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Self-Similar Measures in Multi-Sector Endogenous Growth Models^{*}

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Abstract

We analyze two types of stochastic discrete time multi-sector endogenous growth models, namely a basic Lucas-Uzawa (1988) model and an extended three-sector version as in La Torre and Marsiglio (2010). As in the case of sustained growth the optimal dynamics of the state variables are not stationary, we focus on the dynamics of the capital ratio variables, and we show that, through appropriate log-transformations, they can be converted into affine iterated function systems converging to an invariant distribution supported on some fractal set. This proves that also the steady state of endogenous growth models *i.e.*, the stochastic balanced growth path equilibrium—might have a fractal nature. We also provide some sufficient conditions under which the associated self-similar measures turn out to be either singular or absolutely continuous (for the three-sector model we only consider the singularity).

Keywords: Multi-Sector Growth Models, Endogenous Growth, Fractal, Self-Similar Measure, Singular and Absolutely Continuous Measure

JEL Classification: C61, O41

1 Introduction

Almost thirty years later the seminal work of Boldrin and Montrucchio (1986), it is now well known that also traditional (macro)economic models may give rise to complicated dynamics, including random dynamics eventually converging to (possibly singular) invariant measures supported on fractal sets. Montrucchio and Privileggi (1999) borrowing from the iterated function system (IFS) literature (Hutchinson, 1981; Barnsley, 1989; Vrscay, 1991) firstly show that

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standard stochastic economic growth models may show optimal dynamics defined by an IFS. The traditional one sector growth model with Cobb-Douglas production and logarithmic utility has been extensively studied later. Mitra et al. (2003) show that its optimal path converges to a singular measure supported on a Cantor set, characterizing singularity versus absolute continuity of the invariant probability in terms of the parameters' values. Mitra and Privileggi (2004, 2006, 2009) further generalize the model and provide also an estimate of the Lipschitz constant for the maps of the optimal policy defined by an IFS.¹ Only recently, the analysis has been extended in order to consider two-sector growth models, nowadays predominant in economic growth theory. La Torre et al. (2011) show that in a two sector model with physical and human capital accumulation the optimal dynamics for the state variables can be converted through an appropriate log-transformation into an IFS converging to an invariant measure supported on a generalized Sierpinski gasket.

The aim of this paper it to further extend the analysis of fractal outcomes in optimal economic growth models by studying the behavior of multi-sector endogenous growth models. Indeed, thus far the focus has always been placed on neoclassical growth models, in which at steady state the economic growth rate is null, and nothing has been said on whether also a perpetually growing economy (*i.e.*, an economy experiencing a strictly positive steady state growth rate) may achieve a fractal-type steady state. We thus analyze two alternative models of endogenous growth, specifically a two-sector and a three-sector model, based on the Lucas-Uzawa (1988) and La Torre and Marsiglio (2010) models, respectively. We show that even whenever perpetual growth is admissible the economy may develop along a (stochastic) balanced growth path equilibrium characterized by a fractal nature. However, since in such a framework the optimal dynamics of (physical, human and technological) capital are not stationary, we consider the dynamics of the capital ratio variables (specifically, the physical to human capital and technological to human capital ratios) and show that, through appropriate log-transformations, they can be converted into affine IFS converging to singular distributions supported on some fractal set. We then also provide some sufficient conditions under which the associated self-similar measures turn out to be singular and absolutely continuous.

The paper proceeds as follows. In Section 2 the main results from the IFS theory that we will need in our analysis are briefly recalled. In Section 3 we consider the simplest form of multisector endogenous growth models, namely a Lucas-Uzawa (1988) model driven by human capital accumulation. In Section 4 we analyze an extended version of the model, that is a three sector model, as in La Torre and Marsiglio (2010), in which human capital is endogenously allocated across three (physical, human and knowledge) sectors. For both the models, we derive the optimal dynamics and construct an affine IFS conjugate to the optimal dynamics of stationary variables (the physical to human capital, and, in the latter, also the knowledge to human capital ratios). We provide, directly in terms of parameters of the model, sufficient conditions for the attractor of this conjugate IFS to be a fractal set (the Cantor set for the two sector model and the Sierpinski gasket for the three sector model). We also identify sufficient conditions under which the self-similar measures turn out to be singular and absolutely continuous. Section 6 presents concluding remarks and proposes directions for future research.

¹Other recent applications of the IFS theory showing that some economic growth model converge to an invariant distribution supported on a Cantor set are Marsiglio (2012) and Privileggi and Marsiglio (2013). Specifically, Marsiglio (2012) analyzes a two-sector Solow model, while Privileggi and Marsiglio (2013) consider the sustainability problem in a stochastic economy-environment model.

2 Iterated Function Systems

An Iterated Function Systems (IFS) is a finite collection of contractive maps which are defined on a complete metric space. This collection of maps allows to formalize the notion of selfsimilarity and the definition of invariant set or attractor of the IFS. An Iterated Function System with Probabilities (IFSP), instead, consists of the above collection of IFS maps together with an associated set of probabilities. An IFSP induces a Markov operator on the set of all Borel probability measures and a notion of self-similar invariant measure. More details on these can be found in the fundamental works by Hutchinson (1981) and Barnsley and Demko (1985). Applications of these methods are in image compression, approximation theory, signal analysis, denoising, and density estimation (Freiberg et al., 2011; Kunze et al., 2007; Iacus and La Torre, 2005a, 2005b; La Torre et al., 2006, 2009; La Torre and Mendivil, 2008, 2009; La Torre and Vrscay, 2009; Mendivil and Vrscay, 2002a, 2002b). Now we recall, without proofs, some well known basic properties that will be used in the next sections.

We briefly introduce the notion of Hausdorff measure and Hausdorff dimension (more details can be found in Barnsley, 1989). Let (X, d) be a metric space and let diam (E) denote the diameter of a subset E of X. Let $s \ge 0$ and $\delta > 0$, and define

$$\mathcal{H}^{s}_{\delta}(E) = \inf \left\{ \sum_{k=1}^{\infty} \left[\operatorname{diam}\left(E_{k}\right) \right]^{s} : E \subset \bigcup_{i=1}^{\infty} E_{k}, \operatorname{diam}\left(E_{k}\right) < \delta \right\}.$$

Now let us define

$$\mathcal{H}^{s}(E) = \sup_{\delta > 0} \mathcal{H}^{s}_{\delta}(E) \tag{1}$$

Definition 1 $\mathcal{H}^{s}(E)$ in (1) is called the s-dimensional Hausdorff measure. Furthermore, there exists a unique number $s_{0} \geq 0$ such that $\mathcal{H}^{s}(E) = \infty$ for $0 \leq s < s_{0}$ and $\mathcal{H}^{s}(E) = 0$ for $s > s_{0}$. The number s_{0} is called the Hausdorff dimension of E and it is denoted by $\dim_{H}(E)$.

In what follows, let (X, d) be a complete metric space and $w = \{w_0, \ldots, w_{m-1}\}$ a set of m injective contraction maps $w_i : X \to X$, to be referred to as an m-map IFS. Let $0 < \lambda_i < 1$ denote the contraction factors of w_i and define $\lambda := \max_{i \in \{0,\ldots,m-1\}} \lambda_i$; clearly $0 < \lambda < 1$. Associated with the IFS mappings w_0, \ldots, w_{m-1} there is a set-valued mapping $\hat{w} : \mathcal{K}(X) \to \mathcal{K}(X)$ defined over the space $\mathcal{K}(X)$ of all non-empty compact sets in X as

$$\hat{w}(A) := \bigcup_{i=0}^{m-1} w_i(A), \qquad \forall A \in \mathcal{K}(X),$$
(2)

where $w_i(A) = \{w_i(x) : x \in A\}$ is the image of A under $w_i, i = 0, 1, \ldots, m-1$. Let $\hat{w}^t(A) = \hat{w}[\hat{w}^{t-1}(A)]$ for all $t \ge 1$, with $\hat{w}^0(A) = A$. The Hausdorff distance d_H is defined as

$$d_{H}(A,B) = \max\left\{\sup_{x \in A} \inf_{y \in B} d(x,y), \sup_{x \in B} \inf_{y \in A} d(x,y)\right\}, \qquad \forall A, B \in \mathcal{K}(X)$$

Theorem 1 (Hutchinson, 1981) ($\mathcal{K}(X), d_H$) is a complete metric space and \hat{w} is a contraction mapping on ($\mathcal{K}(X), d_H$):

$$d_{H}\left(\hat{w}\left(A\right),\hat{w}\left(B\right)\right) \leq \lambda d_{H}\left(A,B\right), \qquad \forall A,B \in \mathcal{K}\left(X\right)$$

Therefore, there exists a unique set $A^* \in \mathcal{K}(X)$, such that $\hat{w}(A^*) = A^*$, the so-called attractor (or invariant set) of the IFS \hat{w} . Moreover, for any $A \in \mathcal{K}(X)$, $d_H(\hat{w}^t(A), A^*) \to 0$ as $t \to \infty$.

2.1 Invariant Measures

Let $p = (p_0, p_1, \ldots, p_{m-1}), 0 < p_i < 1, i = 0, 1, \ldots, m-1$, be a partition of unity associated with the IFS mappings w_i , so that $\sum_{i=0}^{m-1} p_i = 1$ (each p_i represents a probability value attached to w_i). Let $\mathcal{M}(X)$ be the space of probability measures defined on the σ -algebra $\mathcal{B}(X)$ of Borel measurable subsets of X and define for some $a \in X$ the set $\mathcal{M}_1(X) = \{\mu \in \mathcal{M}(X) : \int_X d(a, x) d\mu(x) < \infty\}$. Define the *Monge-Kantorovich distance* as

$$d_M(\mu,\nu) = \sup\left\{\int_X f d(\mu-\nu) : f \in Lip_1(X)\right\}, \qquad \forall \mu,\nu \in \mathcal{M}_1(X)$$

where $\operatorname{Lip}_1(X)$ is the set of all Lipschitz functions on X with Lipschitz constant equal to 1. Associated with the IFS (w; p) is the so-called Markov operator, $M : \mathcal{M}_1(X) \to \mathcal{M}_1(X)$, defined as

$$(M\mu)(A) = \sum_{i=0}^{m-1} p_i \mu\left[w_i^{-1}(A)\right], \qquad \forall A \in \mathcal{B}(X),$$

where $w_i^{-1}(S) = \{y \in X : w_i(y) \in S\}$. Let $M^t \mu = M(M^{t-1}\mu)$ for all $t \ge 1$, with $M^0 \mu = \mu$.

