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From Mind to Market: A Global, Dynamic Analysis of R&D*

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Abstract

Existing models of R&D are not easily reconciled with four observable aspects of R&D: initial technologies ("ideas") need to be developed further, only a minority of initial ideas is successfully brought to the market, production and process innovations take place simultaneously (whereby, initially, there is no production at all), and process innovations are implemented for technologies that are destined to leave the market. We present a detailed bifurcation analysis for a dynamic model of R&D that captures these observations in one, unifying framework. As we provide a global analysis, we do not limit initial technologies to carry marginal costs that are below the choke price. We show that there always exists a critical value of initial marginal cost above which the firm does not initiate any (R&D) activity; the path to the saddle-point steady state is never globally optimal. We also sketch some tentative policy implications of our analysis.

Keywords: Bifurcations, Innovation, Monopoly, Optimal control, R&D

JEL: D42, L12, O31

1 Introduction

Around 1900, Swiss chemist Jacques E. Brandenberger invented a transparent film

which he named "cellophane". On December 26, 1923, American chemical company

DuPont acquired from La Cellophane Société Anonyme, an organization to which

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Brandenberger had assigned his various patents, the exclusive rights to its United States cellophane patents. DuPont's hopes for creating a lucrative market were shattered, however, by discovering that cellophane could not be used to wrap up products that require moisture proofing, such as cake and candy. It took DuPont chemist William H. Charch three years and thousands of tests to develop a lacquer that made cellophane moisture proof, an invention that revolutionized the packaging and merchandizing industry. As the manufacturing costs of cellophane continued to decline due to DuPont's ongoing process innovations, so did the price of cellophane. All this contributed to cellophane being used as wrapping material for a variety of products (from food to jewelry), and for its use in various products (including batteries, scotch tape, and dialysis machines). Since the mid-1930's, cellophane has been manufactured continuously. It is still used today.

The development of cellophane, from mind to market, is typical for the life cycle of many new technologies. Research starts long before a prototype sees the light of day; development begins long before the launch of a new product. Ideas abound, but only a small fraction of these ideas leads to successful innovations. For instance, in 1979, Apple was enthusiastic about a design for a computer mouse that they had discovered at Xerox research center. But Apple did not develop this prototype further as the projected production costs of a single mouse would be over \$400.¹ Also, existing technologies tend to leave markets slowly due to incumbents' R&D efforts. For example, Edison's invention of an electric light bulb was bound to replace the gas lamp. Producers of the latter however prolonged the lifetime of this inferior technology through continuous product innovations, including the introduction of the Welsbach mantle that made gas lamps five times more efficient (Utterback, 1994).

These examples illustrate four stylized facts of R&D: (i) initial technologies ("ideas" or "prototypes") need to be developed before they are suitable for large-scale production; (ii) there are many initial technologies, but only a very limited fraction is developed further, and from this fraction only a subset will constitute a successful innovation, (iii) production and the search for further improvements take place simultaneously, and (iv)

¹Years later, Apple came up with a new design which would only cost \$25 to produce. This prototype was subsequently developed into Apple's famous single-button mouse.

there are process innovations for technologies that are destined to leave the market in the foreseeable future.

Existing models of R&D are not easily reconciled with these observations. Static models of R&D are silent about the process from prototype to the first release of new products and production technologies.² Moreover, these models are forced to assume that marginal costs, the proxy for production technologies, are below the choke price (the lowest price at which the quantity sold is zero). This assumption is unlikely to hold for new technologies in their early stages of development.

Dynamic models of R&D are in principle equipped to capture the path from prototype to market penetration. To date, however, neither "innovation race" models nor dynamic versions of static models do so adequately. In essence, models of innovation races examine the time of completion of a cost-saving innovation of known magnitude, whereby the expected time of completion has a one-to-one relationship with R&D expenditures.³ These models exclude the coexistence of production and R&D efforts. Moreover, the R&D process cannot fail, thus transforming the R&D investment decision into a static one. Dynamic versions of static R&D models maintain the assumption that marginal costs are below the choke price at all times.⁴ Initial technologies that are "expensive" are excluded, by assumption, from the analysis. Hence, R&D efforts always lead to the stable equilibrium. Put differently, every initial technology will be developed further, and successfully so. Indeed, without exception this literature provides analyses that are only locally optimal.

In this paper, we provide a general description of the economics of R&D. In particular, for any initial technology level c_0 , we characterize the investment and production path that is globally optimal for a monopolist. Hence, we do not exclude initial technologies with marginal costs above the choke price, and we do not restrict ourselves to solutions that are only locally optimal.⁵ Our model is, therefore, more easily reconciled

²The seminal papers in this literature are d'Aspremont and Jacquemin (1988) and Kamien *et al.* (1992); see De Bondt (1997) for an overview.

³Seminal contributions here include Loury (1979) and Lee and Wilde (1980). Reinganum (1989) surveys this literature.

⁴This literature is still scant; it includes Cellini and Lambertini (2009), Lambertini and Mantovani (2009), and Kováč *et al.* (2010).

⁵As it turns out, a non-trivial part of the parameter space yields stable paths with initial marginal costs so high that production would yield negative mark-ups. At that stage all the monopolist does is to invest in R&D

with the four stylized facts of R&D mentioned earlier.⁶

From a technical point of view, the problem of the monopolist is formulated as an optimal control problem with an infinite time horizon. Its distinctive characteristic is the presence of multiple equilibria while at the same time the Arrow-Mangasarian sufficiency conditions are not met. In models of this type, the qualitative properties of optimal solutions may change if parameters are varied. We therefore use bifurcation analysis⁷, which constitutes the mathematics of a qualitative change, to assess industry dynamics for all initial technologies in conjunction with other parameter combinations (including time preferences and the efficiency of the R&D process).

Our analysis yields four distinct scenarios, three of which remain hidden in local analyses: (i) initial marginal costs are above the choke price and the R&D process is initiated; after some time production starts and marginal costs fall with subsequent R&D investments; (ii) initial marginal costs are above the choke price and the R&D process is not initiated, yielding no production at all; (iii) initial marginal costs are below the choke price and the R&D process is initiated; production starts immediately and marginal costs fall over time, and (iv) initial marginal costs are below the choke price and the R&D process is progressively scaled down; production starts immediately but the technology (and production) dies out over time; the firm leaves the market. The strength of our analysis is that all these cases can emerge from the same unifying framework.

Our analysis is not without policy implications. It shows that market characteristics that affect future profitability have an impact on the monopolist's decision to further develop an initial technology. If market regulations are such that future mark-ups are reduced *a priori*, it could be that the range of initial technologies that is developed further will shrink. The loss of total surplus this reduction brings about constitutes a hidden cost of market regulations if the initial technology consists of marginal costs above the choke price. In that case, there is no immediate production lost if the initial

in order to bring down the costs of production.

⁶As this is the first analysis ever to consider the entire range of initial technologies, we restrict ourselves to the monopoly case. Obviously, competitive forces play an important role as well in what we observe about R&D.

⁷Variations in parameter values can lead to qualitative changes in the solution structure (e.g., some steady states lose their stability, indifference points appear, and so on). Such qualitative changes in the solution structure due to smooth variations in parameters are called *bifurcations* (see Grass *et al.*, 2008).

technology is not developed further. Yet, these indirect costs should be taken into account when assessing properly the trade-off between static and dynamic efficiency.

The rest of the paper is organized as follows. The model is introduced in the next section and its steady states are characterized in Section 3. The properties of the global equilibrium and corresponding industry dynamics are considered in Section 4. Section 5 concludes.

2 The model

Time t is continuous: $t \in [0, \infty)$. A single supplier produces at marginal costs c(t). At every instant, market demand equals production

$$q(t) = A - p(t), \tag{1}$$

where p(t) is the market price, q(t) is the quantity produced, and A is the choke price. The monopolist starts with an initial, exogenous technology $c(0) = c_0$. He can reduce this marginal cost by exerting R&D effort k(t). Marginal cost evolve over time:⁸

$$\frac{dc(t)}{dt} \equiv \dot{c}(t) = c(t) \left(-k(t) + \delta\right),\tag{2}$$

⁸One interpretation of the model is that c_0 constitutes the initial level of technology corresponding to a product innovation which is then followed by subsequent process innovations. This idea can be traced back to the pioneering work by Abernathy and Utterback (1975, 1978). Lambertini and Mantovani (2009) consider a multiproduct monopolist that simultaneously pursues process and product innovations. They show that the path to the saddle-point steady state involves some substitutability between the two innovation types.

where $\delta > 0$ is the constant rate of efficiency reduction due to the ageing of technology and organizational forgetting.⁹ The cost of R&D efforts per unit of time is given by:

$$\Gamma(k(t)) = bk(t)^2, \tag{3}$$

where b > 0 is inversely related to the cost-efficiency of the R&D process. Hence, the R&D process exhibits decreasing returns to scale.¹⁰ The monopolist discounts the future with the constant rate $\rho > 0$.

