



The World's Largest Open Access Agricultural & Applied Economics Digital Library

This document is discoverable and free to researchers across the globe due to the work of AgEcon Search.

Help ensure our sustainability.

Give to AgEcon Search

AgEcon Search

<http://ageconsearch.umn.edu>

aesearch@umn.edu

*Papers downloaded from **AgEcon Search** may be used for non-commercial purposes and personal study only. No other use, including posting to another Internet site, is permitted without permission from the copyright owner (not AgEcon Search), or as allowed under the provisions of Fair Use, U.S. Copyright Act, Title 17 U.S.C.*

No endorsement of AgEcon Search or its fundraising activities by the author(s) of the following work or their employer(s) is intended or implied.

Theory of Stochastic Optimal Economic Growth

by

Lars J. Olson

Dept. of Agricultural and Resource Economics

University of Maryland

College Park, MD 20742-5535 USA

and

Santanu Roy

Department of Economics

Southern Methodist University

Dallas, TX 75275 USA

WP 05-01

Department of Agricultural and Resource Economics

The University of Maryland, College Park

Theory of Stochastic Optimal Economic Growth

Lars J. Olson and Santanu Roy

Olson is Professor of Agricultural and Resource Economics, University of Maryland, College Park, MD 20742 USA (email: lolson@arec.umd.edu)

Roy is Professor of Economics, Southern Methodist University, Dallas, TX 75275 USA (email: sroy@mail.smu.edu)

Abstract: This paper is a survey of the theory of stochastic optimal economic growth.

Key words: economic growth, uncertainty, stochastic growth

May 2004

Copyright @ 2004 by Lars J. Olson and Santanu Roy.

All rights reserved. Readers may make verbatim copies of this document for non-commercial purposes by any means, provided that this copyright notice appears on all such copies.

Theory of Stochastic Optimal Economic Growth.

Lars J. Olson, University of Maryland, College Park, MD.
Santanu Roy, Southern Methodist University, Dallas, TX.

May 4, 2004

1 Introduction

Stochastic optimal growth involves the study of the optimal intertemporal allocation of capital and consumption in an economy where production is subject to random disturbances. The theory traces its roots to the seminal work on deterministic optimal growth by Ramsey [106], Cass [21] and Koopmans [55]. Its influence has been enhanced by research that shows how the convex stochastic growth model can be decentralized to represent the behavior of consumers and firms in a dynamic competitive equilibrium of a productive economy ([102], [115], [15]). This makes the stochastic optimal growth model useful both as a normative exercise and in the development of positive theories of how the economy works. As a consequence, the theory has emerged as one of the central paradigms of dynamic economics. It is based on a simple, yet powerful model that encompasses fundamental questions that are basic to any theory of dynamic economic behavior: What are the characteristics and determinants of optimal policies? What are the economic incentives that govern the optimal intertemporal allocation of resources? What is the transient and long run behavior of variables in the model? Under different assumptions the model admits a rich set of answers to these questions.

Historically, the main focal point of the theory has been issues of aggregate economic growth. At the same time its primary variable, capital, has a flexible interpretation that allows the model and its extensions to represent a wide variety of economic problems ranging from the study of business cycles ([59], [63]) and asset pricing ([14], [15]) to the allocation of renewable natural resources ([77], [82], [83]). Equally important, the model provides a strong theoretical foundation for applied analysis of these problems. The model can be solved numerically and has proved a testing ground for many numerical techniques used today in the analysis of dynamic economic problems.

This chapter provides an overview of key results the *theory* of discounted stochastic optimal growth in discrete time.¹ The paper begins with an analysis of

¹There is a large literature on stochastic growth in continuous time that builds on Merton's

the classical stochastic growth model of Brock and Mirman [18] for a one-sector economy with a convex technology and utility that depends only on consumption.² We then consider extensions of the theory to problems with irreversible investment, increasing returns or a non-convex technology, experimentation and learning, and problems where utility depends on more than consumption alone. We develop the competitive price characterization of optimal policies that can be used to establish the equivalence between optimal and competitive outcomes; our focus, however, is on optimal solutions and their properties. The large literature on dynamic competitive equilibria is, therefore, left to the reader to explore. Likewise, we do not survey the many applications of the stochastic growth model. Instead, we focus on how the theory can be extended in different directions that have proved useful in application. Finally, we provide a glimpse of practical methods for solving the model, but the literature on numerical methods is too large for us to review here.

2 The Classical Framework

2.1 The One Sector Classical Model: Basic Properties

The stochastic growth model has three essential elements: an exogenous stochastic environment corresponding to random productivity disturbances, the production possibilities that determine the set of feasible allocations for consumption, investment and output, and an instantaneous welfare or utility function that represents the preferences of the agent or economic decision-maker. Productivity shocks at dates $t = 1, 2, \dots$, are denoted by $\{r_t\}$, a sequence of i.i.d. real-valued random variables, with common distribution ν on $B(\Phi)$, the Borel σ -field of $\Phi \subset \mathbb{R}$. In particular, Φ is the support of ν and is assumed to be compact. Associated with this stochastic environment is a measure space $(\Omega, \mathcal{F}, \mu)$, where Ω is the set of all real sequences, $\mathcal{F}$ is the σ -field generated by cylinder sets of the form $\prod_{t=0}^{\infty} A_t$, where A_t belongs to $B(\Phi)$ for all t , and μ is the product distribution induced by ν . The statements: for a.e. ω and μ -a.s. mean "except for a subset of Ω of μ -measure zero". The random variable r_t is simply the t^{th} coordinate function on Ω . In the economy, output of a homogeneous consumption/capital good is produced via a production function that is homogeneous of degree one in capital and labor. This allows the economy to be represented in per capita terms where c_t, k_t and y_t denote per capita consumption, capital and output at time t . Given a capital stock at time $t - 1$ and the productivity disturbance at the beginning of period t , $y_t = f(k_{t-1}, r_t)$, where $f : \mathbb{R}_+ \times \Phi \rightarrow \mathbb{R}_+$ is the production function. The *feasible set* for consumption and investment is: $\Gamma(y_t) = \{(c_t, k_t) | 0 \leq c_t, 0 \leq k_t, \text{ and } c_t + k_t \leq y_t\}$.

Each period the economic agent receives utility $u(c_t)$, where $u : \mathbb{R}_+ \rightarrow \mathbb{R}$ and the discount factor for future utility is δ , where $0 \leq \delta < 1$. At the beginning of period t the agent observes y_t and chooses c_t and k_t . The productivity

[78] early work (see also, [16]).

²Previous surveys of stochastic growth such as [81] and [6] focus primarily on this case.

disturbance, r_{t+1} , occurs and a new output, y_{t+1} , is produced. The objective of the agent is to maximize the expected discounted sum of utility over time subject to the feasibility constraints on consumption and capital, and the transition equation that maps capital to output in the following period. Given an initial output, y_0 , the objective is to:

$$\text{Max } E \left[\sum_{t=0}^{\infty} \delta^t u(c_t) \right] \quad \text{subject to: } 0 \leq c_t, 0 \leq k_t, c_t + k_t \leq y_t, y_t = f(k_{t-1}, r_t), t \geq 1. \quad (1)$$

This problem can be formulated as a stochastic dynamic programming problem ([11], [125] and [65]). At date t the *partial history* is $h_t = \{y_0, c_0, k_0, y_1, \dots, c_{t-1}, k_{t-1}, y_t\}$. A *policy*, π , is a sequence $\{\pi_0, \pi_1, \dots\}$, where π_t is a conditional probability on $B(\mathfrak{R}_+)$, given h_t , such that $\pi_t(\Gamma(y_t) \mid h_t) = 1$. Let F be the set of all measurable functions ϕ such that $\phi(y) \in \Gamma(y)$ for all $y \in \mathfrak{R}_+$. A policy is *Markovian* if $\pi_t \in F$ for all t . A Markov policy is *stationary* if there exists a Borel measurable function, $\hat{\pi}(y)$, such that $\pi_t(y) = \hat{\pi}(y)$ for all t . A policy, π , and an initial state, y , induce a *feasible program*, $(\mathbf{y}, \mathbf{c}, \mathbf{k}) = (y_t, c_t, k_t)_{t=0}^{\infty}$, a stochastic process for output, consumption and capital such that $(c_t, k_t) \in \Gamma(y_t)$ and $y_{t+1} = f(k_t, r_{t+1})$ a.s. for all t . Associated with each policy is an expected discounted sum of utility $V_{\pi}(y) = E \sum_{t=0}^{\infty} \delta^t u(c_t)$, where $(\mathbf{y}, \mathbf{c}, \mathbf{k})$ is the feasible program generated by π and f in the obvious manner. A policy, π^* , is optimal if $V_{\pi^*}(y) \geq V_{\pi}(y)$ for all π and y , and the associated program is called an *optimal program*. The value function $V(y)$ is defined on $\mathfrak{R}_+$ by $V(y) = \sup\{V_{\pi}(y) \mid \pi \text{ is a policy}\}$. It follows that π^* is an optimal policy if, and only if, $V_{\pi^*}(y) = V(y)$ for all $y \geq 0$.

Throughout the paper, derivatives are denoted using subscripts, so that u_c represents marginal utility and so on. The production technology and preferences are assumed to satisfy the following assumptions:

- A.1. $f(0, r) = 0$, $f(k, r) > 0$ for all $r \in \Phi$ and all $k > 0$.
- A.2. f is continuous on $\mathfrak{R}_+ \times \Phi$ and for each $r \in \Phi$, $f(\cdot, r)$ is continuously differentiable on $\mathfrak{R}_{++}$.
- A.3. $f_k(k, r) > 0$ and $\inf_{r \in \Phi} f_k(0, r) > 1$.
- A.4. $f(\cdot, r)$ is strictly concave on $\mathfrak{R}_+$ for all $r \in \Phi$.
- A.5. There exists a $\bar{k} > 0$ such that $f(k, r) < k$ for all $k > \bar{k}$ and all $r \in \Phi$.
- A.6. u is continuous on $\mathfrak{R}_+$ and continuously differentiable on $\mathfrak{R}_{++}$.
- A.7. $u_c(c) > 0$ on $\mathfrak{R}_{++}$.
- A.8. u is strictly concave on $\mathfrak{R}_+$.

Under these assumptions the dynamic optimization problem is well defined, the value is finite from any initial state and it satisfies the functional equation:

$$V(y) = \max_{c \in \Gamma(y)} u(c) + \delta \int V(f(y - c, r)) dv(r). \quad (2)$$

Further, there exist stationary optimal policy functions for consumption, $C(y) = \arg \max_{c \in \Gamma(y)} u(c) + \delta \int V(f(y - c, r)) dv(r)$, and capital, $K(y) = y - C(y)$.³

To characterize economic behavior in the model it is important to understand the basic properties of the optimal value and policy functions. Further, such knowledge is necessary to examine how departures from the classical model affect economic outcomes. In the classical model, the feasible set is expanding with a convex graph,⁴ and the production and utility functions are strictly increasing and strictly concave. Using the fact that the functional equation (2) maps the set of continuous, increasing and concave functions into itself this implies (e.g., [124]):

Lemma 1 . *Under A.1-A.8, $V(y)$ is continuous, strictly increasing and strictly concave.*

The value function is a measure of lifetime economic welfare and to a first order approximation is proportional to traditional measures of GDP. The economic implication of Lemma 1 is that small increases in output have small effects on welfare, and that welfare increases at a diminishing rate as output increases.

In the classical model the production, utility, and value functions are strictly concave, so the optimization problem has a unique solution and the optimal policy functions are continuous. Monotonicity properties of $C(y)$ and $K(y)$ are determined by the complementarity⁵ between k and y , and c and y , respectively.

Lemma 2 . *Under A.1-A.8, $C(y)$ and $K(y)$ are single-valued, continuous and increasing functions.*⁶

Proof. The fact that $C(y)$ and $K(y)$ are single-valued and continuous follows from the maximum theorem and the strict concavity of u and f . Let $k \in K(y)$ and $k' \in K(y')$ for $y < y'$. Suppose that $k' < k$. Then $k' \in \Gamma(y)$ and $k \in \Gamma(y')$. Further, $0 < u(y - k) + \delta EV(f(k, r)) - [u(y - k') + \delta EV(f(k', r))] < u(y' - k) + \delta EV(f(k, r)) - [u(y' - k') + \delta EV(f(k', r))] < 0$, where the first and last inequalities follow from the principle of optimality and the middle inequality follow from the fact that A.8 implies u is strictly supermodular in (y, k) . Hence, it must be that $k' > k$. Next suppose that $c' \leq c$. Then, $0 \leq u(c) + \delta EV(f(y - c, r)) - [u(c) + \delta EV(f(y - c', r))] + u(c') + \delta EV(f(y' - c', r)) - [u(c') + \delta EV(f(y' - c, r))] < 0$, where the first inequality follows from the principle of optimality and the last inequality is due to the strict concavity of f and V . Hence, $c' > c$. ■

³Note that existence and all other results in this section continue to hold for logarithmic or CES utility functions that are unbounded below, though $V(0) = u(0) = -\infty$ (e.g., [116]).

⁴The feasible set is expanding if $y \leq y'$ implies $\Gamma(y) \subseteq \Gamma(y')$ and has a convex graph if $\{(c, k, y) \mid (c, k) \in \Gamma(y)\}$ is a convex set.

⁵Formally this is represented by the concept of supermodularity. Let $y \wedge y' = \min[y, y']$ and $y \vee y' = \max[y, y']$. A function $F(k, y)$ is supermodular in (k, y) if $F(k \wedge k', y \wedge y') + F(k \vee k', y \vee y') \geq F(k, y) + F(k', y')$. For C^2 functions this is equivalent to $F_{ky} \geq 0$ so that an increase in one argument raises the marginal value or marginal productivity of the other.