Theorem 2 (Barnsley et al., 2008) If X is a separable complete metric space $(\mathcal{M}_1(X), d_M)$ is a complete metric space; furthermore, if X is compact, then $\mathcal{M}(X) = \mathcal{M}_1(X)$ and both are compact metric spaces under d_M .

M is a contraction mapping on $(\mathcal{M}_1(X), d_M)$, specifically,

$$d_M(M\mu, M\nu) \leq \left(\sum_i p_i \lambda_i\right) d_M(\mu, \nu), \quad \forall \mu, \nu \in \mathcal{M}_1(X),$$

and thus there exists a unique probability measure $\mu^* \in \mathcal{M}_1(X)$, called invariant measure of the IFS (w; p), such that $M\mu^* = \mu^*$.

For any $\mu \in \mathcal{M}_1(X)$, $d_M(M^t\mu, \mu^*) \to 0$ as $t \to \infty$, which implies that $M^t\mu$ converges weakly to μ^* .

If the contraction mappings w_i are assumed to be *similitudes*, *i.e.*, if there exist numbers $0 < \lambda_i < 1$ such that

$$d(w_i(x), w_i(y)) = \lambda_i d(x, y), \qquad \forall x, y \in X, \quad i = 0, \dots, m-1,$$
(3)

the attractor A^* and the invariant measure μ^* are both said to be *self-similar*. As it is common in the IFS literature, however, in what follows we shall refer to any invariant measure μ^* generated by some IFS as 'self-similar'.

2.2 Absolutely Continuous vs. Singular Measures

It is known that the self-similar invariant measure μ^* determined by a IFS (w; p) generated by similitudes can be either *absolutely continuous* or *singular* with respect to the *n*-dimensional Lebesgue measure, according to the following definitions.

Definition 2 Two positive measures μ and ν defined on a measurable space (Ω, Σ) are called singular if there exist two disjoint sets A and B in Σ whose union is Ω such that μ is zero on all measurable subsets of B while ν is zero on all measurable subsets of A. This is denoted by $\mu \perp \nu$. **Definition 3** If μ and ν are two measures defined on a measurable space (Ω, Σ) , we say that ν is absolutely continuous with respect to μ if $\nu(A) = 0$ for any $A \in \Sigma$ such that $\mu(A) = 0$. The absolute continuity of ν with respect to μ is denoted by $\nu \ll \mu$.

This distinction is crucial as in the former case μ^* can be represented by a density depending on some parameters, while in the latter case there is no simple way to represent it—one actually has to list all its values on every point in its support. The mathematical literature so far has dealt with this issue by trying to characterize absolute continuity vs. singularity of μ^* in terms of the parameters characterizing the IFS (w; p), specifically, in terms of λ_i s and p_i s configurations.

Especially the one-dimensional 2-maps IFS $(\lambda x, \lambda x + (1 - \lambda); p, (1 - p))$, with $0 < \lambda < 1$ and $0 , characterized by the same contraction factor <math>\lambda$ in both maps, has received much attention since the first half of the twentieth century, as its invariant measure μ^* is the same as that of the Erdös series $\sum_{s=0}^{\infty} \pm \lambda^s$ [it being understood that the minus sign is taken with probability p and the plus sign with probability (1-p)] translated over the interval [0,1](see Mitra et al., 2003). For p = 1/2 the topic is known as the study of "symmetric infinite Bernoulli convolutions"; an exhaustive survey on the whole history of this subject can be found in Peres et al. (2000). It is straightforward to see that for $0 < \lambda < 1/2$ and any 0 thesupport of μ^* is a Cantor set of Lebesgue measure zero, so that μ^* must be singular, while when $\lambda = 1/2$ and $p = 1/2 \mu^*$ turns out to be the uniform (Lebesgue) measure over [0, 1], which is clearly absolutely continuous. More complex, and hitherto incomplete, is the analysis for parameter values $1/2 \le \lambda < 1$ and $0 , for which the support of <math>\mu^*$ is the 'full' interval [0,1]. Solomvak (1995) made a real breakthrough when he established that, when p = 1/2. for almost every $1/2 < \lambda < 1 \mu^*$ has density in $L^2(\mathbb{R})$ and for almost every $2^{-1/2} < \lambda < 1$ the density is bounded and continuous (see also Peres and Solomyak, 1996, and Peres and Schlag, 2000). To the best of our knowledge, the contribution by Peres and Solomyak (1998), established for a generic family of contracting similitudes on the real line—including the general case $1/2 \le \lambda < 1$ and 0 for the 2-maps IFS above— is still the most advanced availablein the literature and proves useful for our purposes in the next Section 3. We summarize below their results, together with the simpler cases discussed before, on the basis of the analysis in Mitra et al. (2003), so to provide an overall picture of the state of the art.

Theorem 3 Consider the 2-maps IFS $(\lambda x, \lambda x + (1 - \lambda); p, (1 - p))$ on [0, 1], with $0 < \lambda < 1$ and $0 , and let <math>A^*, \mu^*$ be its corresponding self-similar attractor and measure respectively.

- 1. For any $0 < \lambda < 1/2$ and $0 <math>A^*$ is a Cantor set and μ^* is singular.
- 2. If $\lambda = 1/2$, then A^* is the full interval [0, 1] and
 - (a) if $p \neq 1/2$, then μ^* is singular,
 - (b) if p = 1/2, then μ^* is absolutely continuous—it is the uniform (Lebesgue) measure over [0, 1].
- 3. If $1/2 < \lambda < 1$ again A^* is the full interval [0, 1] and
 - (a) if $1/2 < \lambda < p^p (1-p)^{1-p}$ (the entropy curve), then μ^* is singular,
 - (b) (Peres and Solomyak, 1998) if $1/3 \le p \le 2/3$, then μ^* is absolutely continuous for Lebesgue a.e. $p^p (1-p)^{1-p} < \lambda < 1$,

(c) (Peres and Solomyak, 1998) if $0.156 or <math>2/3 , then <math>\mu^*$ is absolutely continuous for Lebesgue a.e. $p^p (1-p)^{1-p} < \lambda < 0.649$, while, for any $1 < \gamma \leq 2$ such that $[p^{\gamma} + (1-p)^{\gamma}]^{1/(\gamma-1)} < 0.649$, μ^* has density in L^{γ} for Lebesgue a.e. $[p^{\gamma} + (1-p)^{\gamma}]^{1/(\gamma-1)} \leq \lambda < 0.649$.

Theorem 3 still leaves room for further research, as the parameter configurations $0 and <math>2/3 for <math>0.649 \le \lambda < 1$ remain unsolved. Figure 5 in Mitra et al. (2003) draws a bifurcation diagram reporting all cases described in Theorem 3.

Only little progress has been made after the results just stated. It is worth mentioning Niu and Xi (2007) who establish singularity of the self-similar measure for very peculiar IFS with m maps that are similitudes on the real line. Here we report the following result, proved in Ngai and Wang (2005), that generalize some of the results of Theorem 3 to m-maps IFS on \mathbb{R}^n having similitudes characterized by different contraction factors $0 < \lambda_i < 1$, namely

$$w_i(x) = \lambda_i Q_i x + \xi_i, \qquad i = 0, \dots, m - 1, \tag{4}$$

where Q_i is a $n \times n$ orthogonal matrix and $\xi_i \in \mathbb{R}^n$ for each *i*. We denote by $A^* \subset \mathbb{R}^n$ the corresponding attractor and by \mathcal{L}^n the *n*-dimensional Lebesgue measure.

Definition 4 The IFS satisfies the open set condition (OSC) if there exists a nonempty open set U such that $w_i(U) \subset U$ for all i = 0, ..., m-1 and $w_i(U) \cap w_i(U) = \emptyset$ for all $i \neq j$.

Theorem 4 (Ngai and Wang, 2005) Let (w; p) be an IFS on \mathbb{R}^n with maps $w_i : \mathbb{R}^n \to \mathbb{R}^n$ defined by (4) and let $p = (p_0, p_1, \ldots, p_{m-1})$ be the associated probability weights. Denote by μ^* the self-similar invariant measure supported over A^* defined by (w; p).

i) If $\prod_{i=0}^{m-1} p_i^{p_i} \lambda_i^{-np_i} > 1$, then μ^* is singular.

ii) If $\prod_{i=0}^{m-1} p_i^{p_i} \lambda_i^{-np_i} = 1$ but $p_i \neq \lambda_i^n$ for some *i*, then μ^* is singular.

iii) If $p_i = \lambda_i^n$ for all i = 0, ..., m - 1, then μ^* is absolutely continuous if and only if the IFS (w; p) satisfies the open set condition (OSC). In this case $\mu^* = \delta \mathcal{L}^n|_{A^*}$, where A^* is the attractor of the IFS, and $\delta = 1/\mathcal{L}^n(A^*)$; that is, μ^* is the uniform (n-dimensional Lebesgue) measure over $A^* \subset \mathbb{R}^n$.

Case (i) generalizes cases 1, 2a and 3a of Theorem 3 altogether to IFS with any number of similitudes on \mathbb{R}^n with different contraction factors λ_i ; case (ii) actually adds an important piece of information by extending the conclusion in 3a of Theorem 3 to the generalized boundary (entropy) curve corresponding to $\lambda = p^p (1-p)^{1-p}$ there; finally, case (iii) generalizes case 2b of Theorem 3. Indeed, in the latter scenario the OSC requires that the image sets of the similitudes of the attractor, $w_i (A^*)$, have only "small overlap" (sometimes called "just touching"), which—because in Theorem 3 the two maps are assumed to have both the same slope, $\lambda_0 = \lambda_1 = \lambda$ —for the IFS ($\lambda x, \lambda x + (1 - \lambda); p, (1 - p)$) translates into $\lambda = p = 1/2$, which implies that A^* is the full interval [0, 1] and the image sets are equal to w_0 ([0, 1]) = [0, 1/2] and w_1 ([0, 1]) = [1/2, 1], having only the unique point x = 1/2 in common (see Schief, 1994). Theorem 1.2 in Ngai and Wang (2005) to some extent generalizes results 3b and 3c of Theorem 3 to IFS with any number of similitudes with different contraction factors, however only for maps on the real line.

In Section 4 we consider a three sector model which gives rise to an affine three-map IFS with constant diagonal linear parts [see (38)]. The IFS maps act on $[0, 1]^2$ and have fixed points

(0,0), (1,0) and (1/2,1). For notational simplicity, we take the linear parts all given by the matrix

$$Q = \begin{bmatrix} a & 0\\ 0 & b \end{bmatrix}.$$
 (5)

There are not many results known about this situation, but here we indicate a simple sufficient condition for singularity of the invariant measure.