Instantaneous profit is:

$$\pi(q,k,c) = (A - q - c)q - bk^2,$$
(4)

yielding total discounted profit:

$$\Pi(q,k,c) = \int_0^\infty \pi(q,k,c)e^{-\rho t}dt.$$
(5)

The optimal control problem for the monopolist is to find controls q^* and k^* that maximize the profit functional Π subject to the state equation (2), the initial condition

⁹In assuming a positive rate of technology (or know-how) depreciation we follow the related theoretical literature (Besanko et al. (2010), Lambertini and Mantovani, 2009). For instance, in Besanko et al. (2010), a firm moves down its learning curve if its gains in know-how through learning are less than the inevitable knowhow losses from organizational forgetting, inducing marginal costs to increase. Empirical documentation also supports the assumption of a positive depreciation rate due to organizational forgetting (Benkard, 2004): "an aircraft producer's stock of production experience is constantly being eroded by turnover, lay offs and simple losses of proficiency at seldom repeated tasks. When producers cut back output, this erosion can even outpace learning, causing the stock of experience to decrease" (Benkard, 2004, p. 590). In our model, know-how gains come from investments in R&D and not from production, but the main idea is the same. Furthermore, production costs include complementary inputs that are typically purchased. Due to their inherent evolution over time, it becomes ever more costly to incorporate them into the production process of a firm that is sluggish in its R&D (e.g., installing a newer version of software on obsolete hardware can significantly slow down performance). The reason for this is that R&D efforts not only generate new information within the firm, but also determine the firm's 'absorptive capacity' - "the firm's ability to identify, assimilate, and exploit knowledge from the environment" (Cohen and Levinthal, 1989, p. 569). Note that the equilibria for a non-positive depreciation rate follow trivially. For δ is zero, consider δ very close to zero. Such a small depreciation rate pushes the level of the initial technology that will not be developed further far beyond the choke price as only minuscule investments are needed to reduce marginal costs over time. Indeed, for negative δ , every initial technology will be developed further as there is an exogenous reduction in marginal costs over time, yielding a steady state with no production costs.

¹⁰Again, we follow the literature with this assumption (see, e.g., d'Aspremont and Jacquemin (1988), Kamien *et al.* (1992), or Qiu (1997)). The evidence is mixed however. Schwartzman (1976) for instance documents significant economies of scale in R&D. But Madsen (2007) is unable to reject the hypothesis of constant returns to scale in the number of patents applied for. Moreover, for the pharmaceutical industry, Vernon and Gusen (1974) and Graves and Langowitz (1993) concludes that firms experience decreasing returns to scale in R&D. And Adams and Griliches (1996) conclude that there are strong diseconomies of scale in the production of scientific articles.

 $c(0) = c_0$, and two control constraints which must hold at all times: $q \ge 0$ and $k \ge 0$. Note that according to (2), if $c_0 > 0$, then c(t) > 0 for all t. The set of all possible states at each time t is given by $c \in [0, \infty)$.

The model has four parameters: A, b, δ , and ρ . The analysis can be simplified by considering a rescaled version of the model which, as defined in Lemma 1, carries only two parameters: ϕ and $\tilde{\rho}$. That is, rescaling the model translates the four-dimensional parameter space into a two dimensional space.

Lemma 1. By choosing the units of t, q, c, and k appropriately, we can assume A = 1, b = 1, and $\delta = 1$. This yields the following, rescaled model:

$$\tilde{\pi}(\tilde{q}, \tilde{k}, \tilde{c}) = (1 - \tilde{q} - \tilde{c})\tilde{q} - \tilde{k}^2,$$
(6)

$$\tilde{\Pi}(\tilde{q}, \tilde{k}, \tilde{c}) = \int_0^\infty \tilde{\pi}(\tilde{q}, \tilde{k}, \tilde{c}) e^{-\tilde{\rho}\tilde{t}} d\tilde{t}$$
(7)

$$\dot{\tilde{c}} = \tilde{c} \left(1 - \phi \tilde{k} \right), \quad \tilde{c}(0) = \tilde{c}_0, \quad \tilde{c} \in [0, \infty) \,\forall \, \tilde{t} \in [0, \infty) \tag{8}$$

$$\tilde{q} \ge 0, \quad \tilde{k} \ge 0 \tag{9}$$

$$\tilde{\rho} > 0, \quad \phi > 0 \tag{10}$$

with conversion rules: $q = A\tilde{q}, c = A\tilde{c}, k = \frac{A}{\sqrt{b}}\tilde{k}, t = \frac{\tilde{t}}{\delta}, \pi = A^2\tilde{\pi}, \phi = \frac{A}{\delta\sqrt{b}}, \rho = \tilde{\rho}\delta$. *Proof.* See Appendix A.

The rescaled model introduces a new parameter $\phi = A/(\delta\sqrt{b})$, which captures the profit potential of a technology in hand: a higher (lower) A implies higher (lower) potential sales revenue, a higher (lower) b implies that each unit of R&D effort costs more (less), whereas a higher (lower) δ implies that each unit of R&D effort decreases the marginal cost by less (more). Therefore, a higher (lower) ϕ corresponds to a higher (lower) profit potential of a technology. For notational convenience, we henceforth omit tildes.

To solve the dynamic optimization problem, we introduce the current-value Pontrya-

gin function¹¹

$$P(c, q, k, \lambda) = (1 - q - c) q - k^{2} + \lambda c (1 - \phi k), \qquad (11)$$

where λ is the current-value co-state variable. It measures the marginal worth of the increment in the state c at time t when moving along the optimal path.¹²As the marginal cost is a "bad", we expect $\lambda(t) \leq 0$ along optimal trajectories.

Pontryagin's Maximum Principle states that if the triple (c^*, q^*, k^*) is an optimal solution, then there exists a function $\lambda(t)$ such that c^*, q^*, k^* , and λ satisfy the following conditions:

1.) q^* and k^* maximize the function P for each t:

$$P(c^*, q^*, k^*, \lambda) = \max_{(q,k) \in \mathbb{R}^2_+} P(c^*, q, k, \lambda),$$
(12)

2.) λ is a solution to the following co-state equation:

$$-\frac{\partial P}{\partial c} = \dot{\lambda} - \rho \lambda \quad \Leftrightarrow \quad \dot{\lambda} = q + (\rho - 1 + \phi k) \lambda, \tag{13}$$

which is evaluated along with the equation for the marginal cost

$$\dot{c} = c(1 - \phi k) \tag{14}$$

and the initial condition $c(0) = c_0$.

Let $q = Q(c, \lambda)$ and $k = K(c, \lambda)$ solve the problem $\max_{(q,k)} P(c, q, k, \lambda)$ for every (c, λ) . We define the current-value Hamilton function

$$H(c,\lambda) = P(c,Q(c,\lambda),K(c,\lambda),\lambda).$$
(15)

The above necessary conditions (12)–(13) for an optimal solution are complemented by

¹¹Also called pre-Hamilton or un-maximized Hamilton function.