⁶Lemma 2 was first established by Brock and Mirman [18]. The monotonicity of $K(y)$ does not depend on the concavity of u or f and can be generalized to the case where $K(y)$ is a correspondence using the methods of Topkis [127] (see also, [119]).

When the stochastic growth model is representative of aggregate economic behavior, it is natural that consumption and investment should always be in the interior of the feasible set. In disaggregate or microeconomic settings, this may not always be true. Since the interiority of optimal policies facilitates the use of differentiable optimization methods it is common to impose an assumption that guarantees interiority.

$$A.9. \lim_{c \downarrow 0} u_c = \infty.$$

Lemma 3 . Under A.1-A.9, $C(y) > 0$ and $K(y) > 0$.

The condition $\lim_{c \downarrow 0} u_c = \infty$ is known as the Inada [44] condition at zero. In the classical model, the intuition for its use is as follows. To invest y yields finite discounted expected marginal value of investment but an infinite marginal utility from consumption. Hence, one can do better by reallocating some output from investment to consumption. Analogous arguments can be used to rule out investment of zero.

When optimal policies are interior, the value function in the classical model is differentiable.

Lemma 4 . (Mirman-Zilcha [84], Lemma 1). Under A.1-A.9, $V(y)$ is differentiable for all $y > 0$ and $V_y(y) = U_c(C(y))$.

Proof. As a concave function, V has left and right-hand derivatives, $V_-(y) \geq V_+(y)$. Let k and c be optimal from y . As $c > 0$, k is feasible from $y + \epsilon$ and $y - \epsilon$ for sufficiently small $\epsilon > 0$. By optimality, $V(y + \epsilon) - V(y) \geq u(c + \epsilon) + \delta EV(f(k, r)) - [u(c) + \delta EV(f(k, r))]$, which implies $V_+(y) \geq u_c(c)$. By a similar argument $V_-(y) \leq u_c(c)$.⁷ ■

In this case output has a unique shadow price given by $V_y(y)$. This shadow price is useful in examining the intertemporal tradeoff between consumption and investment, and in showing that the optimization problem can be decentralized.

Proposition 5 Let (c, k) be an optimal program induced by $C(y), K(y)$. Under A.1-A.9, necessary and sufficient conditions for $C(y), K(y)$ to be optimal are:

$$u_c(c_t) = \delta \int u_c(c_{t+1}(r)) f_k(k_t, r) dv(r). \quad (3)$$

$$\lim_{t \rightarrow \infty} \delta^t E[u_c(c_t) k_t] = 0. \quad (4)$$

⁷ An alternative approach in [12] assumes that the disturbance distribution has a C^n density. This smooths out possible points of discontinuity in the derivative of V . The approach has the advantage that it can be used to obtain higher order differentiability of both V and the optimal policy function, the latter via the implicit function theorem. Santos and Vigo-Aguiar[114] also contains sufficient conditions for the value and policy functions to be C^2 and C^1 , respectively. They use their results to place analytical bounds on the approximation error of a numerical solution.

Proof. The necessary condition (3) is typically proved in one of two ways. The first method is a variational approach that assumes period t output and the period $t+1$ capital stock are optimal. It then examines how a change in period t consumption affects discounted expected utility across the two periods. The second method proceeds as follows. If V is differentiable (Lemma 4) then maximizing the right hand side of equation (2) implies: $u_c(c_t) = \delta \int V_y(f(k_t, r)f_k(k_t, r)dv(r)$. Further, $V_y(y_t) = u_c(c_t)$ by the envelope theorem. Combining these yields (3). As commonly used, this approach requires both interior solutions and a differentiable value function; but a more general statement using inequalities is possible in other cases.

A proof of (4) is given in [86]. ■

Equation (3) is known as the stochastic Ramsey-Euler equation. It is a dynamic optimality condition that equates the marginal utility from consumption to the discounted expected marginal value of investment. The latter can be decomposed into the marginal productivity of investment times the marginal utility from consuming the additional output next period.

Equation (4) is the transversality condition. It implies that marginal utility is bounded in expectation. Mirman and Zilcha [85] show that marginal utilities themselves may be unbounded. It is also important to note that there may be many non-optimal programs that satisfy the Ramsey-Euler equation. The transversality condition selects an optimal program from among those satisfying (3).

One of the most important results in the stochastic growth literature relates to the validity of the fundamental theorems of welfare economics in infinite horizon, stochastic economies. The two basic issues are the existence of prices that support an optimal program and the optimality of a dynamic, competitive equilibrium. In their seminal work Malinvaud [71] and Koopmans [54] make clear that the fundamental welfare theorems do not extend to infinite horizon settings without some additional conditions. The importance of these issues is apparent in [18], [84], [85] and [86] even though prices are often implicit in the necessary and/or sufficient conditions for optimality. Zilcha ([133], [134], [135]) examines the fundamental welfare theorems in a setting in which competitive prices are explicit throughout.

A feasible program $(\mathbf{y}, \mathbf{c}, \mathbf{k})$ is *competitive* if there exists a sequence $\mathbf{p} = (p_t)_{t=0}^{\infty}$ of discounted prices such that $p_t > 0$ a.s. for all t and:

$$\delta^t u(c_t) - p_t c_t \geq \delta^t u(c) - p_t c \text{ a.s. for all } c \geq 0. \quad (5)$$

$$Ep_{t+1}f(k_t, r_{t+1}) - p_t k_t \geq Ep_{t+1}f(k, r_{t+1}) - p_t k \text{ a.s.} \quad (6)$$

Proposition 6 *A feasible program is optimal if and only if it is competitive and satisfies:*

$$\lim_{t \rightarrow \infty} Ep_t k_t = 0. \quad (7)$$

Proof. See [133]. As in Proposition 5, the existence of competitive prices alone is not sufficient to guarantee optimality. For that, the transversality condition (7) is also required. ■

It should be clear that the supporting price p_t is the discounted shadow price of the consumption-capital good. Equation (5) requires that for almost every realized path and every time period, consumption maximize "reduced discounted utility" (utility of consumption minus expenditure). Equation 6 captures intertemporal (expected) profit maximization. When a competitive program is interior it implies $p_t = Ep_{t+1}f_k(k_t, r_{t+1})$. A primary difference between the deterministic and stochastic models is that in the former prices reflect temporal values, while in the latter prices reflect both temporal values and values across different random states of nature. As a consequence, prices and the marginal willingness to substitute consumption are an important determinant of economic behavior even in the long run.

2.2 Stochastic Steady States and Convergence Properties in the One Sector Classical Model

A central concern of optimal growth theory is the study of the long run dynamics of an economy. The deterministic literature focusses on the existence and stability of non-trivial (strictly positive) optimal steady states and on turnpike properties of optimal capital accumulation paths.

An optimal steady state or stationary program is a limit point of an optimal program. If optimal paths from all initial states converge to a steady state then this unique optimal steady state is globally stable and the long run behavior of the economy is independent of initial conditions. When the evolution of capital stocks is stochastic, an optimal program of capital stocks is a sequence of random variables. The optimal policy, the production function, and the random shock map the probability distribution of current capital stocks to the probability distribution of the next period's capital stock. A stochastic steady state is a fixed point of this mapping or a distribution of capital that is invariant under the optimal policy. The stochastic analogue of a globally stable steady state is a unique invariant distribution to which the stochastic process of capital stocks converges from any initial state. In such a steady state the capital stock is not constant over time. Instead, it exhibits endogenous fluctuations in response to random productivity disturbances.⁸

Turnpike theorems study the conditions under which differences in initial conditions have negligible effects on the process of economic growth over long time horizons.⁹ In the deterministic case this involves analyzing when optimal paths from different initial states approach each other asymptotically. The stochastic analogue is convergence in probability, or sometimes, almost sure convergence to zero of an appropriately defined distance between the optimal capital stocks in each period.

⁸One may also look at stronger concepts of optimal steady states [131].

⁹See, [76], [74], [75].

In the classical one sector stochastic optimal growth model the unique optimal stationary policy generates a Markov process of capital stocks k_t . Recall that the optimal investment function $K(y)$ is a continuous and strictly increasing function on $\mathfrak{R}_+$. Define $H(k, r) \equiv K(f(k, r))$ to be the realized capital stock for the next period under the optimal policy. Then, H is continuous in (k, r) and increasing in k . Let S denote the interval $[0, \bar{k}]$. Given $y_0 \in S$, $k_0 = K(y_0) \in S$, the evolution of optimal capital stocks over time is given by:

$$k_t = H(k_{t-1}, r_t) \quad (8)$$

Recall that ν is the common probability distribution of the i.i.d random shocks r_t , with support Φ , a compact subset of $\mathfrak{R}$. Let ν^t be the joint distribution of $r^t \equiv (r_1, \dots, r_t)$ on the product space Φ^t and define $k^n(k_0, r^n) \equiv H(H(\dots(H(k_0, r_1), r_2), \dots), r_n)$. In other words, $k^n(k_0, r^n)$ is the realization of n^{th} -period capital stock k_n , given k_0 and realization $r^n = (r_1, \dots, r_n)$ of random shocks in the first n periods. For any probability measure μ defined on S (and the Borel σ -field generated by S), define the probability measure $\nu^n \mu$ on S by the relation

$$\nu^n \mu(B) = \int_S \nu^n(\{r^n \in \Phi^n \mid k^n(k_0, r^n) \in B\}) \mu(dk_0)$$

where B is any Borel-subset of S . Thus, $\nu^t \mu$ gives us the probability distribution of k_t , when k_0 is distributed according to the probability measure μ . Let S' be a closed interval in S . Then, S' is said to be ν -invariant if $\nu(\{r \in \Phi \mid H(k, r) \in S' \text{ for all } k \in S'\}) = 1$. A probability measure μ on S is said to be an invariant probability measure on S' if the support of μ is a subset of S' and for any Borel set B in S ,

$$\nu \mu(B) = \mu(B) \quad (9)$$

In other words, if k_0 is distributed according to an invariant probability μ , then the distribution of optimal capital stocks in every subsequent period follows the same distribution. The distribution function corresponding to an invariant probability measure is an invariant distribution.

There is a large body of work in the mathematical theory of Markov processes and random dynamical systems that provides sufficient conditions for the existence and stability of invariant distributions for a given stochastic process^{10, 11}. Let

$$H_m(k) = \min_{r \in \Phi} H(k, r) \text{ and } H_M(k) = \max_{r \in \Phi} H(k, r)$$

denote the lower and upper envelopes, respectively, of the transition function $H(k, r)$ defining the Markov process (8). Note that the continuity of H and

¹⁰See, among many others, [35], [41], [42], [2], [8], [9].

¹¹Models of descriptive stochastic growth (such as the stochastic Solow model) where the consumption and investment rules are exogenously specified have also applied these conditions. See, [80], [13] and [107].

the fact that Φ is compact imply that $H_m(k)$ and $H_M(k)$ are well defined and continuous. Further, since H is increasing in k , $H_m(k)$ and $H_M(k)$ are increasing functions.

In addition to the assumptions made in the previous section, the standard proof of existence and global stability of the invariant distribution requires that the production function $f(k, r)$ and the optimal transition function $H(k, r)$ satisfy two additional conditions.

A.10. There does not exist any $k > 0$ and $\tilde{y} \in S$ such that $\nu\{r \mid f(k, r) = \tilde{y}\} = 1$.

A.11. There exists an $\epsilon > 0$ such that $H_m(k) > k$ for all $k \in (0, \epsilon)$.

A.10 requires that every investment level is associated with some non-trivial uncertainty over output. A.11 is a restriction on the optimal policy. It implies that even if the lowest possible output is realized every period, the optimal program from every initial stock is bounded away from zero. In a deterministic model, the optimal policy satisfies this condition as long as marginal productivity at zero is large enough. This is no longer true in the stochastic model. Mirman and Zilcha [85] develop an example where the production function has infinite slope at zero, yet the optimal program from any initial stock comes arbitrarily close to zero with probability one.¹² One can impose restrictions on the production function and distribution of random shocks to ensure that A.11 is satisfied. For example, if there is a strictly positive probability mass on the "worst" production function in the sense that there exists some $\varepsilon > 0$ such that $\nu\{r \mid f(k, r) = \min_{r \in \Phi} f(k, r)\} > \varepsilon, \forall k > 0$, then infinite marginal productivity at zero is sufficient for A.11. For conditions that are applicable to atomless distributions, see [92].

Define the maximal fixed point of H_m by $k_m = \max\{k > 0 \mid H_m(k) = k\}$ and the minimal fixed point of H_M by $k_M = \min\{k > 0 \mid H_M(k) = k\}$. Assumption A.11 implies that $k_m, k_M > 0$.

Lemma 7 $k_m < k_M$.

Proof. Since $H(k, r)$ is continuous in r , there exists $r_m, r_M \in \Phi$ such that $k_m = H_m(k_m) = H(k_m, r_m)$ and $k_M = H_M(k_M) = H(k_M, r_M)$. Further, $f(k_m, r_m) \leq f(k_m, r)$ for all $r \in \Phi$. From the stochastic Ramsey-Euler equation:

$$\begin{aligned} u'(C(f(k_m, r_m))) &= \delta \int_{\Phi} u'(C(f(H(k_m, r_m), r))) f'(H(k_m, r_m), r) \nu(dr) \\ &= \delta \int_{\Phi} u'(C(f(k_m, r))) f'(k_m, r) \nu(dr). \end{aligned}$$

Since u is strictly concave and C is increasing $u'(C(f(k_m, r))) \geq u'(C(f(k_m, r_m)))$ for all $r \in \Phi$. Hence, the inequality above yields $1 \leq \delta \int_{\Phi} f'(k_m, r) \nu(dr)$. Similarly, one can show that $1 \geq \delta \int_{\Phi} f'(k_M, r) \nu(dr)$ so that $\int_{\Phi} f'(k_M, r) \nu(dr) \leq$

¹²Mitra and Roy [92] develop general conditions under which $Prob\{\liminf_{t \rightarrow \infty} k_t = 0\}$ is 0 and 1.