The simplest case is when the attractor of the IFS (the support of μ^*) has dimension strictly less than 2, in which case it is obviously singular; so we start with this case. Define the function $\Phi(s)$ by

$$\Phi(s) = \begin{cases} \max\{b^s, a^s\} & \text{if } s \le 1\\ \max\{ab^{s-1}, ba^{s-1}\} & \text{if } 1 < s \le 2. \end{cases}$$
(6)

Then by results of Falconer and Miao (2007), an upper bound for the dimension \dim_H of the attractor A^* of this IFS is given by

$$\dim_H \left(A^* \right) \le \min \left\{ 2, \Phi^{-1} \left(\frac{1}{3} \right) \right\}.$$
(7)

It can be shown that Φ is a strictly decreasing function of s and thus there is a unique s such that $\Phi(s) = 1/3$. Assuming that $a \leq b$, we have $a^s \leq b^s$ for all s and $ab^{s-1} \leq ba^{s-1}$ for $s \leq 2$. Thus, under $a \leq b$ we have

$$\Phi(s) = \begin{cases} b^s & \text{if } 0 \le s \le 1\\ ba^{s-1} & \text{if } 1 < s \le 2. \end{cases}$$

Thus, we have either

$$\dim_{H} (A^{*}) \leq -\frac{\ln 3}{\ln b} \quad \text{or} \quad \dim_{H} (A^{*}) \leq \frac{\ln \left[a/(3b)\right]}{\ln a}.$$

Therefore, $\dim_H (A^*) < 2$ if ab < 1/3.

The results in Falconer and Miao (2007) require the IFS to satisfy the OSC in order for the dimension result to be exact. Hence, it is also possible that the dimension is less than 2 for a larger range of parameters than that given by ab < 1/3.

The above (dimension-based) result can be extended by proving the following result, which can be thought of as an analogue of Theorem 3a or Theorem 4i. The strategy of the proof is the same as Theorem 4i, but we include it since we use the entropy in an essentially different way. In the sequel we will denote by $\{0, 1, 2\}^{\mathbb{N}}$ the collection of infinite words in the alphabet $\{0, 1, 2\}$ and by $\{0, 1, 2\}^n$ the collection of words of length n.

Theorem 5 Let (w; p) be an IFS on \mathbb{R}^2 having maps $w_i : \mathbb{R}^2 \to \mathbb{R}^2$ defined by $w_i(x) = Qx + \xi_i$, with Q as in (5), $\xi_i \in \mathbb{R}^2$ for i = 0, 1, 2, and let $p = (p_0, p_1, p_2)$ be the associated probability weights. If $ab < p_0^{p_0} p_1^{p_1} p_2^{p_2}$, the invariant measure μ^* defined by (w; p) is singular.

Proof. For $\sigma \in \{0, 1, 2\}^{\mathbb{N}}$ and for $n \in \mathbb{N}$, i = 0, 1, 2, define

$$\sigma(n,i) = \# \left\{ j \le n : \sigma_j = i \right\}.$$

Fix $k \in \mathbb{N}$ and consider the set

$$S_n = \left\{ \sigma \in \{0, 1, 2\}^n : \left| \frac{\sigma(n, i) - np_i}{\sqrt{n}\sqrt{p_i(1 - p_i)}} \right| \le k, i = 0, 1, 2 \right\}$$

so that S_n is the set of k-typical sequences of length n. Then from Chebyshev's inequality we have

$$\Pr\left(S_n\right) \ge 1 - \frac{3}{k^2}$$

which is independent of n. Furthermore, by Theorem 1.3.4 in Roman (1992) we have

$$\#S_n \le (p_0^{p_0} p_1^{p_1} p_2^{p_2})^{-n} \, 3^{C\sqrt{n}}$$

for some constant C > 0. For notational ease define

$$w_{\sigma} = w_{\sigma_1} \circ w_{\sigma_2} \circ \cdots \circ w_{\sigma_n}$$

as the *n*-fold composition given by $\sigma \in \{0, 1, 2\}^{\mathbb{N}}$. Now, let

$$J_n = \bigcup_{\sigma \in S_n} w_\sigma\left(K\right) \subset K$$

Then $\mu(J_n) = \Pr(S_n) \ge 1 - 3/k^2$ uniformly in n and

$$\mathcal{L}^{2}(J_{n}) \leq \sum_{\sigma \in S_{n}} \mathcal{L}^{2}[w_{\sigma}(K)] \leq \sum_{\sigma \in S_{n}} \mathcal{L}^{2}[w_{\sigma}([0,1]^{2})] = \sum_{\sigma \in S_{n}} (ab)^{n} = \#S_{n}(ab)^{n}$$
$$\leq \left(\frac{ab}{p_{0}^{p_{0}}p_{1}^{p_{1}}p_{2}^{p_{2}}}\right)^{n} 3^{C\sqrt{n}} \to 0$$

as $n \to \infty$. Thus, $\mu^* \perp \mathcal{L}^2$.

Notice that the minimum value of $p_0^{p_0} p_1^{p_1} p_2^{p_2}$ is 1/3 and occurs when $p_0 = p_1 = p_2 = 1/3$. Thus if $ab < 1/3 \ \mu^*$ is singular for any choice of p_0, p_1, p_2 and so this result includes the dimension-based result previously mentioned.

Finally, Theorems 22 and 23 in Shmerkin (2006) imply that for any probabilities p_i if the Lebesgue measure of A^* is positive, then either $\mu^* \perp \mathcal{L}^2$ or we have both $\mu^* \ll \mathcal{L}^2$ and $\mathcal{L}^2|_{A^*} \ll \mu^*$ (the two measures are *mutually absolutely continuous*). Specifically, as if $\mathcal{L}^2(A^*) = 0$ we automatically have $\mu^* \perp \mathcal{L}^2$, this always implies that either $\mu^* \perp \mathcal{L}^2$ or they are mutually absolutely continuous.

3 The Lucas-Uzawa Model

We first study the Uzawa-Lucas (1988) model of endogenous growth where the social planner seeks to maximize social welfare subject to the physical and human capital constraints, choosing consumption, $c_t > 0$, and the share of human capital employed in physical production, $0 < u_t < 1$. The welfare is the expected infinite discounted $(0 < \beta < 1$ is the pure rate of time preference) sum of instantaneous utilities [assumed to be logarithmic, $u(c_t) = \ln c_t$]. Physical capital, k_t , accumulates according to the difference between the production of the unique final good and consumption, $k_{t+1} = y_t - c_t$, where production is assumed to use a Cobb-Douglas technology combining physical and the allocated share of human capital, $y_t = k_t^{\alpha} (u_t h_t)^{1-\alpha}$, where $0 < \alpha < 1$ is the physical capital share and u_t the proportion of human capital employed in physical production. The remaining share, $1 - u_t$ is used to produce new human capital, h_t , according to a linear technology; thus, the law of motion of human capital is equal to: $h_{t+1} = b (1 - u_t) h_t$, with b > 0. However, the final sector is affected by exogenous perturbations, z_t , which enter multiplicatively the production function; these shocks are independent and identically distributed, and take on two values: $z_t \in \{q, 1\}$, where 0 < q < 1. We assume that educational choices are not affected at all by eventual shocks. This means that at any time, given the realization of the random shocks, the economy may be in two alternative situations: i) an economic crisis due to a supply shock lowering physical productivity, corresponding to $z_t = q$, and ii) a business-as-usual scenario with no shocks in which the whole economy evolves along its full capacity, corresponding to $z_t = 1$. These two alternative scenarios occur with (constant) probability p and 1 - p, respectively.

The social planner problem consists of choosing the level of consumption and the shares of human capital to allocate to the physical sector in order to maximize social welfare, taking into account the dynamic evolution of physical and human capital, the presence of random shocks $z_t \in \{q, 1\}$, and the given initial conditions k_0 , h_0 , and z_0 :

$$V(k_0, h_0, z_0) = \max_{\{c_t, u_t\}_{t=0}^{\infty}} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \ln c_t$$
(8)

s.t.
$$\begin{cases} k_{t+1} = z_t k_t^{\alpha} \left(u_t h_t \right)^{1-\alpha} - c_t \\ h_{t+1} = b \left(1 - u_t \right) h_t \\ k_0 > 0, \ h_0 > 0, \ z_0 \in \{q, 1\} \text{ are given,} \end{cases}$$
(9)

where \mathbb{E}_0 denotes expectation at time t = 0. The Bellman equation associated to (8) reads as:

$$V(k_t, h_t, z_t) = \max_{c_t, u_t} \left[\ln c_t + \beta \mathbb{E}_t V(k_{t+1}, h_{t+1}, z_{t+1}) \right].$$
(10)

By applying the Verification principle, it is possible to obtain an analytical expression for the value function of the above problem, and consequently derive explicitly the optimal dynamics of physical and human capital (see also Bethmann, 2007, 2013).

Proposition 1 Given the problem in (8) under (9), the following results hold.

i) The Bellman equation (10) has solution given by:

$$V(k_t, h_t, z_t) = \theta + \theta_k \ln k_t + \theta_h \ln h_t + \theta_z \ln z_t,$$

where $\theta_i, j \in \{k, h, z\}$, are defined as follows:

$$\theta_k = \frac{\alpha}{1 - \alpha \beta}, \qquad \theta_h = \frac{1 - \alpha}{(1 - \alpha \beta)(1 - \beta)}, \qquad \theta_z = \frac{1}{1 - \alpha \beta},$$

and the constant term θ is given by:

$$\begin{aligned} \theta &= \frac{1}{1-\beta} \left[\ln\left(1-\alpha\beta\right) + \frac{\alpha\beta}{1-\alpha\beta} \ln\left(\alpha\beta\right) + \frac{1-\alpha}{1-\alpha\beta} \ln\left(1-\beta\right) \\ &+ \frac{\left(1-\alpha\right)\beta}{\left(1-\alpha\beta\right)\left(1-\beta\right)} \ln\beta + \frac{\beta}{\left(1-\alpha\beta\right)} \mathbb{E}\ln z \right]. \end{aligned}$$

ii) The optimal policy rules for consumption, c_t , and share of human capital allocated to physical production, u_t , are respectively given by:

$$c_t = (1 - \alpha\beta) (1 - \beta)^{1-\alpha} z_t k_t^{\alpha} h_t^{1-\alpha}$$
$$u_t \equiv 1 - \beta \qquad \forall t.$$

The optimal dynamics of physical, k_{t+1} , and human, h_{t+1} , capital are the following:

$$k_{t+1} = \alpha \beta \left(1 - \beta\right)^{1-\alpha} z_t k_t^{\alpha} h_t^{1-\alpha} \tag{11}$$

$$h_{t+1} = b\beta h_t \tag{12}$$

The proof of Proposition 1 parallels that of Proposition 1 in La Torre et al. (2011); hence, we omit the details.