¹²More precisely, it measures the marginal worth of the increment in c along all trajectories that satisfy the necessary conditions for an optimal solution (specified in what follows), including Michel's transversality condition (16).

two transversality conditions:13

$$\lim_{t \to \infty} e^{-\rho t} H(c, \lambda) = 0$$
(16)

and

$$\lim_{t \to \infty} e^{-\rho t} \lambda c = 0.$$
⁽¹⁷⁾

If problem (12) has a solution, this solution necessarily satisfies the following Karush-Kuhn-Tucker conditions for all t:

$$\frac{\partial P}{\partial q} = 1 - 2q - c \le 0, \qquad q \frac{\partial P}{\partial q} = 0,$$
 (18)

$$\frac{\partial P}{\partial k} = -2k - \phi \lambda c \le 0, \qquad k \frac{\partial P}{\partial k} = 0.$$
(19)

Conditions (18) imply that either i) q = 0 and $c \ge 1$ or ii) q > 0 and c < 1.¹⁴ In particular: .

$$q^* = \begin{cases} (1-c)/2 & \text{if } c < 1, \\ 0 & \text{if } c \ge 1. \end{cases}$$
(20)

Conditions (19) imply that either i) k = 0 and $\frac{\partial P}{\partial k} \leq 0$ (implying $\lambda \geq 0$) or ii) k > 0and $\frac{\partial P}{\partial k} = 0$. In particular:

$$k^* = \begin{cases} -\frac{\phi}{2}\lambda c & \text{if } \lambda \le 0, \\ 0 & \text{if } \lambda > 0. \end{cases}$$
(21)

The above conditions for optimal production yields two regimes of the state-control system. The first is characterized by positive production; in the second there is no production.

¹³The necessity of (16), which can be restated as $\lim_{t\to\infty} e^{-\rho t} \Pi(c, \lambda) = 0$ and allows exclusion of non-optimal trajectories, was proven by Michel (1982). Kamihigashi (2001) proves the necessity of (17). ¹⁴Observe that in the non-rescaled model, the analogous condition for positive production is that

c(t) < A.

2.1 The system with positive production

In the region where c < 1, using (13), (14), (20), and (21), we obtain the following optimal state and co-state dynamics:

$$\begin{cases} \dot{c} = c \left(1 + \frac{1}{2} \phi^2 \lambda c \right), \\ \dot{\lambda} = \frac{1}{2} (1 - c) + \left(\rho - 1 - \frac{1}{2} \phi^2 \lambda c \right) \lambda, \end{cases}$$
(22)

where $\lambda \leq 0.15$ Equation (21) depends on the co-state and state variable. Note that the correspondence between k and λ is one-to-one if, and only if, $\lambda \leq 0$, and that therefore the system can be re-written in the state-control form.

Differentiate equation (21) with respect to time to obtain the dynamic equation

$$\frac{dk}{dt} \equiv \dot{k} = -\frac{\phi}{2} \left(\dot{\lambda c} + \lambda \dot{c} \right). \tag{23}$$

Use (13) and (14) for $\dot{\lambda}$ and \dot{c} respectively. Note that equation (21) also implies that

$$\lambda = -\frac{2k}{\phi c}.$$
(24)

Substitute this expression into (23), together with (22) and the expression for the optimal output level (q = (1 - c)/2), to obtain:

$$\dot{k} = \rho k - \frac{\phi}{4}c(1-c).$$
 (25)

Hence, the system with positive production (c < 1) in the state-control form consists of the following two differential equations:

$$\begin{cases} \dot{k} = \rho k - \frac{\phi}{4}c(1-c), \\ \dot{c} = c\left(1 - \phi k\right). \end{cases}$$
(26)

¹⁵We show in Appendix F that optimal solution trajectories necessarily satisfy $\lambda \leq 0$ for all $t \in [0, \infty)$. Therefore, the differential equations in the main text, valid for $\lambda \leq 0$, are all one has to consider.

2.2 The system with zero production

In the region $c \ge 1$, substituting (13), (14), and (24) into (23) and imposing q = 0, gives the following expression:

$$\dot{k} = \rho k. \tag{27}$$

Hence, the state-control system with zero production consists of the following two differential equations:

$$\begin{cases} \dot{k} = \rho k, \\ \dot{c} = c \left(1 - \phi k \right). \end{cases}$$
(28)

The state-co-state analogue is:

$$\begin{cases} \dot{c} = c \left(1 + \frac{1}{2} \phi^2 \lambda c \right), \\ \dot{\lambda} = \left(\rho - 1 - \frac{1}{2} \phi^2 \lambda c \right) \lambda. \end{cases}$$
(29)

The Hamilton function, obtained by substituting (21) and appropriate expressions for q from (20) into the Pontryagin function (11), is given by

$$H(c,\lambda) = \begin{cases} \frac{1}{4}(1-c)^2 + \frac{1}{4}\phi^2\lambda^2c^2 + \lambda c & \text{if } c \in [0,1), \\ \frac{1}{4}\phi^2\lambda^2c^2 + \lambda c & \text{if } c \in [1,\infty), \end{cases}$$
(30)

where $\lambda \leq 0$. The state-control analogue is then

$$H(c,k) = \begin{cases} \frac{1}{4}(1-c)^2 + k\left(k - \frac{2}{\phi}\right) & \text{if } c \in [0,1), \\ k\left(k - \frac{2}{\phi}\right) & \text{if } c \in [1,\infty), \end{cases}$$
(31)

where $k \ge 0$.

3 Steady-state solutions

The steady-state solutions of (26) and (28) are obtained by imposing the stationarity conditions $\dot{k} = 0$ and $\dot{c} = 0$.

Lemma 2. The "no-production" state-control system (28) has no steady state in the region where it is defined.

Proof. See Appendix B.

The result of the above lemma is intuitive as a steady state in the "no-production" region would imply that a firm maintains positive R&D investments without ever selling anything.

Consider now system (26). Imposing the stationarity condition $\dot{k} = 0$, we obtain

$$k^{M} = \frac{\phi}{4\rho}c(1-c) \ge 0, \qquad \forall c \in [0,1],$$
 (32)

where superscript *M* stands for *Monopoly*.¹⁶ Steady-state marginal cost follows from inserting (32) into the state dynamics (14) and imposing the stationarity condition $\dot{c} = 0$:

$$\dot{c} = c \left(1 - \frac{\phi^2}{4\rho} c(1-c) \right) = 0.$$
 (33)

This yields:

$$c^M = 0, \quad c^M = \frac{1}{2} \pm V,$$
 (34)

where

$$V = \frac{1}{2}\sqrt{1 - \frac{16\rho}{\phi^2}}.$$
 (35)

Observe that V is real if, and only if, $\phi \ge 4\sqrt{\rho}$, in which case $V \in [0, \frac{1}{2})$ and $c^M \in [0, 1)$. The solution to the system is summarized in Proposition 1 and is depicted in Figure 1.

Proposition 1. If $\phi > 4\sqrt{\rho}$, the state-control system with positive production (26) has three steady states:

i) $(c^M, k^M) = (0, 0)$ is an unstable node,

 $^{^{16}\}mathrm{All}$ steady-state values have superscript M.

ii) $(c^M, k^M) = \left(\frac{1}{2} + V, \frac{1}{\phi}\right)$ is either an unstable node or an unstable focus, and

iii) $(c^M, k^M) = \left(\frac{1}{2} - V, \frac{1}{\phi}\right)$ is a saddle-point steady state.

At $\phi = 4\sqrt{\rho}$, a so-called "saddle-node bifurcation" occurs as the last two steady states coincide and form a so-called "semi-stable" steady state.

If $\phi < 4\sqrt{\rho}$, the state-control system with positive production has one single steady state: the origin $(c^M, k^M) = (0, 0)$, which is unstable.



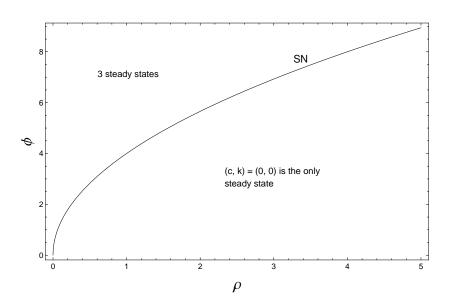


Figure 1: Steady states of the state-control system. The saddle-node bifurcation (SN) curve is defined by $\phi = 4\sqrt{\rho}$. Passing through this curve from above, the two steady states other than origin coincide, form a "semi-stable" steady state, and then disappear, leaving only one steady state to exist.

The intuition for the condition $\phi > 4\sqrt{\rho}$ in the above proposition is that developing a technology further becomes an option (the saddle-point steady state exists) only if for a given discount rate, the profit potential of a considered technology is high enough.