$\int_{\Phi} f'(k_m, r) \nu(dr)$. The fact that $k_m \leq k_M$ follows from the strict concavity of f . Finally, if $k_m = k_M$ then $f(k, r)$ is constant in r which violates A.10. ■

Lemma 7 implies that the highest fixed point of H_m lies below the smallest positive fixed point H_M (see Figure 1). We now state the main result regarding the existence and global stability of the optimal stochastic steady state. For the stochastic process of optimal capital stocks k_t defined by (8), let $F_t(k)$ be the distribution function of k_t i.e., $F_t(k) = \nu^t\{r^t \in \Phi^t \mid k_t \leq k\}$.

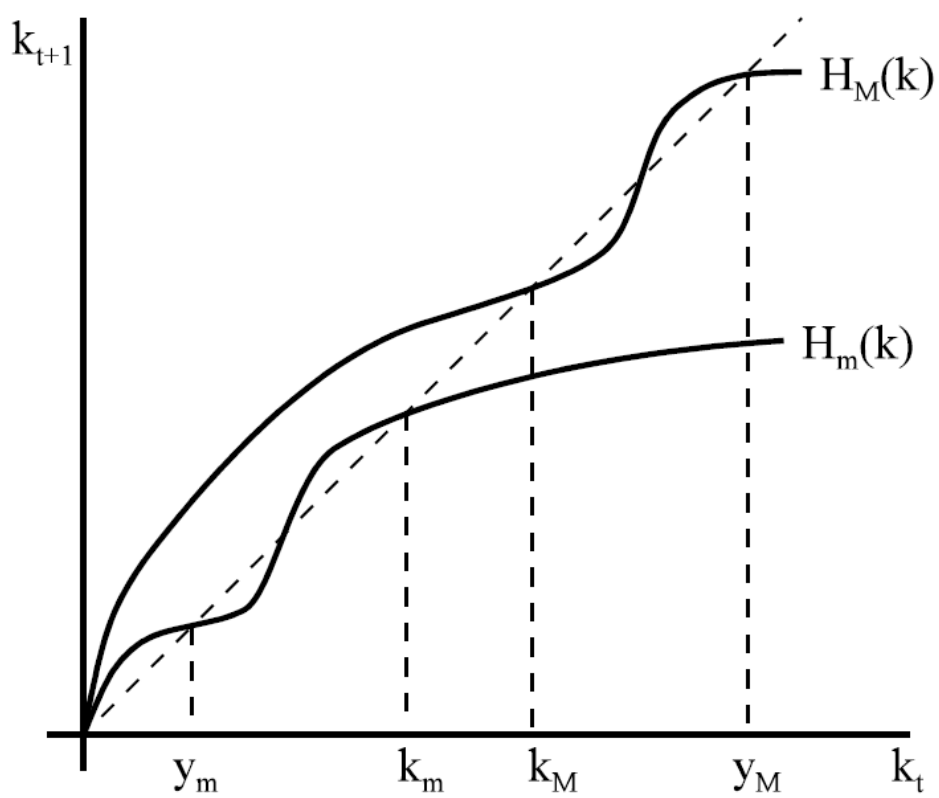
Proposition 8 *Assume A.1 - A.11. There exists a unique non-zero invariant distribution $F(k)$ on S and its support is the interval $[k_m, k_M]$. For any initial capital stock $k_0 > 0$, as $t \rightarrow \infty$, $F_t(k)$ converges uniformly in k (on S) to $F(k)$.*

Proof. (Sketch). Instead of giving a full proof, we sketch the main arguments for the simple case of multiplicative shock ($f(k, r) = rf(k)$) which assumes just two possible values a and b , $0 < a < b < \infty$. Then, $H_m(k) = K(af(k))$ and $H_M(k) = K(bf(k))$. The proof consists of the following key arguments. First, for the Markov process (8), the set of states $(0, k_m)$ and (k_M, ∞) are transient. With probability one, capital stocks move out of these sets in finite time, never to return. Second, once the process enters the set $[k_m, k_M]$ it remains there with probability one. Further, $[k_m, k_M]$ is the smallest ν -invariant set. The intuition behind these arguments can be readily seen from Figure 1. Let $y_m = \min\{k : H_m(k) = k\}$ and $y_M = \max\{k : H_M(k) = k\}$. Then, $0 < y_m \leq k_m < k_M \leq y_M$. From any stock $k \in (0, y_m)$ the optimal capital stocks increase almost surely and reach the set $[y_m, k_M]$ in finite time with probability one. Similarly, from any stock $k \in [y_M, \infty)$ the optimal capital stocks decrease almost surely and reach the set $[k_m, y_M]$ in finite time with probability one. Further, for $k \in [y_m, k_m)$ one can show that the probability that the optimal path from such a stock does not enter $[k_m, k_M]$ in finite time is zero. To move the capital stock y_m to the interval $[k_m, k_M]$ only takes a sufficiently long, but finite run $r_t = b$, such that the realized transition occurs along the function $H_M(k)$. Any such run must occur ω -almost surely as shocks are independent. The same argument shows that $(k_M, y_M]$ is transient. It can also be seen from Figure 1 that no strict subset of $[k_m, k_M]$ is invariant. The next step is to show that a well-known "splitting" condition due to Dubins and Freedman [35] (or some variation/extension) holds on the interval $[k_m, k_M]$. For any $n = 1, 2, \dots$, the probability ν^n is said to split on a ν -invariant subset S' of S if there exists $z \in S'$ and $\eta > 0$ such that:

$$\begin{aligned} \nu^n\{r^n \in \Phi^n \mid k^n(k, r^n) \leq z \text{ for all } k \in S'\} &> \eta \\ \nu^n\{r^n \in \Phi^n \mid k^n(k, r^n) \geq z \text{ for all } k \in S'\} &> \eta. \end{aligned}$$

To verify that the splitting condition holds fix any $z \in (k_m, k_M)$. There exists some $N \geq 1$, such that: (i) if $r_t = a, t = 1, \dots, N$, then $k^N(k_M, r^N) \leq z$, and (ii) if $r_t = b, t = 1, \dots, N$, then $k^N(k_m, r^N) \geq z$. For $0 < \eta < \min\{(\nu(a))^N, (\nu(b))^N\}$, $n = N$, it is easy to see that the splitting condition is satisfied on S' . Dubins

Figure 1: Figure 1



and Freedman [35] then show that this implies there exists a unique invariant distribution F on S' and that $F_t(k)$ converges uniformly in k to $F(k)$.¹³ Finally, since the set $S - S'$ is transient and S' is the smallest ν -invariant set on S , it must be that F is the unique invariant distribution on S and $F_t(k)$ converges uniformly in k to $F(k)$ on S .¹⁴ ■

The basic results on the existence and global stability of an invariant distribution for the classical one sector stochastic model were originally developed in the pioneering work by Brock and Mirman [18] and subsequently refined by Mirman and Zilcha [84]. Majumdar, Mitra and Nyarko [68] were the first to explicitly use the Dubins-Freedman splitting condition. Versions of this problem have also been analyzed by [124] and [42]. [19] [25] contain similar results for the undiscounted model ($\delta = 1$) where optimality is based on the "overtaking criterion".¹⁵ Donaldson and Mehra [34] extend these results to the case of correlated shocks that enter the production function multiplicatively and follow a stationary process.

From an empirical point of view one may be interested in the asymptotic statistical properties of the stochastic processes for capital and consumption. For example, if the law of large numbers holds so that sample averages from time series converge to the mean of the limiting steady state distribution, then one can test a model by comparing the sample average over a sufficiently long period with the theoretical prediction. Alternatively, one can forecast the mean of the long run distribution by using the sample average. The central limit theorem or asymptotic normality of the partial sums can be used for inference of likelihood of values in a parameter space. Many of the conditions that guarantee global stability of an invariant distribution also ensure that both the law of large numbers and the central limit theorem hold. In addition, they imply a minimum bound on the rate of convergence.¹⁶

An important implication of global stability is that the long run behavior of the economy is independent of the initial state. The latter aspect is also brought out in turnpike results that directly examine the conditions under which differences in initial conditions have negligible effects on the process of economic growth over long time horizons. Majumdar and Zilcha [70] establish a "late"

¹³For recent extensions of the result that is applicable to situations where $H(k, r)$ is monotonic but not necessarily continuous and situations where the capital process is multidimensional, see [8],[9].

¹⁴A more traditional approach in theory of Markov processes is to directly verify that the process is irreducible on $[k_m, k_M]$, that intervals disjoint from $[k_m, k_M]$ are transient and an equicontinuity condition on the sequence of probability measures for the capital stock (defined through the stochastic kernel of the Markov process). See, [77]. Another approach is to show that the iterated random functions satisfy a Lipschitz condition and are "contracting on the average" (see, [33]). Stachurski [122] shows that there is always a unique globally stable steady state for the special case of a multiplicative shock where r has a density function that is strictly positive everywhere on $\mathfrak{R}_{++}$. With an interior optimal policy, the structure imposed on the random shock ensures that the system moves with positive probability from any positive stock to any interval on $\mathfrak{R}_{++}$ in *one* step.

¹⁵For convergence in a stochastic open economy, see for example, [28].

¹⁶See, [9]. [123] contains similar results for the case of multiplicative unbounded shocks.

turnpike theorem in a model that is far more general than the classical model of Section 2. Their model allows for unbounded expansion of capital and consumption, time varying utility and a non-stationary stochastic processes of random shocks. Under a condition that requires the elasticity of marginal product to be bounded away from zero (implying a lower bound on the degree of concavity of the production function), they show that the number of periods for which the relative distance between the optimal capital stocks (from any two initial stocks) exceeds any positive threshold is bounded almost surely, where the bound depends on how far apart the initial states are. In other words, optimal paths from different initial states eventually approach each other with probability one. Note that this result is quite independent of whether there is a globally stable invariant distribution. The condition on the elasticity of marginal product ensures that a uniform "value-loss" argument (originally due to Radner [103]) holds.¹⁷ Joshi [49] provides similar turnpike results in a one-sector model with recursive preferences and time varying technology.

Apart from convergence, the other important question in economic growth relates characterization of the properties of the limiting steady state. In other words, what can we say about the relationship between the preferences and technology underlying the economy and the nature of the invariant distribution to which it converges. In the one sector convex deterministic model of optimal growth, there is fairly rich characterization of the steady state. For example, with a strictly concave production function f , the unique steady state or modified golden rule is a capital stock $\hat{k}$ that is the unique maximizer of $[\delta f(k) - k]$, where the latter can be interpreted as the net gain from investment. For the no-discounting case, the steady state is the well-known golden rule capital stock that maximizes the level of sustainable consumption $[f(k) - k]$. There are other decentralized or support price-based characterizations of the optimal steady state. More generally, in the deterministic one sector model, it is possible to look at the steady state as a solution to an independent *static* optimization problem that has desirable economic properties. In the multisector deterministic model, it is a solution to a static optimization and a fixed point problem.

What are the corresponding results in the stochastic case? The answer, surprisingly, is close to none. In fact, there is very little by way of general qualitative characterization of the limit invariant distribution in the stochastic growth literature. One of the reasons behind this is the fact that, unlike the deterministic model, the steady state is not determined solely by the production function and discount factor. Both the utility function (and its curvature) and the distribution of the random shocks play important roles. Specific examples show that for the same technology, discount factor and distribution of random

¹⁷The value loss argument uses support prices of optimal paths to look at the accumulation of shortfalls in values (shadow profits and losses) of input-output combinations along one optimal path relative to another at the other's support prices. Loosely speaking, for two optimal paths that do not approach each other asymptotically, if the value loss is uniformly bounded away from zero over all states and time periods, then the accumulated loss is infinite and that contradicts optimality. [51] contains a turnpike result without requiring uniformity of value-loss across time and states.

shocks, the steady state distribution can change dramatically with variations in the utility function [26]. Further, even for very standard utility and production functions, the form limiting distribution can be very sensitive to parameter values when the shock does not have a continuous distribution.¹⁸ For the case with logarithmic utility, Cobb-Douglas production, and a binary multiplicative shock, Mirman and Zilcha [84] show that the invariant distribution can be degenerate for some parameter values and uniform for others. Montrucchio and Privileggi [93] show that the invariant distribution can also be a Cantor function. Mitra, Montrucchio and Privileggi [91] expand on this example to establish precise bounds on the parameters under which Cantor and more general singular invariant distributions can arise as well as bounds under which the distribution is absolutely continuous. Recently, Mitra and Privileggi [90] extend the example to the class of all iso-elastic utility functions and establish sufficient conditions for a Cantor type invariant distribution.

2.3 Stochastic Steady States and Convergence Properties in the Multisector Classical Model

In the literature on deterministic models of optimal economic growth, the multisector case has been extensively studied. In particular, the literature has focused on two key issues - the existence of an optimal steady state and turnpike results or the convergence properties of optimal paths.¹⁹ In comparison, the stochastic multisector optimal growth literature is relatively thin and there is only a small literature on the existence and stability of steady states in the stochastic, multisector case.