The results highlighted in Proposition 1 are pretty standard in the literature (see Bethmann, 2007, 2013; La Torre et al., 2011). It is also very well-known that the Lucas-Uzawa (1988) framework, because of the linearity in the production of (new) human capital, may generate sustained long-run growth. Specifically, the system (11)–(12) is diverging whenever $b > 1/\beta$ as, in this case, both physical and human capital grow without any bound. Therefore, in such a scenario we need to introduce a change of variables in order to obtain a system converging to a stationary equilibrium. The natural candidate is the physical to human capital ratio, $\chi_t = k_t/h_t$, which, incidentally, reduces the two-dimensional optimal dynamics (11)–(12) into a one-dimensional dynamic. Indeed, the physical to human capital ratio χ_t evolves over time according to the following stochastic nonlinear difference equation:

$$\chi_{t+1} = \frac{\alpha \left(1 - \beta\right)^{1-\alpha}}{b} z_t \chi_t^{\alpha}.$$
(13)

Denoting by σ the linear coefficient in (13),

$$\sigma = \frac{\alpha \left(1 - \beta\right)^{1 - \alpha}}{b},\tag{14}$$

the nonlinear IFS associated to (13) defined by the two maps

$$\begin{cases} f_0(\chi) = \sigma q \chi^{\alpha} & \text{with probability } p \\ f_1(\chi) = \sigma \chi^{\alpha} & \text{with probability } 1 - p \end{cases}$$
(15)

eventually traps the new variable χ into the compact interval $A = \left[(\sigma q)^{1/(1-\alpha)}, \sigma^{1/(1-\alpha)} \right]$, with endpoints corresponding to the fixed points of f_0 and f_1 respectively; such a scenario corresponds to growth rate values for the original variables, k_t and h_t , oscillating between finite positive bounds. If the IFS (15) converges to some invariant measure supported over (possibly a fractal subset of) A, then we con interpret such a situation as a steady state representing a *stochastic balanced growth path* (SBGP) equilibrium, the stochastic equivalent of a typical equilibrium in deterministic endogenous growth theory.

Note that whenever $\alpha > q$ the IFS (15) turns out to be *non-contractive*, as there exists a right neighborhood of the left fixed point $(\sigma q)^{1/(1-\alpha)}$ on which $f'_1 > 1$.² In this case, the results of Section 2, which provide only sufficient conditions for a (contractive) IFS to converge to a unique invariant measure, cannot be directly applied; however, the logarithmic transformation to the nonlinear dynamics in (13) contemplated by the next Proposition 2 allows us to establish the existence of a unique invariant measure for (15) indirectly. Such transformation yields a conjugate affine system which can be represented by an IFS characterized by contractive similitudes.

Proposition 2 The one-to-one logarithmic transformation $\chi_t \to \varphi_t$ defined by:

$$\varphi_t = -\frac{1-\alpha}{\ln q} \ln \chi_t + 1 + \frac{\ln \sigma}{\ln q},\tag{16}$$

²To see this, compute the derivative of the 'higher' map f_1 on the left fixed point, $(\sigma q)^{1/(1-\alpha)}$: $f'_1\left[(\sigma q)^{1/(1-\alpha)}\right] = \sigma \alpha \left[(\sigma q)^{1/(1-\alpha)}\right]^{\alpha-1} = \sigma \alpha/(\sigma q) = \alpha/q$, which is clearly larger than 1 whenever $\alpha > q$. As $f'_0\left[(\sigma q)^{1/(1-\alpha)}\right] = \alpha < 1$ and both f_0 and f_1 are strictly increasing and strictly concave, α/q is the Lipschitz constant of the IFS (15) over the trapping set $\left[(\sigma q)^{1/(1-\alpha)}, \sigma^{1/(1-\alpha)}\right]$. with σ defined in (14), defines a contractive affine IFS equivalent to the nonlinear dynamics in (13) composed of two maps $w_0, w_1 : [0, 1] \rightarrow [0, 1]$, where 0 and 1 are the fixed points of w_0 and w_1 respectively, given by:

$$\begin{cases} w_0(\varphi) = \alpha\varphi & \text{with probability } p \\ w_1(\varphi) = \alpha\varphi + (1-\alpha) & \text{with probability } 1-p. \end{cases}$$
(17)

The IFS (17) converges weakly to a unique self-similar measure supported on an attractor which is either the interval [0, 1] when $1/2 \le \alpha \le 1$ or a Cantor set when $0 \le \alpha < 1/2$.

Proposition 2 follows immediately from Theorems 1 and 2 of Section 2.

Note that the one-sector stochastic optimal growth model discussed in Mitra et al. (2003) exhibits the same optimal dynamics as in (17). Indeed, besides the different constant σ as in (14), the dynamics described by (13) is the same as the optimal dynamics of capital in the one-sector growth model; hence, also the no-overlap condition $0 \leq \alpha < 1/2$ yielding a support which is a Cantor set is exactly the same. The novelty in our model is that here what converges to an invariant measure supported on a Cantor set is a transformation of the physical to human capital ratio (and not a transformation of physical capital); therefore, we have just shown that also an economy experiencing sustained growth can exhibit a long-run pattern related to some fractal attractor. Specifically, the SBGP equilibrium has a fractal nature.

Note that Proposition 2 uses the physical capital share $0 < \alpha < 1$ as the contraction factor to establish convergence of (17). Because (16) is a one-to-one transformation of the nonlinear IFS (15), an immediate consequence of Proposition 2 is the following Corollary 1 establishing weak convergence of (15) to a unique invariant measure also when $\alpha > q$, that is, when it is non-contractive and falls outside the class of IFS considered in Section 2. We shall say that a set $A \subset \mathbb{R}$ is a generalized (topological) Cantor set if it is totally disconnected and perfect.³

Corollary 1 For any parameters' configuration such that $0 < \alpha < 1$, $0 < \beta < 1$, 0 < q < 1, $0 , and <math>b > 1/\beta$, envisaging sustained growth for the stochastic discrete-time Lucas-Uzawa model, the nonlinear IFS (15) weakly converges to a unique invariant measure supported either over the full interval $A^* = \left[(\sigma q)^{1/(1-\alpha)}, \sigma^{1/(1-\alpha)} \right]$ or over some subset of it. In the latter case, whenever $0 < \alpha < 1/2$ the attractor A^* of (15) is a generalized topological Cantor set contained in $\left[(\sigma q)^{1/(1-\alpha)}, \sigma^{1/(1-\alpha)} \right]$.

Proof. As there is a one-to-one correspondence between (13) and (16), the fist part is an immediate consequence of Proposition 2. To establish the second part, we must show that the no-overlap property for the IFS (15) is the same as the no-overlap property $0 \le \alpha < 1/2$ for the IFS (17): as both maps f_0 , f_1 are strictly increasing, the former property (see Mitra and Privileggi, 2009) can be written as $f_0 \left[\sigma^{1/(1-\alpha)}\right] < f_1 \left[(\sigma q)^{1/(1-\alpha)}\right] \iff \sigma q \left[\sigma^{1/(1-\alpha)}\right]^{\alpha} < \sigma \left[(\sigma q)^{1/(1-\alpha)}\right]^{\alpha} \iff q (\sigma)^{\alpha/(1-\alpha)} < (\sigma q)^{\alpha/(1-\alpha)} \iff q < q^{\alpha/(1-\alpha)} \iff q^{(2\alpha-1)/(1-\alpha)} > 1 \iff \alpha < 1/2$, establishing the result.

Corollary 1 states that, if $0 < \alpha < 1/2$, the attractor of the nonlinear IFS (15) is a distorted (due to nonlinearity of the maps in the IFS) Cantor set contained in the interval

³Recall that a set $A \subseteq X$, where (X, d) is a metric space, is said to be *totally disconnected* if its only connected subsets are one-point sets: formally, for any two distinct points x, y in A, there are two non-empty open disjoint sets U and V such that $x \in U$, $y \in V$ and $(U \cap A) \cup (V \cap A) = A$. A set $A \subseteq X$ is said to be *perfect* if it is equal to the set of its accumulation points; that is, it is a closed set which contains no isolated points.

 $\left[\left(\sigma q\right)^{1/(1-\alpha)}, \sigma^{1/(1-\alpha)}\right]$. Mitra and Privileggi (2004, 2006, 2009) dealt with nonlinear IFS generating optimal policies for the one-sector stochastic optimal growth model, starting from a class of the type in (15) and then increasing the degree of generality; however, all results established by the authors hold under assumptions on the primitives of the model guaranteeing that such IFS satisfy the contraction property, on which all results in Section 2 are based. Proposition 2, joint with Corollary 1, expands these results outside the family of contractive IFS, at least for the specific functional forms chosen here. Clearly the logarithmic-Cobb-Douglas specification of our model plays a major role in allowing for the transformation (16).⁴

Remark 1 The literature on IFS actually encompasses a more general result in which the assumption that the IFS mappings are contractions can be slightly relaxed and substituted with a broader notion of "contraction on average"; in this case existence and uniqueness of an invariant measure μ^* for the IFS can still be established (see, e.g., Proposition 5.1 in Diaconis and Friedman, 1999). Specifically, denoting by λ_i the Lipschitz constant associated to the *i*th map of the IFS, for i = 0, ..., m - 1, the IFS is contracting on average if

$$\sum_{i=0}^{m-1} p_i \ln \lambda_i < 0, \tag{18}$$

in which smaller probability weights p_i associated to steeper maps neutralize their effect when they are not contractions, i.e., when $\lambda_i > 1$.

In our model the Lipschitz constants of the two maps in (15) correspond to the slopes of f_0 and f_1 on the smallest fixed point $(\sigma q)^{1/(1-\alpha)}$: $\lambda_0 = \alpha$ and $\lambda_1 = \alpha/q$ respectively (see footnote 2). In this case (18) translates into $p \ln \alpha + (1-p) \ln (\alpha/q) = \ln \alpha - (1-p) \ln q < 0$, which is equivalent to $\alpha/q < e^{1-p}$, clearly a more general assumption than the requirement for both maps to be contractive—corresponding to $\alpha/q < 1$. However, the assumptions of Corollary 1 clearly include the case $\alpha/q \ge e^{1-p}$, stressing further its contribution outside the general theory.