In the saddle-point steady state, instantaneous marginal cost, output, and profits are,

respectively:

$$c^M = \frac{1}{2} - V, \tag{36}$$

$$q^{M} = \frac{1}{4} + \frac{V}{2},\tag{37}$$

$$\pi^{M} = \frac{1}{16} \left(1 + 2V \right)^{2} - \frac{1}{\phi^{2}}.$$
(38)

3.1 Comparative statics

The relevant comparative statics are as follows:

Lemma 3. Steady-state marginal cost is increasing in b, δ , and ρ , and decreasing in ϕ ; steady-state quantity, profit, consumers' surplus, and total surplus are decreasing in b, δ , and ρ , and increasing in ϕ .

Proof. See Appendix D.

Recall that δ accounts for an exogenous decrease in production efficiency due to the ageing of technology and organizational forgetting. The monopolist does not need to exert the desired level of the R&D effort all at once but can smooth it over time. With a convex R&D cost function this enables him to reduce his investment costs. The higher the rate at which the technology depreciates, the less advantageous this smoothing will be, as a larger part of what the monopolist invests just enough to keep his marginal cost constant at the desired level. Hence, the higher is δ , the more costly this maintenance will be, as a larger part of the technology must be replaced in every period.

The higher is *b*, the more it costs a firm to implement a particular R&D effort at every instant. As an increase in either δ or *b* makes implementing the desired R&D effort level more costly, an increase (decrease) in *b* and/or δ increases (decreases) the steady-state level of marginal cost and decreases (increases) the steady-state output, profits and total surplus, as stated in Lemma 3 above.¹⁷

¹⁷Steady-state marginal cost in the non-rescaled model equals $c^M = \left(A - \sqrt{A^2 - 16b\delta\rho}\right)/2$. See Appendix D for the steady-state values of other variables of the non-rescaled model.

4 Global optimality and industry dynamics (bifurcations of optimal vector fields)

Thus far we have focused on the steady state. Obviously, the stable manifold of the saddle-point steady state is one of the candidates for an optimal solution. However, as the following remark shows, it may not (always) be the optimum.

Remark 1. The Pontryagin function of the optimization problem as given in (11) is not jointly concave in the state and control variables. Hence, necessary conditions for an optimum are not necessarily sufficient.

Proof. See Appendix E.

Therefore, we have to carry out a detailed bifurcation analysis to assess the dependence of the solution structure on the model parameters.^{18,19} This yields a complete overview of how optimal R&D investment and concomitant production respond to changes in the initial level of marginal cost and to varying characteristics of the R&D process. To analyze global optimality, we have to inspect the entire state-control space, which is the set of solutions (c(t), k(t)) of the state-control system considered as parameterized curves (trajectories) in a plane. Its general form is sketched in Figure 2.

The dotted vertical line c = 1 separates the region with zero production from the region with a positive level of production. In the left region, the parabola which achieves its maximum at c = 1/2 is the locus $\dot{k} = 0$. The loci representing points where $\dot{c} = 0$ are the horizontal line $k = 1/\phi$ and the vertical line c = 0, which coincides with the k-axis. Loci $\dot{c} = 0$ and $\dot{k} = 0$ intersect in the saddle-point steady state (S_2) and the two unstable steady states (S_1 and S_3). The black arrows in squares indicate the direction of trajectories in each respective region. The horizontal arrows indicate an increase or a decrease in c, whereas the vertical ones refer to changes in k. A number of trajectories

¹⁸For a recent discussion of the use of bifurcation theory in analyzing non-convex optimal control problems with multiple equilibria, see Kiseleva and Wagener (2010).

¹⁹Note that Remark 1 applies to all related papers as well, including Lambertini and Mantovani (2009), Cellini and Lambertini (2009), and Kováč *et al.* (2010). Without further analysis it remains unclear whether these papers discuss truly optimal behavior. A sufficiency proof that does not rely on joint concavity of state and control variables can be given by adapting the Stalford sufficiency theorem (Stalford, 1971) to an infinite horizon setting, as, for instance, in Polasky *et al.* (2011).

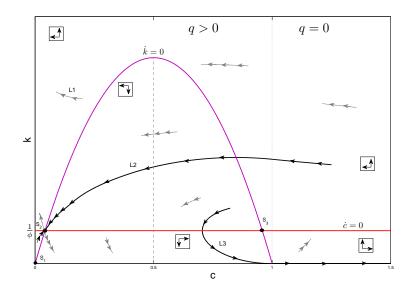


Figure 2: Illustrative sketch of the state-control space.

is indicated by grey arrows. Together, the trajectories cover the entire space. However, there are only two trajectories leading to the saddle-point steady state; these are called "stable paths" or "stable manifolds". They are indicated by the connected black arrows pointing towards S_2 . The lighter arrows pointing away from S_2 indicate the two unstable paths. Trajectories spiral out from S_3 in a counter-clockwise motion.

The solution to the state-control system (28), which yields trajectories in the region with zero production, is given by:

$$k(t) = C_1 e^{\rho t},\tag{39}$$

$$c(t) = C_2 e^{\left(t - \frac{e^{\rho t} \phi C_1}{\rho}\right)},\tag{40}$$

where C_1 and C_2 are positive real constants.

A solution to the state-control system (26), which yields trajectories in the region with positive production, cannot be obtained analytically; we therefore make use of geometrical-numerical techniques.²⁰

A particular trajectory is a *candidate* for an optimal solution if it satisfies all nec-

²⁰See also Judd (1998) and Wagener (2003). All numerical simulations were done in *Matlab*.

essary conditions: those given by Pontryagin's Maximum Principle together with transversality conditions (16) and (17). As Figure 2 shows, there exists a whole range of solutions (c(t), k(t)) to the state-control system. These trajectories all satisfy the conditions of Pontryagin's Maximum Principle. By ruling out those trajectories that do not satisfy the transversality conditions, we are left with the candidates for an optimal solution.

Lemma 4. The points in Figure 2 in the set

$$\{(c,k): c \in (0,1), k = 0\}$$

cannot be a part of any optimal trajectory.

Proof. See Appendix F.

Lemma 5. All trajectories along which $k \to \infty$ and $c \to 0$ as $t \to \infty$ can be ruled out as optimal solutions.

Proof. See Appendix G.

For instance, the trajectory denoted by L_1 in Figure 2 is in the set of Lemma 5.

Lemma 6. The trajectory through the point (c, k) = (1, 0) satisfies the transversality conditions (16) and (17).

Proof. See Appendix H.

In Figure 2, the trajectory of Lemma 6 is labeled L_3 . We call this the "exit trajectory" as it implies that the firm (eventually) exits the market. Indeed, at some stage the exit trajectory enters the region where $c \ge 1$, at which point both R&D investment and production come to a halt.²¹

Notice that the stable path (denoted by L_2 in Figure 2) to the saddle-point steady state (S_2) also satisfies the transversality conditions. Notably, along this trajectory, both c and k converge to finite limits and, therefore, so does the co-state variable (24).

²¹Strictly speaking, the firm's marginal cost continues to increase along the exit trajectory due to the positive depreciation rate δ . However, we interpret any situation in which the firm stays inactive as one in which it has left the market.

As the following corollary to the above lemmas shows, we are left with two candidates for an optimal solution.

Corollary 1. The set of candidates for an optimal solution consists of the stable path of the saddle-point steady state and the trajectory through the point (c, k) = (1, 0).

Proof. See Appendix I.

To find an optimal solution from the candidates, we make use of the following lemma.

Lemma 7. If the triple $(c(\cdot), q(\cdot), k(\cdot))$ satisfies the necessary conditions (12)–(17), then

$$\Pi(c(0), q(0), k(0)) = \int_0^\infty \pi(q, k, c) e^{-\rho t} dt$$

= $\frac{1}{\rho} P(c(0), q(0), k(0), \lambda(0))$
= $\frac{1}{\rho} H(c(0), k(0)),$ (41)

where *H* is the Hamilton function in state-control form.

Proof. See, e.g., Grass *et al.* (2008), ch. 3, pp. 161–162.

Lemma 7 is useful whenever several paths satisfy all the necessary conditions for an optimal solution and we have to determine which one yields the highest discounted profit flow and is thus optimal. According to Lemma 7, these flows are equal to the rescaled values of the Hamiltonian (31) at the corresponding initial points. That is, for a given initial technology c(0), Lemma 7 enables us to determine a firm's total discounted profits when it decides to follow the stable path (Π_1) and when it decides to follow the exit trajectory (Π_2). The option with the largest profits is then an optimal solution.

