a candidate steady state, move a tiny amount away in the direction of the eigenvector of the linearized system, and solve the dynamics backwards to characterize
Figure 6: The Phase Diagram

Note: Numerical solution of the HJB equation using the Matlab program HJB.m.
Figure 7: The Value Function

Note: From the numerical solution of the HJB equation using the Matlab program HJB.m.
the optimal path. This is especially useful in tracing out the spiral dynamics around the unstable middle steady state. For example, if we begin with $x_0 > x_{\text{high}}$, the reverse dynamics move to the right and eventually spiral into the middle steady state (the “upper arm” of the spiral). Alternatively, if we begin with $x_0 = 10,000$, i.e. close to the “low” steady state, and solve backwards, this takes us along the “lower arm” of the spiral back to the unstable middle steady state. Figure 5 in the main text shows the numerical solution of these dynamics.

The advantage of the Hamiltonian approach numerically is that it traces out the spirals “automatically,” so to speak. The HJB value function approach seeks instead to find the optimal allocation. In this application, both play important roles in understanding the transition dynamics.