Figure 1 shows an example of parameters' configuration such that the IFS (15) is noncontractive, having Lipschitz constant, λ_1 , associated to f_1 larger than 1. Specifically, setting the fundamentals as $\alpha = 1/3$, q = 1/6, p = 2/3, $\beta = 0.96$ and b = 1.052 (so that there is sustained growth as $b > 1.042 = 1/\beta$), it turns out that $\sigma = 0.037$ in (14), the fixed points of the two nonlinear maps f_0, f_1 are $\chi_0^* = (\sigma q)^{1/(1-\alpha)} = 0.0005$ and $\chi_1^* = 0.007$ respectively, so that the trapping interval is $[\chi_0^*, \chi_1^*] = [0.0005, 0.007]$, and their Lipschitz constants are $\lambda_0 = f'_0(\chi_0^*) = \alpha = 1/3$ and $\lambda_1 = f'_1(\chi_0^*) = \alpha/q = 2 > e^{1-p} = 1.4 > 1$; thus, the IFS is noncontractive, not even 'on average'. Moreover, as $\alpha = 1/3 < 1/2$, in the figure it can be clearly seen that the images $f_0([\chi_0^*, \chi_1^*])$ and $f_1([\chi_0^*, \chi_1^*])$ do not overlap [because $f_0(\chi_1^*) < f_1(\chi_0^*)]$, so that, according to Corollary 1, the attractor A^* is a generalized topological Cantor set in $[\chi_0^*, \chi_1^*]$.

While we are not aware of results establishing singularity vs. absolute continuity of the invariant measures for nonlinear IFS, Theorem 3 of Section 2 can be applied to provide a partial analysis of the affine IFS (17), which is summarized below. Note that such result is independent of the size of the shock q, the rate of time preference β and the human capital productivity coefficient b.

⁴Incidentally, we note that the same conclusion can be drawn for the one-sector stochastic optimal growth model in Mitra et al. (2003): the nonlinear IFS representing the optimal capital dynamics there converges weakly to a unique invariant measure also when the ratio physical capital share over the technological shock is larger than one, implying that the nonlinear IFS is not a contraction.



FIGURE 1: the nonlinear maps f_0 and f_1 in (15) when $\alpha = 1/3$, q = 1/6, p = 2/3, $\beta = 0.96$ and b = 1.052. Such IFS is non-contractive, as $\lambda_1 = 2 > 1$, and its attractor is a generalized topological Cantor set, as the images of the two maps do not overlap because $f_0(\chi_1^*) < f_1(\chi_0^*)$.

Proposition 3 Let μ^* be the self-similar measure associated to the IFS (17), $(\alpha\varphi, \alpha\varphi + (1-\alpha); p, (1-p))$, on [0, 1].

- i) If $0 < \alpha < p^p (1-p)^{1-p}$, then μ^* is singular.
- ii) If $\alpha = p^p (1-p)^{1-p}$ and $p \neq 1/2$, then μ^* is singular.
- iii) If $\alpha = p = 1/2$, then μ^* is absolutely continuous—it is the uniform (Lebesgue) measure over [0, 1].
- iv) If $1/3 \le p \le 2/3$, then μ^* is absolutely continuous for Lebesgue a.e. $p^p (1-p)^{1-p} < \alpha < 1$.
- **v)** If $0.156 or <math>2/3 , then <math>\mu^*$ is absolutely continuous for Lebesgue a.e. $p^p (1-p)^{1-p} < \alpha < 0.649$, while, for any $1 < \gamma \le 2$ such that $[p^{\gamma} + (1-p)^{\gamma}]^{1/(\gamma-1)} < 0.649$, μ^* has density in L^{γ} for Lebesgue a.e. $[p^{\gamma} + (1-p)^{\gamma}]^{1/(\gamma-1)} \le \alpha < 0.649$.

4 A Three Sector Model

We now analyze an extended Lucas-Uzawa model, as presented in La Torre and Marsiglio (2010) and in Marsiglio (2012), where human capital needs to be endogenously allocated across three sectors: besides the physical and human capital sector as in the standard Lucas-Uzawa

framework, also in the knowledge or technological sector.⁵ In this framework the planner has to choose consumption, c_t , and the share of human capital to allocate into physical, u_t , and knowledge, v_t , production, taking into account the presence of random shocks, which affect both physical and technological capital production. The technological knowledge, a_t , evolves over time because of newly produced knowledge: $a_{t+1} = y_t^a$. Both the production of the final consumption good, y_t^k , and knowledge, y_t^a , use a Cobb-Douglas technology. Physical production combines physical capital, knowledge capital and the allocated share of human capital: $y_t^k = k_t^{\alpha} (u_t h_t)^{\gamma} a_t^{1-\alpha-\gamma}$, with $0 < \alpha < 1$ and, $0 < \gamma < 1 - \alpha$. Knowledge production combines the existing stock of knowledge and the allocated share of human capital: $y_t^a = (v_t h_t)^{\phi} a_t^{1-\phi}$, with $0 < \phi < 1$. The remaining share of human capital, $1 - u_t - v_t > 0$, $0 < u_t, v_t < 1$, is used to generate new human capital according to a linear technology (as in Section 3): $h_{t+1} = b (1 - u_t - v_t) h_t$, with b > 0. We now assume that both the production of the final good and knowledge are affected by exogenous random shocks. As in the previous section, we keep assuming that educational choices are not affected by productivity shocks.

Specifically, the shocks are independent and identically distributed, and take on finite values: $z_t = \{q_1, q_2, 1\}$ and $\eta_t \in \{r, 1\}$, with $0 < q_1 < q_2 < 1$ and 0 < r < 1, are the shocks affecting multiplicatively the physical and knowledge sector, respectively. As in La Torre et al. (2011), we assume that only three pairs of shock values can occur with positive probability, $(z_t, \eta_t) \in \{(q_1, r), (q_2, 1), (1, 1)\}$, each with (constant) probability p_0 , p_1 and p_2 respectively, where $0 < p_i < 1$, i = 0, 1, 2, and $\sum_{i=0}^2 p_i = 1$. Such three shock configurations may be interpreted as: i) a deep economic-financial crisis having wide effects on the economy as a whole and thus involving both production and knowledge sectors, corresponding to $(z_t, \eta_t) = (q_1, r)$, ii) a negative supply shock, as an increase in raw materials' prices (*e.g.*, oil), affecting only final production but not knowledge, corresponding to $(z_t, \eta_t) = (q_2, 1)$, and iii) a business-as-usual scenario, corresponding to $(z_t, \eta_t) = (1, 1)$.

The social planner problem consists of choosing the level of consumption and the shares of human capital to allocate into physical and knowledge production in order to maximize social welfare, taking into account the dynamic evolution of physical, human and technological capital, the presence of random shocks $z_{t+1} = \{q_1, q_2, 1\}$ and $\eta_{t+1} = \{r, 1\}$, and the given initial conditions k_0 , h_0 , a_0 , z_0 and η_0 :

$$V(k_0, h_0, a_0, z_0, \eta_0) = \max_{\{c_t, u_t, v_t\}_{t=0}^{\infty}} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \ln c_t$$
(19)

s.t.
$$\begin{cases} k_{t+1} = z_t k_t^{\alpha} \left(u_t h_t \right)^{\gamma} a_t^{1-\alpha-\gamma} - c_t \\ h_{t+1} = b \left(1 - u_t - v_t \right) h_t \\ a_{t+1} = \eta_t \left(v_t h_t \right)^{\phi} a_t^{1-\phi} \\ k_0 > 0, \ h_0 > 0, \ a_0 > 0, \ z_0 \in \{q_1, q_2, 1\}, \ \eta_0 \in \{r, 1\} \text{ are given,} \end{cases}$$
(20)

where \mathbb{E}_0 denotes expectation at time t = 0. The Bellman equation associated to (19) is:

$$V(k_t, h_t, a_t, z_t, \eta_t) = \max_{c_t, u_t, v_t} \left[\ln c_t + \beta \mathbb{E}_t V(k_{t+1}, h_{t+1}, a_{t+1}, z_{t+1}, \eta_{t+1}) \right].$$
(21)

Following the same steps used in the previous section, it is possible to prove the following.

⁵Note that referring to this sector as technological sector or another form of sector is totally irrelevant for our analysis. What really matters is that the third sector produces a form of capital (different from human and physical capital) which is used in the production of the final consumption good and which is produced using a certain share of human capital. Interpreting this kind of capital as cultural or social capital would work as well; however, in the following, for the sake of simplicity, we refer to this type of capital as technological or knowledge capital as in La Torre and Marsiglio (2010).

Proposition 4 Given the problem in (19) under (20), the following results hold.

i) The Bellman equation (21) has a solution given by:

$$V(k_t, h_t, a_t, z_t, \eta_t) = \theta + \theta_k \ln k_t + \theta_h \ln h_t + \theta_a \ln a_t + \theta_z \ln z_t + \theta_\eta \ln \eta_t,$$

where θ_j , $j \in \{k, h, a, z, \eta\}$ are defined as follows:

$$\theta_k = \frac{\alpha}{1 - \alpha\beta}, \quad \theta_h = \frac{\gamma \left(1 - \beta\right) + \beta \phi \left(1 - \alpha\right)}{\left(1 - \alpha\beta\right) \left(1 - \beta\right) \left[1 - \beta \left(1 - \phi\right)\right]}, \quad \theta_a = \frac{1 - \alpha - \gamma}{\left(1 - \alpha\beta\right) \left[1 - \beta \left(1 - \phi\right)\right]} \\ \theta_z = \frac{1}{1 - \alpha\beta}, \qquad \theta_\eta = \frac{\left(1 - \alpha - \gamma\right) \beta}{\left(1 - \alpha\beta\right) \left[1 - \beta \left(1 - \phi\right)\right]}$$

and the constant term θ is given by:

$$\begin{split} \theta &= \frac{1}{1-\beta} \left\{ \ln\left(1-\alpha\beta\right) + \frac{\gamma}{1-\alpha\beta} \ln \overline{u} + \frac{\alpha\beta}{1-\alpha\beta} \ln\left(\alpha\beta\right) \\ &+ \frac{\beta\gamma\left(1-\beta\right) + \beta^2\phi\left(1-\alpha\right)}{\left(1-\alpha\beta\right)\left(1-\beta\right)\left[1-\beta\left(1-\phi\right)\right]} \ln\left[b\left(1-\overline{u}-\overline{v}\right)\right] \\ &+ \frac{\beta\phi\left(1-\alpha-\gamma\right)}{\left(1-\alpha\beta\right)\left[1-\beta\left(1-\phi\right)\right]} \ln \overline{v} + \frac{\beta}{1-\alpha\beta} \mathbb{E}\ln z + \frac{\left(1-\alpha-\gamma\right)\beta^2}{\left(1-\alpha\beta\right)\left[1-\beta\left(1-\phi\right)\right]} \mathbb{E}\ln\eta \right\}, \end{split}$$

where \overline{u} and \overline{v} are defined in (22) and (23) respectively.