In the deterministic literature, it is well recognized that with discounting, the existence of a globally stable optimal steady state and other turnpike results may not hold in the multisector case, even though it always holds under very mild restrictions in the one-sector model.²⁰ With significant discounting, optimal paths in the multisector model may not be convergent. They may exhibit cyclical and even chaotic dynamics.

A general stochastic multisector optimal growth model with i.i.d. shocks has been analyzed by Brock and Majumdar [17]. The model is a natural extension of the classical one-sector model to the case of m capital goods. For each vector of current capital stocks and realization of the random shock there is a correspondence that defines the set of attainable utilities and capital stocks for the next period, which in turn can be used to define the set of feasible *programs* from any given initial vector of capital stocks. The objective is to maximize the discounted expected sum of utilities, or for the undiscounted case, a stochastic version of the overtaking criterion. The paper imposes four conditions:

- (i) there is a compact set $S' \subset \mathbb{R}_+^m / \{0\}$ such that for any initial vector

¹⁸For the case of multiplicative shock with continuous density, Danthine and Donaldson [27] show that the limiting invariant distribution has a continuous density function.

¹⁹For an excellent review of the basic results see McKenzie [75].

²⁰See, [126].

of capital stocks lying in S' , there exists an *optimal* program such that the stochastic process for capital lies almost surely in S' .

(ii) there exist continuous stationary optimal investment and consumption policies.

(iii) an optimal program is "competitive" relative to a non-trivial price process in a similar sense as in the previous section and satisfies a transversality condition that the expected values of the capital stocks (at the competitive prices) go to zero, for the case of discounting, and is bounded, in the undiscounted case.

(iv) the Hamiltonian system corresponding to the optimal processes has "suitable curvature" so that a stochastic value-loss condition is satisfied.

Under conditions (i) - (iv), Brock and Majumdar show that the distance between the probability distributions of t^{th} -period optimal capital vectors from two distinct initial capital vectors in S' converges to zero as $t \rightarrow \infty$. Further, the difference between the two optimal paths converges to zero in probability. Thus, conditions (i) - (iv) are sufficient to ensure that the optimal paths from alternative capital stocks come close to one another asymptotically and that the long run behavior of optimal paths does not depend on initial conditions. The existence of a unique and globally stable invariant distribution for the stochastic process of optimal capital stocks can also be established under these conditions. Unlike the conditions for global stability of an invariant distribution and other turnpike results in the one-sector stochastic growth model, (i) - (iv) are fairly strong restriction imposed directly on the optimal policy rather than the primitives of the model. Conditions (i) - (iii) are readily satisfied in the one sector stochastic growth model. In the multisector case there are plausible conditions on preferences and technology for (i) and (ii) to hold. For example, Majumdar and Radner [69] consider a stochastic nonlinear activity analysis model in which neoclassical conditions on the technology and preferences are sufficient for (i) and (ii).²¹ Condition (iii) is motivated by the equivalence between optimal programs and competitive programs that satisfy a transversality condition (see, [133], [134], [135]). Condition (iv) is a stochastic extension of conditions for asymptotic stability in the deterministic multisector model due to Cass and Shell [22] and Rockafellar [108] that are, in turn, based on the well known "value loss" argument alluded to in the previous subsection (see also, McKenzie [74]). In particular, condition (iv) requires that the Cass-Shell-Rockafellar version of the value-loss restriction holds uniformly for all states of the environment.

Chang [24] shows that a weaker version of condition (iv) based on *expected value loss* is actually sufficient and further, that the difference between any two optimal paths converges not only in probability, but almost surely. It is worth noting that in a multisector model condition (iv) involves a strong restriction on the extent of discounting in the model, and unlike the one-sector case, it does not follow directly from a restriction on the curvature of the production function.

²¹See also the discussion in [104].

Föllmer and Majumdar [39] follow a somewhat different approach using the theory of martingales to show that even if one does not impose a condition such as (iv), a weaker result is possible. That is, for any two optimal paths, the number of periods for which the value loss exceeds any given positive threshold is finite with probability one. Under uniformity of value loss and a specific distance metric, optimal paths approach each other asymptotically almost surely.

For the case of no discounting with the "overtaking" criterion of optimality, global stability of the stochastic state and other turnpike results can be established under much less restrictive conditions (see, among others, [46], [25], [134]).^{22,23}

3 Extensions of the Classical Framework

3.1 Sustained (Long Run) Growth

The past two decades have seen a renewed interest in the economics of long run growth where unbounded expansion of output, capital and consumption is possible. In the deterministic convex one-sector model, sustained growth is optimal if the marginal productivity of capital at infinity exceeds the discount rate [47]. Much of the literature on stochastic optimal growth theory focusses primarily on models where the technology exhibits bounded growth that rules out indefinite expansion of consumption and capital and sustained long run growth. An exception is the class of models on optimal intertemporal household savings under uncertainty. A portion of this literature considers a linear production function with a multiplicative shock, $f(k, r) = rk$, so that optimal paths may diverge to infinity (see, Phelps [101], Levhari and Srinivasan [62] and subsequent contributions). A closely related literature on the permanent income hypothesis has examined optimal savings where the wealth next period is composed of a deterministic return on current savings (interest income) plus an additive income shock (non-interest income).²⁴

De Hek and Roy [31] examine the possibility of sustained long run growth in optimal consumption and capital stocks in a one sector model with i.i.d. shocks and a concave production function that is not necessarily linear. Consider the model in Section 2 without assumption A.5. In particular, suppose that $f(k, r) = rf(k)$ and let $\theta = \lim_{k \rightarrow \infty} \frac{f(k)}{k}$. They show that under the following two conditions, optimal capital and consumption diverges to infinity with

²²Dutta [37] provides sufficient conditions under which as $\delta \rightarrow 1$, the optimal policies (and value functions) in the discounted stochastic model converge to the optimal policy using two alternative optimality criteria - the undiscounted overtaking criterion and the long run average reward criterion.

²³In a stochastic multisector model with a double infinity of time periods and discount factors close to 1, Yano [131] establishes the existence and continuity in the discount factor of a stronger concept of an optimal stationary program where a stationary program is one where the vector of current capital stocks associated with any infinite realized sequence of past history is time invariant with probability one). See also, [72].

²⁴See [130] and [117] for the undiscounted case, [118] and [120] for the discounted case and also [23].

probability one from every positive initial stock:

- (i) $E[\ln(\theta r)] > 0$
- (ii) $\inf_{y>0} \delta E\left[\frac{u_c(rf(sy))rf_k(sy)}{u_c((1-s)y)}\right] > 1$, where $s = \exp[-E[\ln(\theta r)]]$.

Note that these conditions involve the utility function and its curvature. The possibility of long run growth depend on more than a simple comparison of the discount rate and average marginal product at infinity. Once again, this reflects the general fact that in a stochastic growth model, the utility function and distribution of shocks play important roles in determining the nature of long run behavior of the economy. To illustrate this further we consider a specific example of iso-elastic utility and linear production function for which we can derive the optimal policy explicitly and thus provide an almost exact characterization of the condition for sustained long run growth.

Example 9 $u(c) = \frac{c^{1-\sigma}}{1-\sigma}$, $\sigma > 0$, $\sigma \neq 1$, $f(k, r) = rk$. One can show that the optimal policy function $K(y)$ is linear and given by $K(y) = [\delta E(r^{1-\sigma})]^{\frac{1}{\sigma}} y$ so that

$$k_{t+1} = \alpha r_{t+1} k_t \text{ where } \alpha = [\delta E(r^{1-\sigma})]^{\frac{1}{\sigma}} \quad (10)$$

which implies $\ln k_{t+1} = \ln k_0 + (t+1)[\frac{1}{t+1} \sum_{j=0}^t \ln \alpha r_{j+1}]$. Using the law of large numbers, it is easy to show that an "almost" exact condition for $\ln k_{t+1}$ to diverge to infinity with probability one is that $E[\ln(\alpha r)] > 0$ which can be rewritten as $\sigma E(\ln r) + \ln \delta + \ln E(r^{1-\sigma}) > 0$. This indicates that the risk aversion/intertemporal elasticity of substitution parameter of utility, σ , plays an important role in determining whether sustained growth occurs.

3.2 Stochastic Growth with Irreversible Investment

In the classical framework analyzed in the previous section, investment is either reversible or the existing capital stock depreciates completely at the end of a period. In reality, it is costly to transform capital into consumption and there are limits to how fast the aggregate capital stock depreciates. The stochastic growth model with irreversible investment was first examined by Sargent [115]. In his setting output can either be consumed or invested, but once invested, capital cannot be converted for consumption. Individual agents transact in a competitive market for existing capital. This allows individual investment decisions to be reversed while maintaining the irreversibility of investment in the aggregate. As Sargent shows, irreversibility in the aggregate provides the necessary friction for Tobin's q , the relative price of used to new capital, to diverge from unity. This enables aggregate investment to be positively correlated with q . However, this same friction implies that agents' investment decisions are necessarily a function of their expectations about the future which cannot be summarized by q . The implication is that q -theory of investment functions are of little use for econometric policy evaluation.

The analysis in Sargent is based on the properties of the value function. Olson [98] develops an alternative approach that characterizes optimal policies using stochastic Kuhn-Tucker conditions. Let $f(k, r) = F(k, r) + (1 - d)k$ where k is the depreciation rate of capital. If λ_t is the Lagrange multiplier on the period t irreversibility constraint, $k_{t+1} \geq (1 - d)k_t$, the Ramsey-Euler equation can be written as:

$$u_c(c_t) - \lambda_t = \delta E[u_c(c_{t+1}(r))] f_k(k_t, r) - (1 - d)\lambda_{t+1}(r). \quad (11)$$

Solving for λ_t and substituting forward this can be expressed as:

$$u_c(c_t) = \delta \sum_{i=1}^T (\delta(1 - d))^{i-1} E[u_c(c_{t+i}) F_k(k_{t+i-1}, r_{t+i})] + \delta^T (1 - d)^T E[u_c(c_{t+T})]. \quad (12)$$

This derivation uses the fact that eventual depreciation of the entire capital stock is not optimal so there is a uniform upper bound, T , on the number of time periods for which the irreversibility constraint binds. Sargent's point that agent's decisions are a function of expectations about the future is clearly evident from (12). Evaluating (12) at the minimal and maximal optimal transition functions for capital it can be shown that the support of the limiting distribution under irreversible investment is a subset of the support when investment is reversible.

3.3 Stochastic Growth with Experimentation and Learning

The stochastic growth model has been extended to environments where there is learning about productivity or the capital stock itself. This requires expanding the state space to represent the agent's beliefs. The transition equation for beliefs follows Bayes' rule. In this setting, the possibility of learning affects the optimization problem in two important ways. First, even if information signals are exogenous so that learning is passive and not affected by the current action, the mere prospect of learning may alter current period decisions. Second, when the current action affects how much learning occurs, there is an incentive to experiment to obtain better information. Friexas [40] was the first to examine this problem. Assume output is produced by a technology $f(k, \theta, r)$, where θ is an unknown parameter. The distribution of r is known. Given an initial value for y and current beliefs about θ , the agent chooses consumption and investment. Output in the following period is observed and provides information that can be used to update beliefs about θ . Friexas examines how learning and experimentation affect the initial consumption/investment decision. The learning effect depends on whether learning increases or decreases the marginal value of investment. Friexas then uses Blackwell's [10] theorem to assert that if larger investment yields more information then the experimentation effect leads to an increase in investment. Subsequently, it was shown in [5] and [29] that

this need not always be true. The reason is that investment affects both state variables in the value function so that Blackwell's theorem does not apply. While higher investment may be more informative, the value of information at higher levels of output may be lower. When the second effect dominates, an expected utility maximizer may prefer to invest less even if it is more informative. These tradeoffs have made it difficult to obtain a general set of verifiable conditions to characterize how information affects consumption and investment in the infinite horizon model. Precise results are limited largely to problems where there are only two relevant decision periods.

Nyarko and Olson [97] examine experimentation and learning in a stochastic growth model where there is imperfect information about the capital stock itself. Consumption is observable, but output and investment are not. Beliefs about the state are summarized by a probability distribution over y . After choosing consumption, an information signal is observed that can be used to update beliefs about y . The mapping from beliefs in period t to beliefs in period $t + 1$ is determined jointly by consumption, the information signal and the stochastic production function. Here there is learning about a moving target, in contrast to the case above where the unknown parameter is fixed. Nyarko and Olson show that if $u(0) = -\infty$ ²⁵ then the optimal policy is to assume the worst and optimize against that. That is, the initial state is assumed to be the lower bound of the support of the agent's beliefs about output and the transition equation is $\inf_r f(k, r)$. When information alters the lower bound of the support of the agent's beliefs there is an endogenous capital discovery process. When it does not, the problem with learning has an equivalent, deterministic representation. In that case, output and investment are more volatile than consumption and there is excess saving compared to the case where the capital stock is observable. In cases where $u(0) > -\infty$, the solution either corresponds to that above, or the capital stock becomes zero with strictly positive probability.

4 Non-classical Models of Optimal Stochastic Growth

The models of optimal economic growth under uncertainty reviewed in the previous section are based on the classical assumptions of a convex technology and utility that depends only on consumption. This section reviews some extensions of the theory that allow for non-classical features such as non-convexities and state-dependent utility. These non-classical features imply that even in a one-sector model, continuity and monotonicity properties of optimal policies need not hold and optimal paths need not converge to a unique stochastic steady state. The long run behavior of the economy may depend critically on the initial state.