A short additional remark about profits is in order. Note that as a consequence of Lemma 4, no optimal trajectory can enter the region $\{(c,k) | H(c,k) < 0\}$. For we know that the investment schedule k(t) = 0 for all t will always result in nonnegative profits, implying *a fortiori* that $H(c,k) \ge 0$ on any optimal trajectory. In other words,

following from Lemma 7, the present value of profits along any optimal trajectory is always nonnegative.

Figure 3 provides an example in which the optimal solution is the stable path leading to the saddle-point steady state. Note that the initial marginal costs can be higher than c = 1. In that case, the optimal decision leads the firm to first produce nothing but to invest increasingly in the reduction of its marginal costs. When marginal costs are below one, it starts producing. The level of R&D investments then gradually decreases to its long-run steady-state level. This is exactly the scenario behind the story of cellophane.²²

Note that having a phase of zero production as a part of the stable path is possible only if the range of initial marginal cost is not bounded to be below the choke price. In this phase instantaneous profits are negative, but total discounted profits are positive; the monopolist invests in R&D because once marginal cost is below the choke price, production becomes profitable.²³

In the example of Figure 3, the monopolist optimally selects an appropriate k such that (c_0, k_0) is located on the stable path. However, as the Arrow-Mangasarian sufficiency conditions are not satisfied (see Remark 1), not every investment policy that follows the stable path is necessarily optimal. For a technology with a sufficiently high initial marginal cost c_0 , it can be optimal for a firm to abstain from development, as investing along the stable path would lead to a loss. That is, the solution structure depends on the system's parameters.

Figure 4 presents the bifurcation diagram, which shows the optimal R&D investments for varying parameter values.²⁴ Every combination of the exogenous parameters yields a point on this diagram. It consists of five distinct regions, each representing a particular structure of the set of optimal solutions. In general, either there exists an upper bound of initial marginal cost below which the monopolist selects the stable path

²²Another example is the petrochemical industry. Stobaugh (1988) finds that R&D efforts aimed at process innovations increase every year in the period following the initial product innovation, only to decrease again after that. He also documents that the probability of the next process innovation being major is decreasing over time.

²³Related, the literature on optimal technology adoption predicts a substantial time lag between the discovery of a new technology and its adoption (see, e.g., Hoppe (2002) for an overview). Delaying the adoption provides additional information as to the profitability of a new technology; it is also rational if a better technology is expected to become available in the near future (see also Doraszelski (2004)). In this literature, however, technological progress is exogenous.

²⁴The bifurcation diagram in Figure 4 is drawn for the parameters of the non-rescaled model.

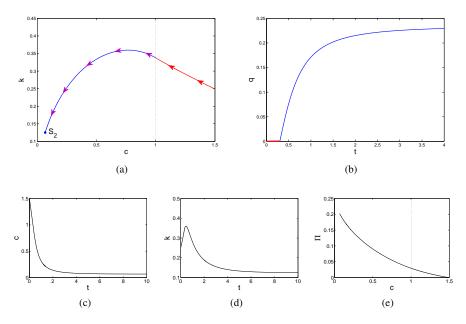


Figure 3: State-control space (a), time paths for quantity (b), marginal cost (c), and R&D effort (d), the correlation diagram between the total discounted profit and marginal cost (e), respectively. All plots show curves for parameter values (ρ , ϕ) = (1,8).

towards the saddle-point steady state, or the (new) technology is always phased out by a firm (eventually) exiting the market. Hence, if it exists, convergence to the saddle-point steady state along the stable path is only locally optimal; it is never globally optimal. In other words, if at all, investing in reducing marginal costs is only optimal for some sufficiently small initial marginal costs, but never for all possible initial marginal costs.

Solutions to the optimization problem for parameters outside the shaded region in Figure 4 are characterized by the presence of threshold values; the stable path of the saddle-point steady state is optimal if the initial marginal cost is below some threshold level, otherwise the exit trajectory is optimal. The threshold point can be either an indifference point²⁵ or a repeller. Indifference points are initial states $c = c_0$ for which the firm is indifferent between several possible investment policies (i.e., the stable path and the exit trajectory). A repeller differs from an indifference point in that at a repelling point, it is optimal to stay at that point forever after.²⁶

²⁵Also called Skiba points or DNSS points (see Grass et al. (2008)).

²⁶More formally, a threshold point separates two basins of attraction: the initial values of the marginal cost below these points constitute the basin of attraction of the saddle-point steady-state; the initial values of

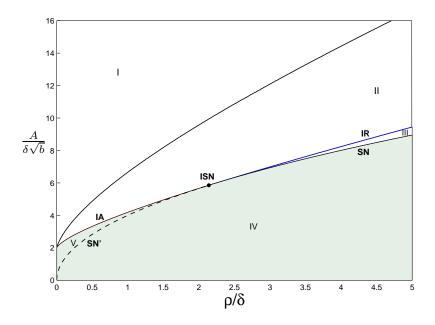


Figure 4: *Bifurcation diagram.* The uppermost curve represents parameter values for which the indifference point between the stable path and the exit trajectory is exactly at c = 1. SN stands for the "saddle-node bifurcation curve", SN' for the "inessential saddle-node bifurcation curve", IA for the "indifference-attractor bifurcation curve", and IR for the "indifference-repeller bifurcation curve". ISN indicates the "indifference-saddle-node point", which is the point from which indifference-attractor, indifference-repeller, saddle-node and inessential saddle-node curves emanate. Its coordinates are $\tilde{\rho} \approx 2.14$ and $\phi \approx 5.85$. Roman numerals (I-V) indicate the corresponding parameter regions discussed in the main text. The shaded region (regions IV and V) indicates the parameter region for which the firm always exits the market. Notice that the axes are excluded from the admissible parameter space. The curve representing indifference points at c = 1 obtains a value of $\phi \approx 2.00$ for $\tilde{\rho} = 1 \times 10^{-5}$.

4.1 Region I

In *Region I*, the point of indifference is above one. Therefore, the system will converge to the saddle-point steady state for all initial technologies $c_0 \in (0, 1]$.²⁷ A typical example is depicted in Figure 5. The arrows indicate the direction of motion along optimal trajectories. The two regions (production, no production) are split by the dotted vertical line c = 1. For a given ρ , *Region I* is characterized by a relatively high ϕ , that is, by relatively large demand and/or high R&D efficiency. In such a favorable

the marginal cost above these points constitute the basin of attraction to exit the market. An indifference point lies in both basins of attractions, a repeller in neither.

²⁷Note that $c_0 = 0$ corresponds to an unstable steady state; for this trivial value of the marginal cost, the monopolist stays at that point forever.

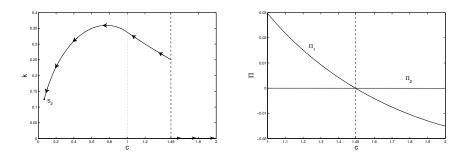


Figure 5: State-control space and the total profit function when the indifference point is in the region with zero production. The plots are drawn for parameter values $(\rho, \phi) = (1, 8)$.

environment, the monopolist can compensate for his early stage losses if, initially, production is postponed as that would yield negative mark-ups. There exists, however, a finite upper bound $\hat{c} > 1$ for the initial marginal costs c_0 beyond which future profits are not enough to compensate for short-run losses ($\hat{c} \approx 1.49$ in the case considered). In that case, the initial technology will not be developed further. In Figure 5, the point of indifference (indicated by a dashed vertical line) occurs where the total profit function for the stable path (Π_1) in the right-hand picture obtains a value of zero (which is the value of the total profit function for the exit trajectory Π_2 in the region $c \ge 1$). As the total profit beyond this point is negative (the firm makes a net loss), the firm prefers not to invest at all. Starting from any initial value of marginal cost below the indifference point, the firm's optimal decision is to follow the stable path. The exact position of the indifference point is defined in Lemma 8.

Lemma 8. If an indifference point between the stable path and the exit trajectory exists in the region with zero production $(c \ge 1)$, it is given by the value of the marginal cost $\hat{c} \ge 1$ for which the point $(c, k) = (\hat{c}, \frac{2}{\phi})$ lies on the stable path. For $c_0 = \hat{c}$, total discounted profits are zero for both trajectories.

Proof. See Appendix J.