ii) The optimal policy rules for consumption, c_t , share of human capital allocated to physical, u_t , and knowledge, v_t , production are respectively given by:

$$c_{t} = (1 - \alpha\beta) \,\overline{u}^{\gamma} z_{t} k_{t}^{\alpha} h_{t}^{\gamma} a_{t}^{1 - \alpha - \gamma},$$

$$u_{t} \equiv \frac{\gamma \left(1 - \beta\right) \left(1 - \beta + \beta\phi\right)}{\gamma \left(1 - \beta\right) + \beta\phi \left(1 - \alpha\right)} = \overline{u} \qquad \forall t,$$

$$(22)$$

$$v_t \equiv \frac{\beta \phi \left(1 - \alpha - \gamma\right) \left(1 - \beta\right)}{\gamma \left(1 - \beta\right) + \beta \phi \left(1 - \alpha\right)} = \overline{v} \qquad \forall t.$$
(23)

The optimal dynamics of physical, k_{t+1} , human, h_{t+1} , and technological, a_{t+1} , capital are the following:

$$k_{t+1} = \alpha \beta \overline{u}^{\gamma} z_t k_t^{\alpha} h_t^{\gamma} a_t^{1-\alpha-\gamma}, \qquad (24)$$

$$h_{t+1} = b \left(1 - \overline{u} - \overline{v}\right) h_t, \tag{25}$$

$$a_{t+1} = \overline{v}^{\phi} \eta_t h_t^{\phi} a_t^{1-\phi}.$$
(26)

The proof of Proposition 4 is a long and (algebraically) tedious extension of that of Proposition 1 in La Torre et al. (2011); hence, we omit it.

Because the human capital sector employs a linear technology, the economy may experience sustained long-run (La Torre and Marsiglio, 2010); specifically, whenever $b > 1/(1 - \overline{u} - \overline{v})$ physical, human and technological capital grow without any bound according to (24), (25) and (26). Again, by introducing the physical to human capital, $\chi_t = k_t/h_t$, and the knowledge to human capital, $\omega_t = a_t/h_t$, ratio variables, we can recast the previous system into a twodimensional nonlinear system converging to some compact trapping subset of \mathbb{R}^2 , which may provide the basis for a stationary SBGP equilibrium:

$$\begin{cases} \chi_{t+1} = \Delta z_t \chi_t^{\alpha} \omega_t^{1-\alpha-\gamma} \\ \omega_{t+1} = \Theta \eta_t \omega_t^{1-\phi}, \end{cases}$$
(27)

where

$$\Delta = \frac{\alpha \beta \overline{u}^{\gamma}}{b\left(1 - \overline{u} - \overline{v}\right)} \quad \text{and} \quad \Theta = \frac{\overline{v}^{\phi}}{b\left(1 - \overline{u} - \overline{v}\right)}.$$
(28)

Under our assumptions on the stochastic process governing the exogenous shocks, the nonlinear IFS associated to (27) is defined by the following three maps $f_i : \mathbb{R}^2 \to \mathbb{R}^2$:

$$\begin{cases} f_0(\chi,\omega) = \left(\Delta q_1 \chi^{\alpha} \omega^{1-\alpha-\gamma}, \Theta r \omega^{1-\phi}\right) & \text{with probability } p_0 \\ f_1(\chi,\omega) = \left(\Delta q_2 \chi^{\alpha} \omega^{1-\alpha-\gamma}, \Theta \omega^{1-\phi}\right) & \text{with probability } p_1 \\ f_2(\chi,\omega) = \left(\Delta \chi^{\alpha} \omega^{1-\alpha-\gamma}, \Theta \omega^{1-\phi}\right) & \text{with probability } p_2, \end{cases}$$
(29)

which has trapping set A defined by the rectangle in \mathbb{R}^2 containing all three fixed points of each map in (29), that is, the rectangle with opposite vertex having coordinates

$$(\chi_0^*, \omega_0^*) = \left(\left[\Delta q_1 \left(\Theta r \right)^{(1-\alpha-\gamma)/\phi} \right]^{1/(1-\alpha)}, \left(\Theta r \right)^{1/\phi} \right)$$

$$(\chi_2^*, \omega_2^*) = \left(\left[\Delta \Theta^{(1-\alpha-\gamma)/\phi} \right]^{1/(1-\alpha)}, \Theta^{1/\phi} \right);$$
(30)

such a scenario corresponds to growth rate values for the original variables, k_t , h_t and a_t , oscillating between finite positive bounds. If the IFS (29) converges to some invariant measure supported over (possibly a fractal subset of) A, then we have a stationary SBGP equilibrium.

Following the same approach pursued in the proof of Proposition 2, the next proposition shows that, for specific sets of values for parameters α , β , γ , ϕ , b, q_1 , q_2 , r and through a specific one-to-one logarithmic transformation, an affine system conjugate to (27) exists and defines a IFS converging to an invariant measure supported on a generalized Sierpinski gasket⁶ with vertices (0,0), (1/2,1) and (1,0).

Of course a different definition of the below one-to-one logarithmic transformation could lead to an alternative IFS system conjugate to (27): we do not investigate this issue here, but for future research we plan to explore the case of IFS defined only on complete (non-compact) spaces.

Proposition 5 Assume that $\phi \neq 1 - \alpha$ and parameters q_1, q_2 satisfy

$$\begin{array}{ll}
q_1 < q_2^2 & if \ \phi < 1 - \alpha, \\
q_1 > q_2^2 & if \ \phi > 1 - \alpha,
\end{array}$$
(31)

and let:

$$r = \left(q_1/q_2^2\right)^{\frac{1-\alpha-\phi}{1-\alpha-\gamma}}.$$
(32)

Then, the one-to-one logarithmic transformation $(\chi_t, \omega_t) \to (\varphi_t, \psi_t)$ defined by:

$$\varphi_t = \rho_1 \ln \chi_t + \rho_2 \ln \omega_t + \rho_3 \tag{33}$$

$$\psi_t = \rho_4 \ln \omega_t + \rho_5, \tag{34}$$

where:

$$\rho_1 = -\frac{1-\alpha}{2\ln q_2}, \qquad \rho_2 = \frac{(1-\alpha-\gamma)(1-\alpha)}{2(1-\alpha-\phi)\ln q_2}, \tag{35}$$

$$\rho_3 = 1 + \frac{1}{2\ln q_2} \left(\ln \Delta - \frac{1 - \alpha - \gamma}{1 - \alpha - \phi} \ln \Theta \right), \tag{36}$$

$$\rho_4 = \frac{(1-\alpha-\gamma)\phi}{(1-\alpha-\phi)}\ln\left(\frac{q_2^2}{q_1}\right), \qquad \rho_5 = 1 + \frac{(1-\alpha-\gamma)}{(1-\alpha-\phi)}\ln\left(\frac{q_1}{q_2^2}\right)\ln\Theta, \tag{37}$$

⁶By generalized Sierpinski gasket we mean a standard Sierpinski gasket whose pre-fractals are composed by triangles that may or may not overlap, depending on the parameters configuration.

defines a contractive affine IFS which is equivalent to the nonlinear dynamics in (27) composed of three maps $w_0, w_1, w_2 : \mathbb{R}^2 \to \mathbb{R}^2$, given by:

$$\begin{cases} w_0(\varphi,\psi) = (\alpha\varphi,(1-\phi)\psi) & \text{with probability } p_0\\ w_1(\varphi,\psi) = (\alpha\varphi+(1-\alpha)/2,(1-\phi)\psi+\phi) & \text{with probability } p_1\\ w_2(\varphi,\psi) = (\alpha\varphi+(1-\alpha),(1-\phi)\psi) & \text{with probability } p_2, \end{cases}$$
(38)

The IFS (38) converges weakly to a unique self-similar measure supported on a generalized Sierpinski gasket contained in the square $[0,1]^2$ and with vertices (0,0), (1/2,1) and (1,0).

The proof of Proposition 5 resembles closely that of Proposition 2 in La Torre et al. (2011); hence, we omit it.

By rewriting the IFS in (38) as

$$\begin{bmatrix} \varphi_{t+1} \\ \psi_{t+1} \end{bmatrix} = \begin{bmatrix} \alpha & 0 \\ 0 & 1-\phi \end{bmatrix} \begin{bmatrix} \varphi_t \\ \psi_t \end{bmatrix} + \begin{bmatrix} \zeta_t \\ \vartheta_t \end{bmatrix},$$
(39)

we can see that the (conjugate) random vector $(\zeta_t, \vartheta_t) \in \mathbb{R}^2$ taking on the three values (0, 0), $((1-\alpha)/2, \phi)$ and $(1-\alpha, 0)$ corresponds respectively to the three scenarios $(q_1, r), (q_2, 1)$ and (1,1) for the original random values (z_t,η_t) . Condition (32) relates the value r of the shock affecting the knowledge sector—which, under (31), is bound to be always strictly less than 1—to the two values q_1 and q_2 of the shock affecting the production sector. We believe this restriction not to be excessively restrictive as it leaves a certain degree of freedom in the choice of the values of two out of three exogenous shock parameters, being only the third parameter dependent upon the first two. Note that the condition $\phi \neq 1 - \alpha$ required in Proposition 5 precludes the possibility of generating the standard Sierpinski gasket with vertices (0,0), (1/2, 1) and (1, 0) through (38), as its construction postulates that $\phi = 1 - \alpha = 1/2$ must hold. Hence, the attractor of (38) must always be a generalized Sierpinski gasket, that is, a Sierpinski gasket-like set whose $prefractals^7$ are composed by triangles that do not overlap for smaller values of α , independently of values of ϕ , while they tend to overlap for larger values of α and smaller values of ϕ . For example, as for any pair of values $\alpha \geq 2/3$ and $\phi = 1 - \alpha$ the attractor becomes the full triangle of vertices (0,0), (1/2,1) and (1,0), there is always overlap whenever $\alpha \geq 2/3$ and $\phi < 1 - \alpha$ (see the examples in the next section).