²⁵This assumption holds for the class of all constant relative risk averse utility functions with coefficient at least one.

4.1 Stochastic Growth with Non-convex Technology

Non-convexities enter the production technology of an economy through numerous sources, such as fixed costs, threshold effects, increasing returns to scale, economies of scope, and depensation in the reproduction of natural resources. In applications of optimal growth models to areas such as environmental management there is also the need to study the implications of a non-convex technology. A separate chapter of this handbook focuses on optimal growth in non-convex economies. In this subsection, we concentrate on explaining how a non-concave production function (non-convex technology) alters the basic results of the classical stochastic growth literature reviewed in the previous subsections.

Majumdar, Mitra and Nyarko [68] were the first to comprehensively analyze the problem of optimal stochastic growth in a one sector model where the production function, $f(k, r)$, is not necessarily concave, though it exhibits bounded growth.²⁶ In this framework, the set of feasible programs is not necessarily convex and therefore, the value function for the dynamic optimization problem is not necessarily concave even though the utility function satisfies classical concavity restrictions. This non-convexity means that the maximization problem on the right hand side of the functional equation may have multiple solutions so that instead of a unique optimal policy function, the solution is characterized by a measurable selection from an upper semi-continuous optimal policy correspondence. Further, there need not exist any continuous selection and every policy function may exhibit jump discontinuities on a set that is at most countable. Also, non-convexity in the economy implies that the optimal path is not necessarily decentralizable - in particular, support prices may not exist.

As the value function is not necessarily concave, the expected future marginal value of capital may be increasing in current investment.²⁷ This, in turn, implies that optimal consumption may actually decline with an increase in output.²⁸ Indeed, in the deterministic model it has been shown that there may not exist an optimal consumption function that is globally monotonic. The optimal investment policy correspondence is, however, an ascending correspondence. Further, if the utility function is strictly concave, then it can be shown that every measurable selection from this correspondence is non-decreasing and an optimal investment policy function $K(y)$ is always non-decreasing in output.²⁹

²⁶Some notable contributions to deterministic optimal growth with a non-convex technology include [67],[32].

²⁷The term "marginal" is used loosely here as the value function is not necessarily differentiable no matter how smooth the utility and production functions are.

²⁸Unlike both the classical stochastic model and the deterministic model with non-concave production function, it is difficult to guarantee that optimal consumption is strictly positive in the stochastic model with non-concave production, even if Inada conditions are imposed on the utility and production functions. An interior optimal policy is ensured in [68] by assuming that $u(0) = -\infty$, which is a very strong restriction on the class of admissible utility functions. More recently, [94] establish interiority by assuming the Inada condition on utility, sufficiently high marginal productivity at zero, and that the random shock is multiplicative and has a density function so that the maximand on the right hand side of the functional equation of dynamic programming is smooth.

²⁹If the utility function is concave but not strictly concave, then there may be an optimal

A central question is whether there exists a globally stable invariant distribution. In the deterministic literature with non-concave production functions, it has been shown that there may be a multiplicity of steady states and the limit of the optimal path of capital stocks may depend on the initial state. For example, with an S-shaped production function, it is quite possible that optimal paths from small stocks converge to zero (extinction), while for initial stocks above a critical level,³⁰ optimal paths converge to a strictly positive optimal steady state. This initial state dependence can be expected to be true in the stochastic model too.

Consider the model of Section 2 without assumption A.4. For any measurable selection from the optimal policy correspondence, the transition function $H(k, r)$ for the Markov process of optimal capital stocks (8) is non-decreasing in k , but not necessarily continuous.³¹ Recall that k_m, k_M are the largest positive fixed point of the lowest transition function $H_m(k)$ and the smallest positive fixed point of the highest transition function $H_M(k)$, respectively. A critical step in the proof of global stability in Proposition 8 is Lemma 7 that showed $k_m < k_M$. Indeed, if A.10 and A.11 hold and $k_m < k_M$, then even if assumption A.4 does not hold, the proof of Proposition 8 goes through and there exists a globally stable invariant distribution. However, in the non-convex model it is quite possible that $k_m > k_M$ so that Lemma 7 does not hold. To see what happens in that case, suppose that A.10 and A.11 hold and optimal policy is interior ($0 < K(y) < y$ for all $y > 0$). Once again, confine attention to the case where $f(k, r) = rf(k)$ where r assumes one of two possible values a, b . As before, let $y_m > 0$ be the smallest positive fixed point of the lowest transition function $H_m(k)$. Then, for all $k \in (0, y_m)$, $H_M(k) > H_m(k) > k$ and $H_M(y_m) > H_m(y_m) = y_m$ so that $y_m < k_M$. Similarly, it is easy to show that $k_m < y_M$, where y_M is the largest fixed point for the highest transition function $H_M(k)$. Thus, $k_m > k_M$ implies $y_m < k_M < k_m < y_M$. This configuration is depicted in Figure 2. It is easy to check that the two disjoint intervals $[y_m, k_M]$ and $[k_m, y_M]$ are both ν -invariant; from any initial state in either interval, the optimal capital process remains in that interval almost surely. For $k_0 \in (0, k_M]$, all optimal paths eventually enter and stay in the interval $[y_m, k_M]$ while for $k_0 \in [k_m, \infty)$, all optimal paths eventually enter and stay in the interval $[k_m, y_M]$. There is no globally stable invariant distribution. Using arguments based on the splitting condition referred to earlier, Majumdar, Mitra and Nyarko [68] show that if $k_m > k_M$, then for all $k_0 \in (0, k_M]$, the distribution of capital stocks converges to the same invariant distribution whose support is $[y_m, k_M]$, while for all $k_0 \in [k_m, \infty)$, the distribution of capital stocks converges to another invariant distribution

investment function that is non-monotonic though, even in that case, there is at least one optimal investment function that is non-decreasing.

³⁰This critical level is referred to as a safe standard of conservation in the literature on renewable resource economics.

³¹An innovative approach to the non-convex model can be found in Amir [1]. It takes advantage of the averaging associated with the random disturbances to derive conditions for the monotonicity of optimal policies and higher order differentiability of the value function. As in [12], differentiability of the optimal policy functions follows from the implicit function theorem.

whose support is $[k_m, y_M]$. For any fixed initial stock in the intermediate range (k_M, k_m) , the optimal path may enter either of the two invariant sets and remain there, depending on the realization of random shocks. This last possibility illustrates an aspect of path dependence that has no parallel in the deterministic literature.

In general, non-convexities in production may lead to multiple invariant distributions. However, if production is "sufficiently stochastic", then there exists a globally stable invariant distribution despite the non-convexity [68]. Here, the precise condition that ensures global stability is

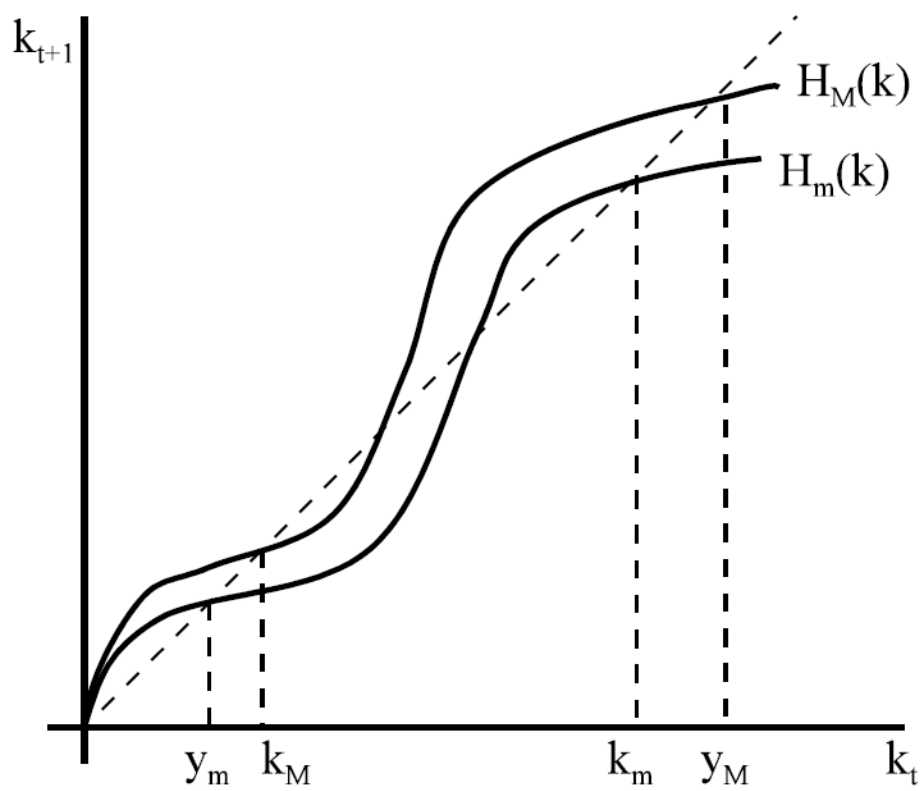
A.12. There exists some $\vartheta > 0$ in S such that $\nu(\{r \in \Phi \mid f(k, r) \leq \vartheta \text{ for each } k \in S\}) > 0$ and $\nu(\{r \in \Phi \mid f(k, r) \geq \vartheta \text{ for each } k \in S\}) > 0$. This is a condition on the production function, not the transition function for the optimal capital process. It captures the idea that the output that results from any given investment is sufficiently spread out, i.e., the technology exhibits sufficient variability. Under this condition, if we let $z = K(\vartheta)$, then one can easily verify that the splitting condition described in the proof of Proposition 8 is immediately satisfied. This ensures global stability. Hence, the possibility of multiple stochastic steady states depends on the stochasticity of the model. This is another instance where the stochastic growth model (with sufficient uncertainty) is qualitatively different from the deterministic analogue. We summarize the above discussion in the next proposition:

Proposition 10 *Assume A.1 – A.7, A.10, A.11 and that optimal policy is interior. Then, (i) if $k_m < k_M$, then there is a unique invariant distribution on S whose support is $[k_m, k_M]$ and from every $k_0 > 0$, the optimal capital stocks converge in distribution (uniformly) to this invariant distribution; (ii) if $k_m \geq k_M$, then for all $k_0 \in (0, k_M]$, the distribution of optimal capital stocks converges to an invariant distribution whose support is $[y_m, k_M]$, while for all $k_0 \in [k_m, \infty)$, the distribution of capital stocks converges to another invariant distribution whose support is $[k_m, y_M]$. If, further, A.12 holds, then the conclusion in (i) always holds.*

As in Section 3.1, A.11 implicitly imposes restrictions on the technology. For example, in [68] it is obtained from the model primitives by assuming (in addition to a condition for interiority of optimal policy) that the random shock has finite support and that the marginal productivity at zero is infinite. The latter is a rather serious restriction on the class of admissible non-concave production functions. It rules out the S-shaped production function that is a widely used canonical form to capture increasing returns to scale and other threshold effects.³² In a recent paper, Nishimura, Rudnicki and Stachurski [94] analyze a non-convex model with multiplicative i.i.d. random shock that has a density function that is strictly positive on $\mathbb{R}_{++}$. Under restrictions on the expectation of the random shock, they show that the Markov process of optimal capital stocks either converges to zero from every initial state or there is a globally

³²Mitra and Roy [92] provide weaker conditions that can ensure A.11 even when the marginal productivity at zero is finite and the distribution of the random shock is absolutely continuous.

Figure 2: Figure 2



stable non-zero steady state. To place their results in context, their assumption on the density function automatically satisfies the "very stochastic" assumption in [68] discussed above. Their result does not require Inada conditions on the production function and, in fact, allows the marginal product at zero to be less than one with positive probability.

The literature on non-convex stochastic growth also develops turnpike conditions under which optimal paths approach each other asymptotically. In a model with non-convex and non-stationary technology Joshi [50] uses monotonicity properties of the optimal policy and a supermartingale process generated by the stochastic Ramsey-Euler equation to show that, under a strong "value loss" condition that is uniform with respect to time and state, the asymptotic distance between optimal paths from two distinct initial states converges to zero with probability one. However, as in the case of turnpike theorems in the stochastic multisector convex models, the uniform value loss condition is not very transparent in terms of its implications for the model primitives.

One of the interesting questions in stochastic growth models with non-convexity is the possibility of extinction where optimal paths converge to zero. This is particularly important in applications of the optimal growth model to problems of renewable resource management where utility reflects the net benefit from harvesting and the production function reflects natural biological growth. Assuming a bounded growth production function and i.i.d. shocks that have compact support, Mitra and Roy [92] show that there are only three possibilities: (i) optimal paths from all initial states get arbitrarily close to zero infinitely often with probability one (this includes extinction in finite time), (ii) optimal paths from all initial states are bounded away from zero with probability one, and (iii) there exists a critical capital stock or safe standard above which all optimal paths are bounded away from zero with probability one. They develop sufficient conditions on the preferences and technology that leads to each of the above outcomes. In contrast to the deterministic literature, these conditions involve not just the discount factor and marginal productivity, but also marginal utility - one compares the discount rate to expected "welfare-modified" return on investment (marginal productivity) as in the condition in Proposition 11. Another result on optimal extinction is due to Kamihigashi [53] who shows that if the marginal productivity at zero is finite, then sufficient variability in the random shock implies that *all* feasible programs (including therefore, the optimal program) converge to zero almost surely.