Observe that in Figure 5, the indifference point $\hat{c} \approx 1.49$ corresponds to the value of $k = 2/\phi = 0.25$ on the stable path, as implied by Lemma 8. For this value of k, the

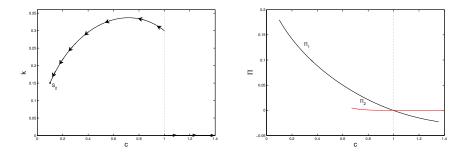


Figure 6: State-control space and the total profit functions when the indifference point is on the boundary between the two regions ($\hat{c} = 1$). The plots are drawn for parameter values (ρ, ϕ) = (1, 6.655).

Hamilton function (31) obtains a value of zero and, therefore, by Lemma 7, the total discounted profits (Π_1) are zero as well (see the right plot in Figure 5). For $c > \hat{c}$, the value of k on the stable path is smaller than $2/\phi$ and so the total profits are negative; whereas for $1 \le c < \hat{c}$, the value of k on the stable path is larger than $2/\phi$ and so the total profits are positive.

Region I is bounded from below by the curve for which the point of indifference is at c = 1. This particular situation is depicted in Figure 6: for all initial values of the marginal cost $c_0 \in (0, 1)$, the optimal-solution trajectory is the stable path of the saddle-point steady state, while for $c_0 \in (1, \infty)$, it is optimal not to produce and to invest anything; at $c_0 = \hat{c} = 1$, either strategy is optimal. The right-hand picture shows that the total profit Π_1 for the stable path is zero at c = 1. For comparison, the total profit Π_2 on the exit trajectory, which passes through the point (c, k) = (1, 0), is plotted up to its conjugate point (the last point on the trajectory where marginal cost is still decreasing; see also footnote 30 below).

The stable path in the region with zero production $(c \ge 1)$ is characterized by a decreasing k and an increasing c as $t \to -\infty$ (see (39)–(40) and Figure 5). From Lemma 8 follows that the indifference point between the stable path and the exit trajectory is at c = 1 when the stable path reaches $k = 2/\phi$ at c = 1 (see Figure 6). As investment is increasing in ϕ (for a given ρ), we can find for every ρ a high enough ϕ such that the indifference point is either at c = 1 or in the region with $c \ge 1$. A

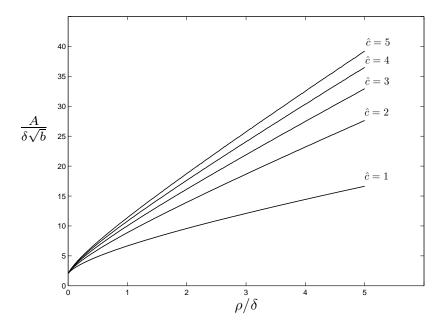


Figure 7: Dependence of the indifference point \hat{c} on the model parameters. Curves are drawn for five fixed values of \hat{c} . Given ρ , a higher value of A or a lower value of b and/or δ increases the value of marginal cost at the indifference point, and vice versa. This effect is larger, the smaller the value of ρ . An increase in ρ reduces the critical value of marginal cost, ceteris paribus.

higher ϕ for a given ρ in *Region I* moves the indifference point further beyond c = 1. Consequently, *Region I* is not bounded from the right. Moreover, as k decreases along the stable path as $t \to -\infty$, at some instant the stable path, having a value of k at c = 1above $k = 2/\phi$, reaches the point with $k = 2/\phi$ for some c > 1 as $t \to -\infty$. Therefore, for parameters in *Region I*, we always have an indifference point at some $c = \hat{c} \ge 1$ and, in particular, the convergence to the saddle-point steady-state is for these parameters guaranteed if $c_0 \in (0, 1]$, but not for all $c_0 \in (1, \infty)$.

Figure 7 illustrates how the indifference point varies with the parameters.²⁸ Clearly, for a given discount rate ρ , the more efficient is the R&D process, the larger will be the range of initial marginal costs that lead the monopolist to select the stable path. Likewise, a higher choke price corresponds to a higher value of initial marginal cost at the indifference point.

²⁸Figure 7 is drawn for the parameters of the non-rescaled model.

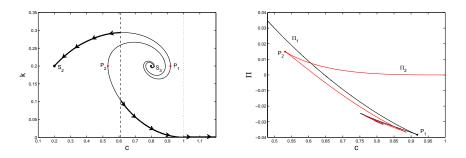


Figure 8: State-control space and the total profit functions when the indifference point is within the region with positive production. The plots are drawn for parameter values $(\rho, \phi) = (1, 5)$. The optimal-solution trajectories are marked by thick lines. The two candidates for the optimal solution, which both spiral out of the unstable steady state S_3 , are the stable path and the exit trajectory.

4.2 Regions II and III

In *Region II*, the indifference point \hat{c} is below c = 1. This implies in particular that at the indifference point the monopolist produces a positive quantity. Figure 8 illustrates this case. The indifference point occurs where the two respective total profit functions intersect, as indicated in the right-hand picture. For the particular parameter values considered in Figure 8, the indifference point occurs at $\hat{c} \approx 0.61$. Although it is also profitable to invest in R&D for some initial technology level $c_0 \in (\hat{c}, 1)$, the technology is not promising enough for the monopolist to select the stable path towards the saddlepoint steady-state. Rather, he invests in R&D at some smaller rate that retards the decay of the technology level optimally; but eventually the technology will leave the market, and the monopolist with it.²⁹

For an even higher initial level of marginal cost ($c_0 \ge 1$), the monopolist does not initiate any activity; the new technology is not developed at all and the monopolist does not enter the market.³⁰

 $^{^{29}}$ An alternative possibility is that the marginal cost is bounded from above (e.g., by its initial level). Such a state constraint is straightforward to implement. When a limit value of *c* is above the choke price, the result is not changed as the exit trajectory enters the zero production region at the same instant as before. However, when a limit value is below the choke price, a different scenario evolves: once the firm has reached the limit value of *c*, it stops investing in R&D, but it continues to produce a positive amount. We thank an anonymous referee for making us aware of this alternative possibility.

³⁰Conjugate points P_1 and P_2 indicate the last element on the respective trajectory (when starting from its end) where the marginal cost is still increasing/decreasing. Considering a particular trajectory beyond this point would give us two possible k-points for each c, of which the one beyond the P point is suboptimal. Observe that in the plot of the profits, the value of the total profit for the exit trajectory (starting from its end)

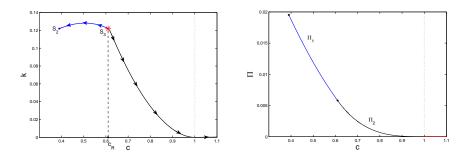


Figure 9: State-control space and the total profit functions when there is a repeller point (c_R) in the region with positive production. The plots are drawn for parameter values $(\rho, \phi) = (4, 8.2)$. The stable path of the saddle-point steady state and the exit trajectory both flow out of a repelling point, which is the unstable node S_3 (corresponding to c = 0.6098).

Utterback (1994) documents several examples of technologies where process innovations continue to be realized although it is known that these technologies will leave the market in the near future. For instance, steam-powered saws and conveyors were first used in the ice harvesting industry after the introduction of machine-made ice. And self-setting shutters, small plate cameras, and celluloid substitutes for glass were introduced in the dry plate photography at the time that Kodak introduced roll film cameras. These examples correspond well to the scenario predicted by the "exit trajectory".³¹

Region III is similar to Region II in that for $c_0 \ge 1$ the firm does not initiate any activity and only for sufficiently low initial marginal costs $c_0 \in (0, c_R)$ is there convergence to the saddle-point steady state. There is, however, a small qualitative difference between the two regions: in Region III, the threshold point is not an indifference point but a repeller (see Figure 9). Starting close to the repeller, the firm will "linger" near to it for a long time before deciding on developing the technology (if $c_0 < c_R$), or on phasing it out eventually (if $c_0 > c_R$).³²

Region III is separated from Region II by a type-2 indifference-repeller bifurcation

increases all along the way to P_2 (where it obtains a cusp point), beyond which it starts decreasing.

³¹Note that if $c_0 \in (\hat{c}, 1)$, the alternative strategy of producing but not investing anything in R&D, which corresponds to "following" the *c*-axis below the exit trajectory in Figure 8, yields lower total discounted profits than following the exit trajectory (recall Lemma 4).