The IFS in (38) is two-dimensional, mapping \mathbb{R}^2 into itself; therefore, in order to pursue singularity vs. absolute continuity analysis on its self-similar measure we should rely on Theorem 4 of Section 2. Unfortunately, this theorem can be applied only to affine IFS which are similitudes. As a matter of fact, the requirement that $\phi \neq 1-\alpha$ —which turns out to provide the key step in the proof of Proposition 5—in principle precludes any application of this Theorem, as the IFS (38) can be a similitude if and only if the contraction factors of the two components of each map w_i are the same; that is, if and only if $\alpha = 1 - \phi$. In other words, while on the one hand Theorem 4 is sufficiently general to encompass maps w_i in the IFS with different contraction factors [which is not the case with (38)], on the other hand the requirement that they are similitudes according to (3)—*i.e.*, they are defined through some orthogonal matrix Q_i as in (4)—implies that each map w_i must apply the same contraction factor to both variables, φ and ψ , as the contraction shrink must be the same along any direction in \mathbb{R}^2 .

However, by applying Theorem 5 of Section 2 to the IFS (39) with linear part given by

$$Q = \left[\begin{array}{cc} \alpha & 0 \\ 0 & 1 - \phi \end{array} \right],$$

⁷The sets obtained after each iteration of the map (38) are called *prefractals*.

it is possible to establish the following sufficient condition for the singularity of the invariant measure. As there are three maps in (38), we can set only two out of the three probabilities associated to each map w_i , say p_0 and p_1 , as the third must complement to 1: $p_2 = 1 - p_0 - p_1$. Let

$$E(p_0, p_1) = p_0^{p_0} p_1^{p_1} \left(1 - p_0 - p_1\right)^{1 - p_0 - p_1}$$
(40)

denote the (exponential of the) entropy of the Bernoulli process underlying the exogenous shocks in our model.

Proposition 6 If $\alpha (1 - \phi) < E(p_0, p_1)$, with $E(p_0, p_1)$ defined in (40), then the invariant measure μ^* of the IFS (38) is singular.

Figure 2 illustrates Proposition 6: any value for the product $\alpha (1 - \phi)$ below the entropy curve, *i.e.*, satisfying $0 < \alpha (1 - \phi) < E(p_0, p_1)$, characterizes a dynamic defined by the IFS (38) that weakly converges to a singular self-similar measure μ^* , which is supported on a generalized Sierpinski gasket contained in the square $[0, 1]^2$ and with vertices (0, 0), (1/2, 1) and (1, 0). Nothing can be said on the possible absolute continuity of μ^* when $\alpha (1 - \phi) \ge E(p_0, p_1)$. Note that when two out of the three probabilities p_i are both close to 0 the invariant measure μ^* turns out to be singular for most values of α and ϕ .



FIGURE 2: if $\alpha (1 - \phi) < E(p_0, p_1)$ on the triangle $\{(p_0, p_1) : p_0 \ge 0, p_1 \ge 0, p_0 + p_1 \le 1\}$, then the invariant measure μ^* of the IFS (38) is singular.

Consistently with Proposition 3, Proposition 6 states that singular invariant measures are associated with economies having small physical capital share in final good production (α); the novelty for the three-sector version is that the same effect can also be determined by a large human capital share in the knowledge sector producing (ϕ), while the human capital share in final production (γ), does not affect the nature of the invariant measure. Again, such result is independent of the size of the shocks q_1 and q_2 [and thus of r as given in (32)], the rate of time preference β and the human capital productivity coefficient b. Note that empirical estimates of the physical capital share are about 1/3 (Bernankeand Gurkaynak, 2002); thus, in view of Propositions 3 and 6, empirically relevant values of α can be considered 'small'. While clear empirical estimates of the human capital share in knowledge production do not exist, it is reasonable to believe the technological sector to be human capital intensive (Lucas, 1988), such that relevant values of the parameter ϕ should be considered 'large'. Both these results go in the same direction to ensure that the inequality in Proposition 6 is met, meaning that for an empirically realistic model's parametrization the invariant measure μ^* is likely to be singular.

Remark 2 If on the one hand the restriction $\phi \neq 1 - \alpha$ required by Proposition 5 seems harmless, as it allows the result to hold generally, on the other hand the special case $\phi = 1 - \alpha$ looks quite appealing for at least two (related) reasons: 1) it would allow the construction of 'symmetric' Sierpinski gaskets by means of three-sector endogenous growth models (as well as through two-sector, Ramsey-type growth models as in La Torre et al., 2011), and, more importantly, 2) it would render Theorem 4 available for constructing examples of absolutely continuous self-similar measures supported over the attractor.

As a matter of fact, as it can be easily grasped by looking at conditions (35)-(37), the restriction $\phi \neq 1 - \alpha$ is crucial in the proof of Proposition 5 (see also the proof of Proposition 2 in La Torre et al., 2011), and we could only do without it by setting $\gamma = \phi$; the latter position, however, when combined with our desired condition $\phi = 1 - \alpha$, would yield $\gamma = 1 - \alpha$, which, when substituted into (20) or into (27), implies the disappearance of the knowledge sector in our model. In other words, the only way to obtain a symmetric Sierpinski gasket as the attractor of a three-sector growth model is by eliminating one of the sectors, thus annihilating the very nature of the model itself, transforming it into the two-dimensional one discussed in Section 3, which, after detrending its two state variables k_t and h_t , exhibits a one-dimensional dynamic possibly converging to a Cantor set.

Apparently, the two logarithmic transformations defined by (33)-(34) in Proposition 5 become linearly dependent whenever the slopes of the affine maps in (38) are the same, that is, whenever $\alpha = 1 - \phi$ (i.e., $\phi = 1 - \alpha$), thus ruling out any chance of keeping the two-dimensional nature of the original dynamic (27) through the transformation, as it necessarily collapses into the one-dimensional dynamic (17) tackled in Section 3. Any attempt to obtain a symmetric Sierpinski gasket is doomed to fail from the start because symmetry itself kills one dimension of the problem.⁸ We speculate that this property may be more general: even by adding more sectors (dimensions) to the model, whenever one tries to get an affine IFS composed of similitudes out of the optimal dynamics all dimensions but one are bound to be lost, while being left only with a one-dimensional IFS, possibly converging to a Cantor set. We plan to further address this issue in future research.

Remark 3 As it happens for the IFS (15), also the nonlinear IFS (29) may be non-contractive, because there could exist a neighborhood of the smallest fixed point (χ_0^*, ω_0^*) defined in (30) on which at least one of the two components of the largest map $f_2(\chi, \omega)$ in (29) has (maximum) slope steeper than 1, which, in turn, implies that its Lipschitz constant is larger than 1, $\lambda_2 > 1$. Again, in such circumstances the results of Section 2 cannot be directly applied; nonetheless, Proposition 5 establishes indirectly the existence of a unique invariant measure for (29) even in such non-contractive cases. Because, for any i = 0, 1, 2, the nonlinear maps $f_i(\chi, \omega)$ in (29) have components which are strictly increasing and strictly concave, their Lipschitz constants λ_i

⁸We thank L. Gardini for raising this observation during the MDEF 2014 workshop.

can be computed as the largest eigenvalue of their Jacobian matrix evaluated at the smallest fixed point (χ_0^*, ω_0^*) defined in (30), which can be easily checked to be $\lambda_0 = \max \{\alpha, 1 - \phi\}, \lambda_1 = \max \{\alpha q_2/q_1, (1 - \phi)/r\}$ and $\lambda_2 = \max \{\alpha/q_1, (1 - \phi)/r\}$ respectively. Clearly, under our assumptions it may well occur that either $\alpha/q_1 > 1$ or $(1 - \phi)/r > 1$ hold; if this is the case, the nonlinear IFS (29) turns out to be non-contractive, but, as Proposition 5 still applies, such IFS converges anyway to a unique invariant measure.

For example, this is certainly the case when we set $\alpha = 1/3$, $\phi = \gamma = 0.2$, $q_1 = 0.2$, $q_2 = 0.8$ [so that the first inequalities in (31) hold and, according to (32), r = 0.313], $\beta = 0.96$ and b = 1.052 [so that there is sustained growth as $b > 1.042 = 1/(1 - \overline{u} - \overline{v})$], because $\lambda_2 = \max \{ \alpha/q_1, (1 - \phi)/r \} = \max \{ 1.667, 2.56 \} = 2.56 > 1$. Furthermore, if we set $p_0 = 1/3$ and take any positive value $0 < p_1 < 2/3$ so that $0 < p_2 = 2/3 - p_1 < 1$ holds al well, it turns out that the IFS (29) is not even 'contracting on average' according to (18). Indeed, noting that $\lambda_0 = \max \{ \alpha, 1 - \phi \} = \max \{ 1/3, 0.8 \} = 0.8$ and $\lambda_1 = \max \{ \alpha q_2/q_1, (1 - \phi)/r \} = \max \{ 1.333, 2.56 \} = 2.56$, we have

$$\sum_{i=0}^{2} p_i \ln \lambda_i = \frac{1}{3} \ln (0.8) + (p_1 + p_2) \ln (2.56) = \frac{1}{3} \ln (0.8) + \frac{2}{3} \ln (2.56) \simeq 0.552 > 0.552 = 0.552$$

5 Examples

Below we present some examples of attractors for certain parameterizations of the model described in Section 4. In all examples we keep constant the parameter $\beta = 0.96$ and set $b = 1/(1 - \overline{u} - \overline{v}) + 0.01$, where \overline{u} and \overline{v} are defined in (22) and (23) respectively, so to have always sustained growth. In the first four examples we set $\gamma = \phi$, $q_1 = 0.2$, $q_2 = 0.6$, while parameter r is defined according to (32). Note that the restriction $\gamma = \phi$ implies that the exponent in (32) equals to 1, so that $r = q_1/q_2^2 \equiv 0.556$ and $q_1 = 0.2 < 0.36 = q_2^2$ holds, which, in turn, implies that we must choose values for the key parameters α, ϕ satisfying the first inequality in (31), *i.e.*, $\phi < 1 - \alpha$ must hold. In the first three examples we also link parameter ϕ to our choices of parameter α according to

$$\phi = 1 - \alpha - 0.001, \tag{41}$$

so that, while on the one hand the first inequality in (31) is satisfied, on the other hand we focus on almost symmetrical generalized Sierpinski gaskets.⁹

Figure 3(a) shows a good estimate of the generalized Sierpinski gasket, which is the attractor of the affine IFS (38), obtained after 8 iterations of the operator (2) associated with it,¹⁰ starting from the triangle of vertices (0,0), (1/2, 1) and (1,0), for the benchmark case $\alpha = 1/2$, to which, from (41), corresponds $\phi = 0.499$. Figure 3(b) reports an estimate of a distorted nonlinear generalized Sierpinski gasket obtained after 8 iterations of the operator (2) associated to the nonlinear IFS (29), which, according to Proposition 5, corresponds to the attractor in Figure 3(a) for the original nonlinear optimal dynamic described in (27). Recalling (28), the three

⁹Recall that the condition $\phi \neq 1 - \alpha$ in Proposition 5 rules out symmetrical Sierpinski gaskets as attractors of IFS (38).