4.2 Stochastic Growth with Stock-Dependent Utility

For some important capital theoretic allocation problems welfare depends on both consumption and the beginning of period output, as represented by $u(c, y)$.³³ Utility is assumed to be nondecreasing in y , jointly concave in (c, y) and A.7 is

³³Such models include the allocation of natural capital or renewable resources and the effects of wealth on consumption-savings behavior.

no longer imposed.³⁴ In the deterministic case stock-dependent utility has two important consequences. The first consequence arises if investment and output are substitutes in utility in the sense that $u(y - k, y)$ is submodular in k and y . In that case, an interior optimal investment policy may be decreasing in output. At the same time, there may be intervals of the state space on which corner solutions are optimal and the optimal transition function coincides with the production function. Combining these two possibilities opens the door for the optimal transition function to be like a tent map, or even more complex. When this happens an optimal program may exhibit nonlinear dynamics including cycles or chaos [3]. The second important consequence is that multiple optimal steady states are possible, even if the utility function is supermodular in k and y and the optimal investment policy is monotone (Kurz [58]). In such cases, the asymptotic behavior of an optimal program depends on the initial state.

The first analysis of stochastic models with stock-dependent utility can be traced to the literature on renewable resource allocation. In that literature, the production function represents biological growth of the renewable resource and the random shock represents the effect of environmental disturbances on resource growth. The state variable is the resource stock (output) at the beginning of the period. Stock-dependent utility arises when the harvest costs depend on the resource stock or when the resource stock has amenity or other social value. Early papers ([45], [105], [121]) focused on the case where $u_{cc}(c, y) + u_{cy}(c, y) = 0$. In this case the direct and indirect utility effects of an increase in output offset exactly and investment and output are neither strict complements nor strict substitutes in utility. An interior optimal investment policy is simultaneously nondecreasing and nonincreasing in output. That is, the optimal investment policy is a constant investment policy, which in the presence of fixed costs becomes an (s,S) inventory rule. Mendelsohn and Sobel [77] prove monotonicity of the optimal investment policy under the supermodularity condition $u_{cc}(c, y) + u_{cy}(c, y) \leq 0$. Nyarko and Olson [95] show that the optimal consumption policy is nondecreasing when $u_{cy}(c, y) \geq 0$ is imposed in addition to the concavity of u and f . They also use the Dubins and Freedman splitting condition to characterize the convergence of optimal programs to a limiting distribution. Without additional restrictions the invariant distribution may not be unique and the long run behavior of an optimal program may depend on initial conditions.³⁵ Subsequently, Nyarko and Olson [96] show that additional sufficient conditions for the existence of a unique invariant distribution are: (i) $u_c(c, y) = 0$ implies $u_y > 0$ for sufficiently large y , and (ii) for all $y > 0, c \in \Gamma(y)$ and $\alpha > 1$, if $u_c(c, y) > 0$ and $u_c(\alpha c, \alpha y) > 0$ then $\frac{u_y(c, y)}{u_c(c, y)} \geq \frac{u_y(\alpha c, \alpha y)}{u_c(\alpha c, \alpha y)}$. The last assumption is a complementarity condition that implies that the slope of indifference curves for u decrease as output and consumption increase along a

³⁴In renewable resource allocation problems welfare declines if consumption exceeds the quantity that equates demand and supply.

³⁵One, but not the only, mechanism by which this can happen is for one invariant distribution to exist where state is large enough that its shadow price is zero, $V_y = u_c(C(y)) = 0$, while another invariant distribution exists at smaller states with a strictly positive shadow price.

ray through the origin in (c, y) space. Nyarko and Olson provide examples to show that multiple invariant distributions can be optimal when either (i) or (ii) are violated. The existence of a unique invariant distribution is also ensured when there is sufficient divergence between production in the best and worst states. [68] and [96] show that there is more than one way to define sufficient variation in production. The underlying intuition is the same. A model with multiple limiting distributions can be transformed into one with a unique invariant distribution by the mixing that results from increasing the variance in production. On the other hand, if the variability in production is small enough and if $u(y - k, y)$ is submodular in (k, y) , then an optimal program may oscillate between cyclic sets [3].

The economic possibilities associated with the stochastic growth model expand considerably when a non-convex production technology is combined with stock-dependent utility. To date this combination has primarily been used to examine the conditions under which capital stocks remain strictly bounded away from zero, issues related to conservation and extinction. In the deterministic model with both non-convex production and stock-dependent utility it is possible for there to be disjoint intervals in the state space from which an optimal program converges to zero. That is, an optimal program starting from intermediate states may remain bounded away from zero, while optimal programs starting from lower or higher states converge to zero [99]. The addition of random productivity disturbances leads to the somewhat surprising possibility that a first-order improvement in the distribution of disturbances can reduce the set of initial states from which optimal output and capital stocks have a positive lower bound.

One useful technique to analyze some questions in the non-convex model is to examine behavior under the convex-hull of the technology. If capital stocks under an optimal program always remain in an interval for which the convex-hull coincides with the non-convex technology then the two optimization problems coincide on that interval. This can be used, for example, to provide conditions for the existence of a safe standard of conservation.

Proposition 11 *Assume $u(y - k, y)$ is supermodular in (k, y) , u is increasing in c , and f is concave in k for all r . Let $\underline{f}(k) = \inf_r f(k, r)$. If*

$$\inf_{z \in [0, k]} \delta E \left[\frac{u_c(f(k, r) - z, f(k, r)) + u_y(f(k, r) - z, f(k, r))}{u_c(\underline{f}(k) - z, \underline{f}(k))} \right] > 1 \quad (13)$$

then $\liminf y_t \geq k$ for all $y_0 \geq k$.

A general version of this result in the model with non-convex technology and stock-dependent utility can be found in Olson and Roy [100], along with other results dealing with conservation or extinction. The conclusions depend on the joint properties of the technology, utility, and the distribution of disturbances. As can be seen above, $\underline{f}(k)$ or productivity under the worst disturbance is an important determinant of conservation or extinction.

5 Comparative Dynamics

An important question in stochastic growth theory is the sensitivity of optimal decision rules and paths with respect to preference and technology parameters that describe the underlying economy. In a one sector model, continuity of optimal investment and consumption decision with respect to various parameters of the model generally holds under far weaker assumptions than those described in Section 2.³⁶

The theory of monotone comparative statics using supermodular functions and complementarity developed in Topkis [127] has been extended to stochastic dynamic models (see, for example, [119], [42]). One can apply results from this literature to derive the comparative dynamics of the optimal policy function with respect to various preference and technology parameters by looking at the maximization problem on the right hand side of the functional equation of dynamic programming [43]. Most of these results have been derived in a one sector framework.

Danthine and Donaldson [26] show that an increase in the discount factor increases optimal investment and shifts the distribution of optimal capital stocks to the right and hence, the invariant distribution to which the stocks converge.³⁷ Moreover, they show that an increase in the curvature of the utility function (loosely speaking, an increase in risk aversion), leads to higher consumption (i.e., lower investment) at low levels of output, and lower consumption (i.e., higher investment) at high levels of output; further, the range of the limiting distribution expands as risk aversion increases.³⁸

Another important issue in comparative dynamics is the effect of a change in the degree of riskiness or volatility of the random shocks. This relates to a central concern in macroeconomics about the relationship between riskiness of productive assets and the optimal intertemporal precautionary saving decisions of individuals as well as more aggregative analysis of the relationship between growth and economic fluctuations (see for example, [48]). Unfortunately, there is no general characterization of the effect of a second order stochastic dominance change in the distribution of shocks on the optimal policy.³⁹

In the specific case of optimal savings under uncertainty⁴⁰ discussed in Example 9, one can characterize the comparative dynamics of riskiness fairly tightly. From (10), we have

³⁶See for example, [38]. Conditions for parametric continuity of stationary distributions of Markov processes are discussed, among others, by [124] and [60]. These properties are important for numerical simulations.

³⁷Dutta [36] shows that lengthening the time horizon for a fixed discount factor and increasing the discount factor for a fixed time horizon are, in a precise sense, equivalent.

³⁸In the case of logarithmic utility, Cobb-Douglas production with multiplicative shock, an increase in the discount factor increases the variance of capital stock and output. Danthine and Donaldson [27] provide sufficient conditions for this to occur. They also characterize conditions under which an increase in the curvature of the utility function has a similar effect.

³⁹An exception is the model of optimal dynamic consumption with deterministic linear interest and additive labor income shock. See, for example, [79].

⁴⁰See, among others, [101], [62], [109], [61].

$$K(y) = \alpha y, E\left[\frac{k_{t+1}}{k_t}\right] = \alpha E(r_{t+1}), \text{ where } \alpha = [\delta E(r^{1-\sigma})]^{\frac{1}{\sigma}}$$

so that the propensity to invest/save and the expected growth rate of capital are both proportional to α and the latter is increasing (decreasing) in riskiness of the random shock if $\sigma > (<)1$ because $r^{1-\sigma}$ is a convex (concave) function of r in that case. Thus, depending on the curvature of the utility function, an increase riskiness may increase or decrease optimal investment and cause a first order increase or decrease in the distribution of optimal capital stocks. Roughly speaking, if utility is more concave than the logarithmic function, an increase in riskiness of the random shock increases the optimal savings rate and the expected rate of growth. The reverse holds if utility is less concave than the logarithmic function. In the case of log utility, the optimal policy depends only on the average realization of the random shock and not on its higher moments.⁴¹

A less ambitious question relates to a comparison of the moments of the limiting distribution to the steady state in the deterministic model.

Example 12 $u(c) = \ln c$ and $f(k, r) = rk^\beta, 0 < \beta < 1$. For this example Mirman and Zilcha [84] show that the optimal investment policy is given by $K(y) = \beta\delta y$. Danthine and Donaldson [27] use this to analyze the properties of the optimal program for capital:

$$k_t = (\beta\delta)^{1+\beta+\dots+\beta^{t-1}} k_0^{\beta^t} r_0^{\beta^t} r_1^{\beta^{t-1}} \dots r_{t-1}^\beta r_t.$$

This implies:

$$Ek_t = (\beta\delta)^{\sum_{s=0}^{t-1} \beta^s} k_0^{\beta^t} \prod_{s=0}^{t-1} E(r_{t-s-1}^{\beta^s}).$$

Assume $E(r_t) = 1$ and a non-degenerate distribution for r_t . Then taking limits as $t \rightarrow \infty$, the first moment of the limiting invariant distribution of capital satisfies: $Ek = (\beta\delta)^{\frac{1}{1-\beta}} L$, where, $L = \lim_{t \rightarrow \infty} \prod_{s=0}^{t-1} E(r_{t-s-1}^{\beta^s})$. Jensen's inequality implies $L < 1$. Expected consumption and output in the limiting invariant distribution are given by: $Ec = (1 - \beta\delta)(\beta\delta)^{\frac{\beta}{1-\beta}} L$, $Ey = (\beta\delta)^{\frac{\beta}{1-\beta}} L$. In the deterministic version of the model (where $r_t = 1$ almost surely), steady state capital, consumption and output are given by: $k = (\beta\delta)^{\frac{1}{1-\beta}}$, $c = (1 - \beta\delta)(\beta\delta)^{\frac{\beta}{1-\beta}}$, $y = (\alpha\delta)^{\frac{\beta}{1-\beta}}$. This shows that in the stochastic model, the steady state distribution has smaller average capital stock, output and consumption than in the certainty equivalent version of the model.

In Example 12, uncertainty only affects the evolution of an optimal program and not the optimal policy function itself. This simplifies the task of characterizing the effect of uncertainty on the limiting distribution. In general, uncertainty will also affect the optimal policy function. As we have seen earlier, in the case

⁴¹For an extension of this kind of result to a model of endogenous growth see [30].

of iso-elastic utility and linear production, uncertainty may increase or reduce optimal investment depending on the nature of the utility function. This makes it difficult to compare the moments of the limiting distribution of capital for the stochastic model with its certainty equivalent.

6 Solving the Stochastic Growth Model

The stochastic growth model is inherently nonlinear. There is no known general, closed form solution. Instead, analysis of the model with general functional forms aims to qualitatively characterize optimal policies and the resulting implications for economic behavior. There are two main approaches to achieving more specific solutions, all of which require assumptions regarding functional forms for production and utility. The only cases with known closed form analytical solutions are those discussed in Examples 9 and 12 the linear-quadratic case.

Approximation and numerical methods are the alternative when an analytical solution is not available.⁴² By far the most common approximation technique is to linearize the Euler equations around the steady state of the model, an idea pioneered by Magill [64] in continuous time and Kydland and Prescott [59] in discrete time. This approach was subsequently extended by [56], [57], [20] and many others. In a model with Cobb-Douglas technology and CES/CRRA utility, [129] develops central limit and large deviation principles that characterize the manner in which capital trajectories in the stochastic model converge to those in the deterministic case as the standard deviation of the random shock goes to zero. In practice, most approximation methods are not entirely analytical and the approximate solution is analyzed using simulations where the underlying parameters are calibrated to data. Solutions are accurate in the neighborhood of a stochastic steady state with support on a small interior interval. Approximation methods are less useful in situations where the disturbance term has support on a large interval, where the solution is not interior, where second order effects are important, and in the study of transition dynamics.

Numerical dynamic programming can be also be used to solve the parametric specifications of the stochastic growth model. The two most common approaches involve iteration of discrete or parametric approximations to the value or policy functions. Recent surveys of numerical methods can be found in [110], [52], [73], and [111]. Once the model is solved, the policy functions can be used to compute moments for the limiting distributions of the economic variables of interest. The main advantage of numerical dynamic programming is that attention need not be restricted to a neighborhood of the steady state. This allows one to investigate almost any question of interest within the context of a given parametric specification, including a study of global dynamics. The primary disadvantage has to do with robustness to model specification, calibration and choice of numerical method.⁴³ In general, different numerical

⁴² An early survey and comparison of different methods can be found in [128].