³²Observe in Figure 9 that in the case of a repeller, the two trajectories never occur at the same time. Hence, the corresponding total profit functions (the right-hand picture) do not intersect.

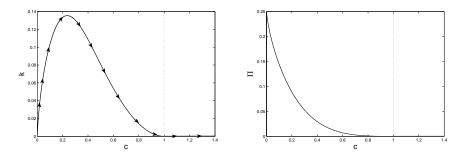


Figure 10: State-control space and the total profit function for the case with no saddlepoint steady state. Parameter values are $(\rho, \phi) = (1, 3)$. In this situation (and any other situation in which $\phi < 4\sqrt{\rho}$), the exit trajectory covers the whole space and it is the only solution. Observe that the total discounted profit is positive for c < 1.

curve (IR₁(2)), at which a repeller changes into an indifference point.³³ Fixing ϕ and varying ρ , it can be shown that the exit trajectory of a repeller lies below that of an indifference point.³⁴ This is also intuitive: a lower discount rate ρ (in the case of an indifference point) implies larger investments in R&D as the monopolist values future profits relatively more. Hence, for each initial value of marginal cost, the exit trajectory corresponding to a lower discount rate lies above the exit trajectory corresponding to a higher discount rate.

4.3 Regions IV and V

Region IV is separated from Region III by a saddle-node bifurcation curve, given by $\phi = 4\sqrt{\rho}$. This curve represents parameter combinations for which the saddle and the unstable node S_3 coincide and form a semi-stable steady state. After the bifurcation, both the saddle and the unstable node disappear. In such a case (see Figure 10), the only solution left is the exit trajectory: for $c_0 \ge 1$ the firm never initiates any activity, while for $c_0 \in (0, 1)$ the firm does invest in R&D but it will leave the market at some future instant. This is again intuitive as in Region IV we have that ϕ is relatively small (for a

³³Before the bifurcation (*Region II*), the unstable steady state (S_3) has two complex eigenvalues with positive real parts. The stable path and the exit trajectory both spiral out from it, which yields an indifference point to the left of the unstable steady state (focus). At the bifurcation, both eigenvalues become real and equal, such that all trajectories move away from the unstable steady state in the direction of a unique eigenvector. After the bifurcation, the steady state has two real positive eigenvalues and two corresponding (and distinct) eigenvectors.

 $^{^{34}\}text{Fixing }\rho$ and varying ϕ leads to the same conclusion.

given ρ). That is, R&D is relatively inefficient and/or market demand is relatively small. Note that in Figure 10, R&D investments are initially increasing over time for very low values of marginal costs. These investments reduce the pace at which the monopolist will leave the market; for low initial marginal cost it is profitable to slow down this process in order to profit optimally from the relatively favorable initial technology.

Finally, *Region V* resembles *Region IV* in that only the exit trajectory is optimal (see Figure 11). This is in spite of the fact that points in *Region V* have higher ϕ values than points in *Region IV* with the same value of ρ , and that the saddle-point steady state (S_2) together with its stable path (denoted by T_1) exists. A distinctive feature of *Region V* is that the exit trajectory (denoted by T_2) intersects the vertical line through the saddle-point steady state. Whenever this is the case, it is never optimal to be at the saddle, or to follow any stable path leading towards it. Rather, the exit trajectory is optimal, yielding a higher discounted profit flow: that is, the value of the Hamiltonian at any point of the exit trajectory is higher than at points with the same value of marginal cost on the stable path.³⁵

Observe that in the upper right graph of Figure 11 total discounted profit evaluated along the stable path (Π_1) is always below that evaluated along the exit trajectory (Π_2). For the sake of completeness, the second stable path (denoted by T_4) and the unstable path (denoted by T_3) are plotted jointly with the corresponding total discounted profit functions (Π_4 and Π_3 , respectively). The latter are also both lower than Π_2 . Therefore, as said, for parameter values in *Region V*, the exit trajectory is always optimal and the saddle-point steady state is never approached, neither from above nor from below.³⁶ Put differently, the saddle-point steady state is not even locally optimal.

The IA₁ curve that bounds *Region V* from above is an indifference-attractor bifurcation curve. At the IA₁ bifurcation (depicted in the second row of Figure 11), the exit trajectory (T_2) coincides with the unstable path (T_3) and yields a higher total discounted profit than the stable path converging to the saddle-point steady state from the right (see

³⁵This follows from the fact that the derivative of the Hamilton function with respect to k is negative for $k < 1/\phi$ (which is also the k-coordinate of the saddle-point steady state). ³⁶The second stable path is not drawn in the previous figures so as not to blur the exposition. The policy

³⁶The second stable path is not drawn in the previous figures so as not to blur the exposition. The policy that it induces implies that the long-run optimal marginal cost is just higher than the initial value of the marginal cost.

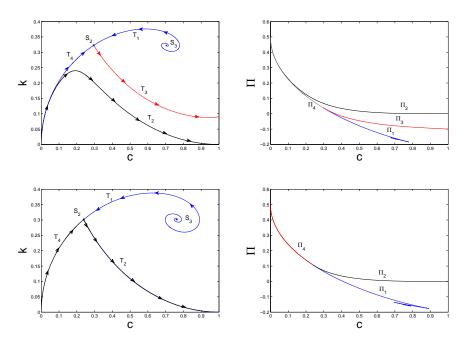


Figure 11: State-control space and the total profit functions before and at the indifferenceattractor bifurcation. The plots in the upper two graphs (before the bifurcation) are drawn for $(\rho, \phi) = (0.5, 3.1)$. The plots in the lower two graphs show that the bifurcation occurs at $(\rho, \phi) \approx (0.5, 3.32)$.

also Kiseleva and Wagener (2011) and Wagener (2003)). Hence, there is no convergence to the saddle from above. However, the exit trajectory does not intersect with any vertical line in the region to the left of the saddle-point steady state. Consequently, the left-hand stable path remains the only candidate for an optimal solution. That is, for initial marginal cost lower than the saddle-point steady-state value, the optimal trajectory approaches the saddle-point steady state from below.³⁷

All in all, before the bifurcation, the exit trajectory lies below the unstable path of the saddle; after the bifurcation, it lies above it. At the bifurcation, the two trajectories change their relative positions, the saddle-point steady state becomes locally optimal and an indifference point is created. That is, we enter *Region II*. In other words, moving from *Region V*, where the firm always exits the market, by increasing ϕ while keeping ρ fixed, we arrive at *Region II*, where staying in the market might be an optimal

³⁷Notice that for the values of initial marginal cost lower than the saddle-point steady state, the saddle is also approached from below for parameter values in *Regions I, II, and III* as the left-hand stable path of the saddle point remains the only candidate for the optimal solution in the respective region.

solution, depending on the initial level c_0 of the marginal cost. The interpretation is straightforward: staying in the market becomes an option if R&D becomes more efficient and/or if market demand increases.

The curve bounding *Region V* from below is an inessential saddle-node bifurcation curve (SN'_1) . It is similar to the SN_1 curve in that two new steady states appear when crossing it. However, it is *inessential* for the optimization problem: the bifurcation does not entail a qualitative change of the optimal strategy, as the steady states involved in the SN'_1 bifurcation are not associated with the optimal strategy; as above, the exit trajectory below the unstable path of the saddle remains the optimal trajectory. Consequently, while the state-control system does bifurcate, the structure of the optimal solutions does not.

Together, *Region IV* and *Region V* represent the parameter region for which there is never convergence to the saddle-point steady state because it is either not optimal to approach it (*Region V*) or because there is no saddle-point steady state at all (*Region* IV).³⁸

4.4 Unexpected changes in the parameters

Note that our model is deterministic in the sense that the monopolist does not form expectations about possible changes in exogenous parameters. One interpretation of this implicit assumption is that the monopolist is confronted, in fact, with great uncertainty: even the probability distributions over exogenous parameters are unknown. Alternatively, the implicit assumption is that the monopolist does not expect any change in the value of any exogenous parameter. Whatever the underlying motivation, in this section we illustrate that unexpected changes in exogenous parameters are straightforward to implement in our framework.³⁹

Utterback (1994) documents how producers of some of today's widely-used con-

³⁸Cellini and Lambertini (2009) restrict the values of the marginal cost to $c \in [0, 1]$ and assume that the saddle-point steady state is always optimally approached. Clearly, their conclusions hold only for parameters in *Region I* as only there we have convergence to the steady state for initial marginal costs $c \in (0, 1]$. A look at Figure 4, however, reveals that four other regions exist where the convergence is not guaranteed for all $c \in (0, 1]$.