¹⁰The Maple 18 code generating all attractor approximations in this section is available from the authors upon request.

fixed points in the latter case are computed as

$$\begin{aligned} (\chi_0^*, \omega_0^*) &= \left(\left[\Delta q_1 \left(\Theta r \right)^{(1-\alpha-\gamma)/\phi} \right]^{1/(1-\alpha)}, \left(\Theta r \right)^{1/\phi} \right) = (0.00036, 0.00002) \\ (\chi_1^*, \omega_1^*) &= \left(\left[\Delta q_2 \left(\Theta r \right)^{(1-\alpha-\gamma)/\phi} \right]^{1/(1-\alpha)}, \Theta^{1/\phi} \right) = (0.00321, 0.00007) \\ (\chi_2^*, \omega_2^*) &= \left(\left[\Delta \Theta^{(1-\alpha-\gamma)/\phi} \right]^{1/(1-\alpha)}, \Theta^{1/\phi} \right) = (0.00891, 0.00007) . \end{aligned}$$

As $\alpha (1 - \phi) = 0.2505 < 1/3 = \min \{ E(p_0, p_1) : p_0 \ge 0, p_1 \ge 0, p_0 + p_1 \le 1 \} = E(1/3, 1/3),$ according to Proposition 6 the invariant measure μ^* supported over the attractor in Figure 3(a) must be singular for any choice of probabilities p_0, p_1 .



FIGURE 3: (a) first 8 iterations of the IFS (38) for $\alpha = 0.5$, $\phi = 0.499$, and (b) its corresponding distorted nonlinear counterpart obtained after 8 iterations of the IFS (29).

Figure 4(a) shows the estimate of the attractor of the affine IFS (38), obtained through the same construction as before, for $\alpha = 0.4$, to which, according to (41), corresponds $\phi = 0.599$. Figure 4(b) reports an estimate of the distorted nonlinear generalized Sierpinski gasket which is the attractor of the corresponding nonlinear IFS (29). By (28), the three fixed points in the latter case now are

$$(\chi_0^*, \omega_0^*) = (0.00054, 0.00002), \quad (\chi_1^*, \omega_1^*) = (0.00337, 0.00006), \quad (\chi_2^*, \omega_2^*) = (0.00789, 0.00006).$$

Again, as $\alpha (1 - \phi) = 0.1604 < 1/3$, the invariant measure μ^* supported over the generalized Sierpinski gasket in Figure 4(a) is singular for any choice of probabilities p_0, p_1 . This property is clearly confirmed by the strong no-overlapping of the prefractals in Figures 4(a) and 4(b).

Figure 5(a) shows the estimate of the attractor of the affine IFS (38), again obtained through the same construction as before, for $\alpha = 0.632$, to which, according to (41), corresponds $\phi = 0.367$. Figure 5(b) reports the attractor of the corresponding nonlinear IFS (29). By (28), the fixed points in this case are

$$(\chi_0^*, \omega_0^*) = (0.00012, 0.00002), \quad (\chi_1^*, \omega_1^*) = (0.00245, 0.00010), \quad (\chi_2^*, \omega_2^*) = (0.00984, 0.00010).$$



FIGURE 4: (a) first 8 iterations of the IFS (38) for $\alpha = 0.4$, $\phi = 0.599$, and (b) its corresponding distorted nonlinear counterpart obtained after 8 iterations of the IFS (29).



FIGURE 5: (a) first 8 iterations of the IFS (38) for $\alpha = 0.632$, $\phi = 0.367$, and (b) its corresponding distorted nonlinear counterpart obtained after 8 iterations of the IFS (29).

Because now $\alpha (1 - \phi) = 0.4001 > 1/3$, according to Proposition 6 we cannot exclude that the invariant measure μ^* supported over the attractor in Figure 5(a) may be absolutely continuous for some values of probabilities p_0, p_1 , possibly for $p_0 = p_1 = 1/3$ or values around 1/3. In fact, both generalized Sierpinski gaskets in Figure 5 exhibit enough overlapping of the prefractals to allow for positive 2-dimensional Lebesgue measure of the attractor itself and an absolutely continuous invariant measure supported on it. From Figure 5 it is understood that for larger values of α and smaller values of ϕ the extent of the overlapping in the prefrectals increases. Indeed, for $\alpha = 2/3$ and $\phi = 1 - \alpha = 1/3$ the attractor of (38) becomes the full triangle

of vertices (0,0), (1/2,1) and (1,0), with Lebesgue measure equals to 1/2; unfortunately, as $\phi = 1 - \alpha$ is ruled out in the assumptions of Proposition 5, we cannot relate this scenario to the nonlinear IFS (29).

In Figure 6 we relax condition (41) and consider the values $\alpha = \phi = 1/3$, still satisfying the first inequality in (31), $\phi < 1 - \alpha$, so that no changes are required for the other parameters. The fixed points of the distorted nonlinear generalized Sierpinski gasket in Figure 6(b) are

 $(\chi_0^*,\omega_0^*) = (0.00013, 0.00314), \quad (\chi_1^*,\omega_1^*) = (0.00163, 0.01829), \quad (\chi_2^*,\omega_2^*) = (0.00351, 0.01829).$

As $\alpha (1 - \phi) = 2/9 < 1/3$, also in this case the invariant measure μ^* supported over the attractor in Figure 6(a) is singular for any choice of probabilities p_0, p_1 . Again this property is confirmed by the no-overlapping of the prefractals.



FIGURE 6: (a) first 8 iterations of the IFS (38) for $\alpha = \phi = 1/3$, and (b) its corresponding distorted nonlinear counterpart obtained after 8 iterations of the IFS (29).

Finally, in Figure 7 we consider an example with larger values of both α and ϕ in which the second inequality in (31) holds: $\alpha = 0.7$ and $\phi = 0.5$, clearly satisfying $\phi > 1 - \alpha$. Because under the assumptions in the model and from Proposition 5 both $q_1 < q_2$ and $q_1 > q_2^2$ must be satisfied, we set $q_1 = 0.4$ and $q_2 = 0.6$, so that, according to (32), r = 0.81. As also $\gamma < 1 - \alpha$ must hold, we set $\gamma = 0.2$. The fixed points of the distorted nonlinear generalized Sierpinski gasket in Figure 7(b) are

$$(\chi_0^*, \omega_0^*) = (0.00022, 0.00813), \quad (\chi_1^*, \omega_1^*) = (0.00099, 0.01239), \quad (\chi_2^*, \omega_2^*) = (0.00541, 0.01239).$$

In this case $\alpha (1 - \phi) = 0.35 > 1/3$, so that, according to Proposition 6, an absolutely continuous invariant measure μ^* supported over the attractor in Figure 7(a) cannot be ruled out for some values of probabilities p_0, p_1 , possibly for $p_0 = p_1 = 1/3$. As a matter of fact, both generalized Sierpinski gaskets in Figure 7 exhibit enough overlapping of the prefractals so that the attractor may have positive Lebesgue measure, possibly with an absolutely continuous invariant measure supported on it.



FIGURE 7: (a) first 8 iterations of the IFS (38) for $\alpha = 0.7$, $\phi = 0.5$, $q_1 = 0.4$, $q_2 = 0.6$, r = 0.81 and $\gamma = 0.2$, and (b) its corresponding distorted nonlinear counterpart obtained after 8 iterations of the IFS (29).

6 Conclusions

In this paper we extend the analysis of stochastic discrete-time optimal growth models to consider the multi-sectorial framework in the context of sustained growth. We consider first the simplest case, namely the Lucas-Uzawa (1988) two-sector model, and then an extended three sector model, as in La Torre and Marsiglio (2010). Both the models exhibit two peculiar features: the log-Cobb-Douglas structure of preferences and production functions in each sector allows for a closed form solution of the Bellman equation, thus permitting to explicitly compute the optimal dynamics of the state variables; moreover, through simple log-transformations of the capital ratio variable dynamics we are able to show that for a sufficiently rich set of parameters' configurations the model economy converges to an invariant measure supported on some fractal set (a generalized Cantor set in the case of the two sector model, and a generalized Sierpinski gasket in the case of the three sector model).

By exploiting some recent results on the IFS theory, we are also able to establish some sufficient conditions under which the invariant measure turns out to be either singular or absolute continuous with respect to Lebesgue measure (for the three-sector model we only consider the singularity with respect to Lebesgue measure). By comparing the outcomes of the two and three sector model, it is clear that the latter framework is much richer and allows for a greater variety of alternative configurations in terms of singularity and absolute continuity of the selfsimilar measure associated to the IFS. We are also able to show that, in both models, for a rich range of parameter value (an also for an empirically realistic model's parametrization) the invariant measure is likely to be singular.

This paper significantly extends the literature on stochastic growth and fractal attractors, by showing that also stochastic BGP equilibria can have a fractal nature and showing that for empirically relevant parameterizations the invariant measure is singular. Despite the new insights provided by these results, new questions for future research naturally arise. In particular, since in Section 4 we are able only to comment on the singularity of the invariant measure, the issue related to its eventual absolute continuity is still unsolved; it would be interesting at least to build an example converging to some absolutely continuous invariant measure. Moreover, since the literature has only focused thus far on log-linear transformations of the optimal dynamics associated with stochastic growth models (the same approach we use in our paper), it might be interesting to try characterizing the singularity vs. absolute continuity of the invariant measure directly for the original nonlinear IFS—*i.e.*, without transforming them into affine IFS—in both two and three-sector models. These issues are left for future research.

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