⁴³ For algorithms generated by a contraction mapping of modulus δ , the approximation

procedures can yield substantially different results so care must be exercised in their implementation.

7 Conclusion

The literature on optimal stochastic growth theory is over three decades old now and there are many important ways in which the theory has contributed to our understanding of capital accumulation, growth and more generally, optimal intertemporal resource allocation. In this section, we summarize some of the contributions of the stochastic growth literature and we point out where the introduction of uncertainty has done little to alter the conclusions of the deterministic model.

First, stochastic growth theory has provided a different explanation of economic volatility. In contrast to the deterministic case, an optimal program in the stochastic model is a sequence of random variables generated jointly by optimal decisions and random productivity disturbances. Realized capital paths fluctuate even when the optimal policy is time stationary and well-behaved. This way of looking at economic volatility has been successfully utilized by the business cycle literature as a way to capture various stylized facts about economic fluctuations.

Second, in the stochastic growth model the utility function plays a prominent role in determining the long run behavior of the economy. Indeed, even in a one-sector model and for the same production technology, discount factor and distribution of random shock, the limiting steady state distribution can differ wildly according to the specification of the utility function. The role of the utility function is also seen in conditions for long run growth and avoidance of extinction. This role is absent in deterministic models.

Because the utility function affects long run behavior in the stochastic growth model it is possible to examine how optimal paths and the limiting distribution are affected by changes in the riskiness of productive assets, risk aversion, or the willingness to substitute consumption across time. Unfortunately, not much general analytical characterization is available outside a few examples in the log-linear family. These examples nonetheless serve to illustrate how the qualitative nature of comparative dynamics can depend on the parameters of the utility function.

Third, key qualitative features of optimal policies such as continuity and monotonicity are not significantly altered by the presence of uncertainty in the production technology.

Fourth, extending results on the existence and global stability of an optimal steady state to the stochastic model requires verifying that the transition law

error is bounded by $\|V_n - V_{n+1}\| < \epsilon/(1 - \beta)$, where ϵ is the tolerance level under the given metric and V_n is the n^{th} iterate of the algorithm. Santos [112] shows how the Euler equation residuals can be used to bound the approximation error for other types of algorithms. Santos and Peralta-Alva [113] examine when the simulated moments from a numerical solution to converge to their exact values as the approximation errors converge to zero.

for the optimal process satisfies certain conditions, which are discussed above. This has necessitated strong technical assumptions that have no counterpart in the deterministic literature. In the multisector case, this difficulty has been more pronounced and the conditions for global stability of a stochastic steady state are only specified in terms of the transition law for the optimal process, making it difficult to evaluate their economic implications.

Fifth, in non-classical one-sector models that generate multiple invariant distributions that act as local attractors, it has been shown that if the volatility of technological disturbances is increased sufficiently, one can establish global convergence of optimal processes to a unique stochastic steady state. Loosely speaking, higher stochasticity in the production technology makes it more likely that realized optimal paths exhibit a high degree of economic fluctuations over time, but it also increases the likelihood that the distribution of optimal capital stocks converges globally to a unique invariant distribution independent of the initial state. In other words, greater production uncertainty may be associated with higher economic volatility and at the same time, may ensure long run "convergence" in probability distribution of economies that differ in their initial states. This is a fundamental insight into the process of growth and fluctuations in an economy.

Finally, the stochastic growth literature has followed the deterministic literature very closely in establishing a set of turnpike results that show how optimal paths approach each other almost surely in the long run.

As for the important theoretical questions that remain unanswered, our brief survey indicates that a general characterization of the stochastic steady state, or invariant distribution, is lacking. Steps toward such a characterization would improve our understanding of the forces that determine long run economic behavior in a convergent stochastic economy. We not only need to understand how complex the limiting distribution can be, but also have some idea of the relationship between the fundamentals of the model and the properties of the limiting distribution. That is, what do technology and preferences imply about the nature of the limiting distribution? Much work remains to be done there.

Other important open questions in the one sector model are: a *complete* characterization of conditions under which optimal paths converge to zero almost surely, to a non-trivial invariant distribution and diverge to infinity almost surely (the existing literature only provides strong sufficient conditions for each of these events); relaxing the conditions for convergence and stability in the non-convex model; and the question of asymptotic convergence in versions of the model with non-monotone optimal investment policy (such as the stock-dependent model). Developing more transparent conditions for convergence and stability in the multisector stochastic model and conditions for sustained long run growth in such models are also problems that remain open to the current generation of growth theorists.

Finally, the methodology of stochastic optimal growth is increasingly applied to other problems of dynamic resource allocation ranging from models of financial markets and macroeconomic fluctuations to the management of natural and environmental assets. These applications often require extensions and modifica-

tions to the basic framework in order to suit the stylized facts that characterize the fundamentals. This, in turn, poses new questions for the growth theorist. The development of new applications and extensions of existing ones may well continue to be the most fruitful source of new ideas related to the stochastic growth model.

References

- [1] Amir, R., 1997, "A new look at optimal growth under uncertainty", *Journal of Economic Dynamics and Control* 22(1), 67-86.
- [2] Athreya, K.B. , 2003, "Stationary measures for some Markov chain models in ecology and economics", *Economic Theory* 23(1), 107-22.
- [3] Benhabib, J. and K. Nishimura, 1989, Stochastic equilibrium oscillations, *International Economic Review* 30(1), 85-102.
- [4] Benveniste, L.M. and J.A. Scheinkman, 1979, On the differentiability of the value function in dynamic models of economics, *Econometrica* 47(3), 727-32.
- [5] Bertocchi, G. and M. Spagat, 1998, "Growth under uncertainty with experimentation", *Journal of Economic Dynamics and Control* 23(2), 209-31.
- [6] Bhattacharya, R.N., and M. Majumdar, 1981, "Stochastic models in mathematical economics" in *Proceedings of the Indian Statistical Institute Golden Jubilee International Conference on "Statistics: Applications and New Directions"*, Indian Statistical Institute, Calcutta.
- [7] Bhattacharya, R.N. and M. Majumdar, 1999, "On a theorem of Dubins and Freedman", *Journal of Theoretical Probability* 12(4), 1067-87.
- [8] Bhattacharya, R.N. and M. Majumdar, 2001, "On a class of random dynamical systems", *Journal of Economic Theory* 96(1/2), 208-29.
- [9] Bhattacharya, R.N. and M. Majumdar, 2003, "Random dynamical systems: a review", *Economic Theory* 23(1), 13-38.
- [10] Blackwell, D. 1953. Equivalent comparison of experiments, *Annals of Mathematical Statistics* 24, 265-72.
- [11] Blackwell, D., 1965, "Discounted dynamic programming", *Annals of Mathematical Statistics* 36, 226-35.
- [12] Blume, L., Easley, D., and M. O'Hara, 1982, "Characterization of optimal plans for stochastic dynamic programs", *Journal of Economic Theory* 28(2), 221-34.

- [13] Boylan, E.S., 1976, "On properties of steady state measures for one-sector growth models", *International Economic Review* 17(3), 783-85.
- [14] Brock, W.A., 1979, "An integration of stochastic growth theory and the theory of finance, Part I: The growth model", in *General Equilibrium, Growth and Trade*, (eds., J.R. Green and J.A.Scheinkman), Academic Press, New York.
- [15] Brock, W.A., 1982, "Asset prices in an uncertain economy", in *The Economics of Information and Uncertainty* (ed., J.J. McCall), University of Chicago Press, Chicago.
- [16] Brock, W.A. and M.J.P. Magill, 1979, "Dynamics under uncertainty", *Econometrica* 47(4), 843-68.
- [17] Brock, W.A. and M. Majumdar, 1978, "Global asymptotic stability results for multisector models of optimal growth under uncertainty where future utilities are discounted", *Journal of Economic Theory* 18(2), 225-43.
- [18] Brock, W.A. and L.J. Mirman, 1972, "Optimal economic growth and uncertainty: the discounted case", *Journal of Economic Theory* 4(3), 479-513.
- [19] Brock, W.A. and L.J. Mirman, 1973, "Optimal economic growth and uncertainty: The no discounting case", *International Economic Review* 14(3), 560-73.
- [20] Campbell, J. Y., 1994, "Inspecting the mechanism: an analytical approach to the stochastic growth model", *Journal of Monetary Economics* 33(3), 463-506.
- [21] Cass, D., 1965, "Optimal growth in an aggregative model of capital accumulation", *Review of Economic Studies* 32, 233-40.
- [22] Cass, D. and K. Shell, 1976, "The structure and stability of competitive dynamical systems", *Journal of Economic Theory* 12(1), 31-70.
- [23] Chamberlain, G. and C.A. Wilson, 2000, "Optimal intertemporal consumption under uncertainty", *Review of Economic Dynamics* 3(3), 365-95.
- [24] Chang, F.-R., 1982, "A note on the stochastic value loss assumption: global asymptotic stability results for multisector models of optimal growth under uncertainty when future utilities are discounted", *Journal of Economic Theory* 26(1), 164-70.
- [25] Dana, R.A., 1974, "Evaluation of development programs in a stationary stochastic economy with bounded primary resources", *Proceedings of the Warsaw Symposium on Mathematical Models in Economics*.

- [26] Danthine, J.-P., and J.B. Donaldson, 1981a, "Stochastic properties of fast vs. slow growing economies", *Econometrica* 49(4), 1007-33.
- [27] Danthine, J.-P., and J.B. Donaldson, 1981b, "Certainty planning in an uncertain world: a reconsideration", *Review of Economic Studies* 48(3), 507-10.
- [28] Datta, M., 1999, "Optimal accumulation in a small open economy with technological uncertainty", *Economic Theory* 13(1), 207-19.
- [29] Datta, M., L.J. Mirman and E.E. Schlee, 2002, "Optimal experimentation in signal dependent decision problems", *International Economic Review* 43(2), 577-608.
- [30] de Hek, P., 1999, " On endogenous growth under uncertainty", *International Economic Review* 40(3), 727-44.
- [31] de Hek, P. and S. Roy, 2001, "On sustained growth under uncertainty", *International Economic Review*, 42(3), 801-14.
- [32] Dechert, W.D. and K. Nishimura, 1983, "A complete characterization of optimal growth paths in an aggregated model with a nonconcave production function", *Journal of Economic Theory* 31(2), 332-54.
- [33] Diaconis, P. , and D. Freedman, 1999, "Iterated random functions", *SIAM Review* 41(1), 45-76.
- [34] Donaldson, J.B., and R. Mehra, 1983, "Stochastic growth with correlated production shocks", *Journal of Economic Theory* 29(2), 282-312.
- [35] Dubins, L.E., and D. Freedman, 1966, "Invariant probabilities of certain Markov processes", *Annals of Mathematical Statistics* 37, 837-47.
- [36] Dutta, P.K., 1987, "Capital deepening and impatience-equivalence in stochastic aggregative growth models", *Journal of Economic Dynamics and Control* 11(4), 519-30.
- [37] Dutta, P.K., 1991, "What do discounted optima converge to? A theory of discount rate asymptotics in economic models", *Journal of Economic Theory* 55(1), 64-94.
- [38] Dutta, P.K., Majumdar, M. and R. Sundaram, 1994, "Parametric continuity in dynamic programming problems", *Journal of Economic Dynamics and Control* 18(6), 1069-92.
- [39] Föllmer, H., and M. Majumdar, 1978, "On the asymptotic behavior of stochastic economic processes: Two examples from intertemporal allocation under uncertainty", *Journal of Mathematical Economics* 5(3), 275-88.
- [40] Freixas, X., 1981, "Optimal growth with experimentation", *Journal of Economic Theory* 24(2), 296-309.

- [41] Futia, C., 1982, "Invariant distributions and the limiting behavior of markovian economic models", *Econometrica* 50(2), 377-408.
- [42] Hopenhayn, H., and E.C. Prescott, 1992, "Stochastic monotonicity and stationary distributions for dynamic economies", *Econometrica* 60(6), 1387-1406.
- [43] Huggett, M., 2003, "When are comparative dynamics monotone?", *Review of Economic Dynamics* 6(1), 1-11.
- [44] Inada, K-I, 1963, "On a two-sector model of economic growth: comments and a generalization," *Review of Economic Studies* 30(2), 119-27.
- [45] Jaquette, D.L., 1972, "A discrete time population control model", *Math. Biosciences* 15, 231-252.
- [46] Jeanjean, P., 1974, "Optimal development programs under uncertainty: The undiscounted case", *Journal of Economic Theory* 7(1), 66-92.
- [47] Jones, L.E., and R.E. Manuelli, 1997, "The sources of growth", *Journal of Economic Dynamics and Control* 27(1), 75-114.
- [48] Jones, L.E., Manuelli, R., Siu, H.E. and E. Stacchetti, 2003, "Fluctuations in convex models of endogenous growth I: growth effects", Mimeo.
- [49] Joshi, S., 1995, "Recursive utility and optimal growth under uncertainty", *Journal of Mathematical Economics* 24(6), 601-17.
- [50] Joshi, S., 1997, "Turnpike theorems in nonconvex nonstationary environments", *International Economic Review* 38(1), 225-48.
- [51] Joshi, S., 2003, "The stochastic turnpike property without uniformity in convex aggregate growth models", *Journal of Economic Dynamics and Control* 27(7), 1289-1315.
- [52] Judd, K.L., 1998, *Numerical Methods in Economics*, MIT Press, Cambridge, MA.
- [53] Kamihigashi, T., 2003, "Almost sure convergence to zero in stochastic growth models", Discussion Paper No. 140, RIEB, Kobe University.
- [54] Koopmans, T.C., 1957, *Three Essays on the State of Economic Science*, McGraw-Hill, New York.
- [55] Koopmans, T., 1965, "On the concept of economic growth" in *Semaine d'Etude sur le Rôle de l'Analyse Econometrique dans la Formulation de plans de Développement*, Pontifican Academiae Scientiarum Scripta Varia No. 28, Vatican.
- [56] King, R.G., Plosser, C.I., and S.T. Rebelo, 1988, "Production, growth and business cycles I: The basic neoclassical model", *Journal of Monetary Economics* 21(2/3), 195-232.

- [57] King, R.G., C.I. Plosser and S.T. Rebelo, 2002, "Production, growth and business cycles: technical appendix", *Computational Economics* 20(1-2), 87-116.
- [58] Kurz, M., 1968, "Optimal economic growth and wealth effects", *International Economic Review* 9, 348-357.
- [59] Kydland, F. and E.C. Prescott, 1982, "Time to build and aggregate fluctuations", *Econometrica* 50(6), 1345-70.
- [60] Le Van, C. and J. Stachurski, 2004, "Parametric continuity of stationary distributions", Research Paper No. 899, Dept. of economics, The University of Melbourne, Australia.
- [61] Leland, H., 1974, "Optimal growth in a stochastic environment", *Review of Economic Studies* 41(1), 75-86.
- [62] Levhari, D. & T.N.Srinivasan, 1969, "Optimal savings under uncertainty", *Review of Economic Studies* 36, 153-63.
- [63] Long, J.B. and C.I. Plosser, 1983, "Real business cycles," *Journal of Political Economy* 91(1), 39-69.
- [64] Magill, M., 1977, "A local analysis of n-sector capital accumulation under uncertainty", *Journal of Economic Theory* 15(1), 211-18.
- [65] Maitra, A., 1968, "Discounted dynamic programming on compact metric spaces", *Sankhya* Ser. A 30, 211-16.
- [66] Majumdar, M., 1982, "A note on learning and optimal decisions with a partially observable state space" in *Essays in the Economics of Renewable Resources* (Eds., L.J. Mirman and D. Spulber), North Holland, Amsterdam.
- [67] Majumdar, M. and T. Mitra, 1982, "Intertemporal allocation with a non-convex technology: the aggregative framework", *Journal of Economic Theory* 27(1), 101-36.
- [68] Majumdar, M., Mitra, T. and Y. Nyarko, 1989, "Dynamic optimization under uncertainty: non-convex feasible sets" in *Joan Robinson and Modern Economic Theory* (ed. G.R. Feiwel), Macmillan Press, New York, 1989, 545-90.
- [69] Majumdar, M. and R. Radner, 1983, "Stationary optimal policies with discounting in a stochastic activity analysis model", *Econometrica* 51(6), 1821-37.
- [70] Majumdar, M. and I. Zilcha, 1987, "Optimal growth in a stochastic environment: some sensitivity and turnpike results", *Journal of Economic Theory* 43(1), 116-33.

- [71] Malinvaud, E., 1953, "Capital accumulation and efficient allocation of resources", *Econometrica* 21, 233-68.
- [72] Marimon, R., 1989, "Stochastic turnpike property and stationary equilibrium", *Journal of Economic Theory* 47(2), 282-306.
- [73] Marimon, R. and A. Scott, 1999, *Computational Methods for the Study of Dynamic Economies*, Oxford, Oxford University Press.
- [74] McKenzie, L.W., 1976, "Turnpike theory", *Econometrica* 44(5), 841-65.
- [75] McKenzie, L.W., 1986, "Optimal economic growth, turnpike theorems and comparative dynamics" in *Handbook of Mathematical Economics: Volume III* (Eds., K.J. Arrow and M.D. Intriligator), North Holland, Amsterdam.
- [76] McKenzie, L.W., 1998, "Turnpikes", *American Economic Review* 88(2), 1-14.
- [77] Mendelssohn, R., and M. Sobel, 1980, "Capital accumulation and the optimization of renewable resource models", *Journal of Economic Theory* 23(2), 243-60.
- [78] Merton, R.C., 1975, "An asymptotic theory of growth under uncertainty", *Review of Economic Studies* 42(3), 375-93.
- [79] Miller, B., 1976, "The effect on optimal consumption of increased uncertainty in labor income in the multiperiod case", *Journal of Economic Theory* 13(1), 154-67.
- [80] Mirman, L.J. , 1972, "On the existence of steady-state measures for one sector growth models", *International Economic Review* 13(2), 271-86.
- [81] Mirman, L.J., 1980, "One sector economic growth and uncertainty: a survey" in *Stochastic Programming* (ed., M.A.H. Dempster), Academic Press, London, 537-67.
- [82] Mirman, L.J. and D.F. Spulber, 1984, "Uncertainty and markets for renewable resources," *Journal of Economic Dynamics and Control* 8(3), 239-64.
- [83] Mirman, L.J. and Spulber, D.F. 1985. "Fishery regulation with harvest uncertainty," *International Economic Review* 26(3), 731-46.
- [84] Mirman, L.J. and I. Zilcha, 1975, "On optimal growth under uncertainty", *Journal of Economic Theory* 11(3), 329-39.
- [85] Mirman, L.J. and I. Zilcha, 1976, "Unbounded shadow prices for optimal stochastic growth models with uncertain technology", *International Economic Review* 17(1), 121-32.
- [86] Mirman, L.J. and I. Zilcha, 1977, "Characterizing optimal policies in a one-sector model of economic growth under uncertainty", *Journal of Economic Theory* 14(2), 389-401.

- [87] Mirrlees, J.A., 1974, "Optimal accumulation under uncertainty: the case of stationary returns of investment" in *Allocation under Uncertainty: Equilibrium and Optimality* (Ed., J. Drèzé), Wiley, New York.
- [88] Mitra, K., 1998, "On capital accumulation paths in a neoclassical stochastic growth model", *Economic Theory* 11, 457-64.
- [89] Mitra, T., and Y. Nyarko, 1991, "On the existence of optimal processes in nonstationary environments", *Journal of Economics* 53(3), 245-70.
- [90] Mitra, T. and F. Privileggi, forthcoming, "Cantor type invariant distributions in the theory of optimal growth under uncertainty", *Journal of Difference Equations and Applications*.
- [91] Mitra, T., Montrucchio, L., and F. Privileggi, 2003, "The nature of steady states in models of optimal growth under uncertainty", *Economic Theory* 23(1), 39-71.
- [92] Mitra, T. and S. Roy, 2003, "Optimal exploitation of renewable resources under uncertainty and the extinction of species", CAE Working Paper 03-10, Cornell University, NY.
- [93] Montrucchio, L., and F. Privileggi, 1999, "Fractal steady states in stochastic optimal control models", *Annals of Operations Research* 88, 183-97
- [94] Nishimura, K., Rudnicki, R., and J. Stachurski, 2003, "Stochastic optimal growth with non-convexities", Mimeo., Institute of Economic Research, Kyoto University.
- [95] Nyarko, Y. and L.J. Olson, 1991, "Stochastic dynamic models with stock-dependent rewards", *Journal of Economic Theory* 55(1), 161-68.
- [96] Nyarko, Y. and L.J. Olson, 1994, "Stochastic growth when utility depends on both consumption and the stock level", *Economic Theory* 4(5), 791-97.
- [97] Nyarko, Y. and L.J. Olson, 1996, "Optimal growth with unobservable resources and learning", *Journal of Economic Behavior and Organization* 29(3), 465-91.
- [98] Olson, L.J., 1989, "Stochastic growth with irreversible investment", *Journal of Economic Theory* 47(1), 101-29.
- [99] Olson, L.J. and S. Roy, 1996, "On conservation of renewable resources with stock-dependent return and nonconcave production", *Journal of Economic Theory* 70(1), 133-157.
- [100] Olson, L.J. and S. Roy, 2000, "Dynamic efficiency of conservation of renewable resources under uncertainty", *Journal of Economic Theory* 95(2), 186-214.

- [101] Phelps, E.S., 1962, "The accumulation of risky capital: a sequential utility analysis", *Econometrica* 30, 729-43.
- [102] Prescott, E.C., and R. Mehra, 1980, "Recursive competitive equilibrium: the case of homogenous households", *Econometrica* 48(6), 1365-79.
- [103] Radner, R., 1961, "Paths of economic growth that are optimal with regard only to final states", *Review of Economic Studies* 28, 98-104.
- [104] Radner, R., 1973, "Optimal stationary consumption with stochastic production and resources", *Journal of Economic Theory* 6(1), 68-90.
- [105] Reed, W.J., 1974, "A stochastic model for the economic management of a renewable resource", *Mathematical Biosciences* 22(4), 313-37.
- [106] Ramsey, F.P., 1928, "A mathematical theory of savings", *Economic Journal* 38, 543-59.
- [107] Razin, A. and J.A. Yahav, 1979, "On stochastic models of economic growth", *International Economic Review* 20(3), 599-604.
- [108] Rockafellar, R.T., 1976, "Saddle points of Hamiltonian systems in convex Lagrange problems having a nonzero discount rate", *Journal of Economic Theory* 12(1), 71-113.
- [109] Rothschild, M., and J.E Stiglitz, 1971, "Increasing risk II: its economic consequences," *Journal of Economic Theory* 3(1), 66-84.
- [110] Rust, J. 1996. "Numerical Dynamic Programming in Economics," in *Handbook of Computational Economics* (Eds, H.M. Amman, D.A. Kendrick and J. Rust) Elsevier, Amsterdam, 619-729.
- [111] Santos, M., 1999, "Numerical Solutions of Dynamic Economic Models," in *Handbook of Macroeconomics* (Eds., J.B. Taylor and M. Woodford) v. 1A, ch. 5, Elsevier, Amsterdam.
- [112] Santos, M.S., 2000, "Accuracy of numerical solutions using the Euler equation residuals", *Econometrica* 68(6), 1377-1402.
- [113] Santos, M.S., and A. Peralta-Alva, 2003, "Accuracy of simulations for stochastic dynamic models", Mimeo.
- [114] Santos, M.S., and J.T. Vigo-Aguiar, 1998, "Analysis of a numerical dynamic programming algorithm applied to economic models", *Econometrica* 66(2), 409-26.
- [115] Sargent, T.J., 1980, "Tobin's 'q' and the rate of investment in general equilibrium" in *On The State of Macroeconomics* (eds., K. Brunner and A.H. Meltzer), North-Holland, Amsterdam, 107-54.

- [116] Schäl, M., 1975, "Conditions for optimality in dynamic programming and for the limit of n-stage optimal policies to be optimal", *Z. Wahrscheinlichkeitstheorie verw. Gebiete* 32, 179-96.
- [117] Schechteman, J., 1976, "An income fluctuation problem", *Journal of Economic Theory* 12(2), 218-41.
- [118] Schechteman, J. and V. Escudero, 1977, "Some results on an 'An income fluctuations problem'", *Journal of Economic Theory* 16, 151-66.
- [119] Serfozo, R., 1976, "Monotone optimal policies for Markov decision processes", *Math. Prog. Study* 6, 202-15.
- [120] Sotomayor, M., 1984, "On income fluctuations and capital gains", *Journal of Economic Theory* 32(1), 14-35.
- [121] Spulber, D.F., 1982, "Adaptive harvesting of a renewable resource and stable equilibrium" in *Essays in the Economics of Renewable Resources* (eds. L.J. Mirman and D.F. Spulber) North Holland, Amsterdam.
- [122] Stachurski, J., 2002, "Stochastic optimal growth with unbounded shock", *Journal of Economic Theory* 106(1), 40-65.
- [123] Stachurski, J., 2003, "Stochastic growth: asymptotic distributions", *Economic Theory* 21(4), 913-19
- [124] Stokey, N.L., Lucas, R.E., and E.C. Prescott, 1989, *Recursive Methods in Economic Dynamics*, Harvard University Press, Massachusetts.
- [125] Strauch, R., 1966, "Negative dynamic programming", *Annals of Mathematical Statistics* 37, 871-90.
- [126] Sutherland, W.R.S., 1970, "On optimal development in a multi-sectoral economy: The discounted case", *Review of Economic Studies* 37, 585-89.
- [127] Topkis, D., 1978, "Minimizing a submodular function on a lattice", *Operations Research* 26(2), 305-21.
- [128] Taylor, J.B. and H. Uhlig, 1990, "Solving nonlinear stochastic growth models: a comparison of alternative solution methods", *J. Bus. Econ. Stat.* 8(1), 1-17 .
- [129] Williams, N., 2003, "Small noise asymptotics for a stochastic growth model", NBER working paper w10194, forthcoming *Journal of Economic Theory*.
- [130] Yaari, M.E., 1976, "A law of large numbers in the theory of consumer's choice under uncertainty", *Journal of Economic Theory* 12, 202-17.
- [131] Yano, M., 1989, "Comparative statics in dynamic stochastic models: differential analysis of a stochastic modified golden rule in a Banach space", *Journal of Mathematical Economics* 18(2), 169-85.

- [132] Zilcha, I., 1975, "Weakly maximal optimal stationary programs under uncertainty", *International Economic Review* 16(3), 796-99.
- [133] Zilcha, I., 1976a, "Characterization by prices of optimal programs under uncertainty", *Journal of Mathematical Economics* 3, 173-84.
- [134] Zilcha, I., 1976b, "On competitive prices in a multisector economy with stochastic production and resources", *Review of Economic Studies* 43(3), 431-38.
- [135] Zilcha, I., 1978, "Transversality conditions in a multisector economy under uncertainty", *Econometrica* 46(3), 505-26.