 $c \in (0, 1]$. ³⁹Introducing stochasticity into the model by having the monopolist forming expectations about the evolution of exogenous parameters is an extension that we leave for future research.

sumer products considerably underestimated their market potential (including photographic cameras, which were supposed to be limited to a small cadre of professional portrait takers, and computers, which IBM thought to be limited to massive processing operations). Yet other products unexpectedly became destined to leave the market (such as typewriters and gas lamps) due to the emergence of a superior technology (computerized word processing and electric lamps respectively). We thus illustrate how unexpected changes in exogenous parameters can be incorporated into the model by examining unexpected changes in demand, as reflected in changes of A (examining unexpected changes in other exogenous parameters would be likewise incorporated).

Suppose the monopolist is confronted with some initial marginal cost c_0 such that Region I applies. If $c_0 > \hat{c}$ for a given set of parameters, the development of this technology is unprofitable. However, a sufficient increase in A, which increases the profit potential of a technology, can then lead to a situation whereby $c_0 < \hat{c}$ (recall from Figure 7 that \hat{c} is increasing in A); an unprofitable development project has turned into a profitable one. The firm then follows the stable path to the saddle-point steady state. Once the firm is following the stable path, any upward move in A pushes the firm onto the stable path characterized by larger R&D efforts k for all c and a lower long-run steady-state value of marginal cost.

Alternatively, there is a sudden drop in A because, for instance, a new and superior technology is entering the market. And suppose this drop is large enough to bring us to Region IV (where the exit trajectory is globally optimal). This situation is depicted in Figure 12. The final value of marginal costs on the stable path before the parameter change (c_1) translates into a new "initial value" of marginal costs (c_2) on the exit trajectory corresponding to the value of parameters after the change in A. From the point of the change onwards, the monopolist follows the exit trajectory and eventually exits the market.⁴⁰

⁴⁰If the fall in A is large, it can well happen that after the change marginal cost exceeds the choke price $(c_2 > 1)$ such that production and R&D are terminated immediately. Alternatively, if the drop in A is sufficiently small, such that for c_2 the stable path is still optimal, the firm follows the new stable path characterized by lower k for all c and higher long-run steady-state marginal cost.

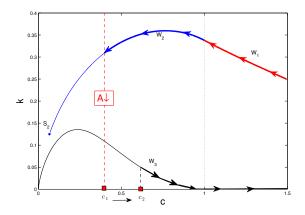


Figure 12: An unexpected fall in the demand parameter A. The upper trajectory (stable path) is drawn for $(\rho, \phi) = (1, 8)$. Beginning in the area with zero production, the firm starts investing along the stable path (W_1) and after some time also producing (W_2) . The vertical dashed line indicates the moment at which a sudden fall in A occurs. For the new value of parameters $(\rho, \phi) = (1, 3)$, the firm starts following the exit trajectory (W_3) .

4.5 Welfare analysis

In this section, we highlight one particular insight from our model. Recall that the level of marginal cost at the indifference point (the marginal cost level for which the monopolist is indifferent between developing the technology further and opting out) is positively related to the profit potential of the new technology (see the discussion in relation to Figure 7). Figure 13 (panel (*a*)) shows the total discounted value of total surplus along the monopolist's *optimal trajectory* for two situations: a relatively high profit potential (black curve), and a lower profit potential (grey curve). Total surplus is only positive for all initial marginal cost below the indifference point (obviously, when there is no production, there is no surplus). The figure illustrates an implication of our analysis: market characteristics that influence future production affect the decision of the monopolist to develop a new technology. In particular, if market regulations are such that future mark-ups are reduced, for instance because the monopolist is forced to price closer to its marginal cost, the range of initial technologies that it is profitable to develop shrinks ($\hat{c}_2 < \hat{c}_1$). The loss of total surplus that this reduced range brings about is a hidden cost of competition policies if the indifference point is above c = 1, as in that

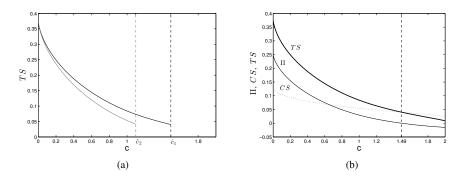


Figure 13: Panel (a) shows the total discounted value of total surplus for $(\rho, \phi) = (1, 8)$ (black) and $(\rho, \phi) = (1, 7)$ (grey). Panel (b) shows total discounted profits (thin solid curve), consumer surplus (dotted curve) and total surplus (thick solid curve) along the stable path for different initial values of marginal costs. Beyond the indifference point $\hat{c} \approx 1.49$, total profits are negative, but consumer surplus and total surplus are positive. $(\rho, \phi) = (1, 8)$.

case, initially there is no production. Hence, not developing the technology does not surface as an immediate reduction in total surplus. Indeed, for a proper assessment of the trade-off between static and dynamic efficiency these hidden costs should be taken into account.

As an illustration, consider first Figure 13, panel (*b*). It displays total discounted profits, consumer surplus and total surplus along the *stable path* of the saddle steady state, in case the indifference point is in the region of zero production. Note that there is a small range of initial marginal cost, just above the indifference point, that would induce the monopolist to abandon the technology at the cost of giving up consumer surplus (and hence, total surplus). Likewise, Figure 14 compares total discounted profits and total surplus along the stable path and the exit trajectory (up to their respective conjugate points; see footnote 30) when the indifference point is in the region of positive production. In this case, total surplus on the stable path between \hat{c}_1 and \hat{c}_2 is higher than on the exit trajectory (the trajectory chosen by the monopolist; see also the discussion of Figure 8). In both these cases the monopolist gives up on development 'too early'; had it chosen the stable path (rather than not developing the technology at all, or choosing the exit trajectory), more surplus would have been generated. As is well known, this discrepancy between the private and social indifference point is caused by the inability

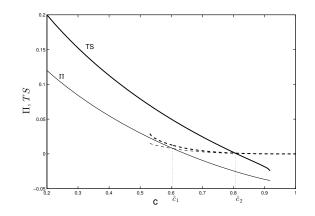


Figure 14: Total discounted profits (thin) and total surplus (thick) along the stable path (solid) and the exit trajectory (dashed) when the indifference point is in the region of positive production; $(\rho, \phi) = (1, 5)$.

of the monopolist to appropriate the entire surplus it generates by its R&D activities.⁴¹ What is revealed by our analysis, however, is that the cost of this discrepancy can remain hidden if the decision of the monopolist affects production that has not been started yet.

5 Concluding remarks

Existing models of R&D are difficult to reconcile with empirical aspects of R&D. These include the observation that many initial technologies ("prototypes" or "ideas") need to be developed further before they can enter the market, that only a minority of initial technologies is successfully brought to the market, that production starts only after an initial stage during which the technology is developed further, and that process innovations are implemented for technologies that are destined to leave the market. In this paper, we have developed a dynamic model of R&D that describes the R&D investment decision of a monopolist, and that is better aligned with these empirical observations.

A distinguishing feature of our analysis is that we provide a global analysis, as opposed to all related papers. Accordingly, we do not limit initial technologies to carry

⁴¹Simulating the model over a wide range of parameter configurations, we never encountered a case with a social indifference point lower than the monopolist's one.

marginal costs that are below the choke price. The part of the parameter space for which such technologies could eventually lead to the saddle-point steady state is not negligible indeed. More generally, we have shown that there always exists a critical value of initial marginal cost above which the firm does not initiate any (R&D) activity; the stable path of the saddle-point steady state is never globally optimal.

Our analysis suggests a careful look at competition policies that reduce mark-ups in products markets. Reduced future profits diminish the profitability to develop further initial technologies, in particular those that come with marginal costs above the choke price. Not developing further these technologies does not surface as a cost of competition policy as there is no production yet that will be taken from the market. Obviously, a proper assessment of a competition policies requires an analysis that includes firm competition, an exercise that is part of our future research agenda.

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Appendices

A Proof of Lemma 1

A rescaled variable or parameter is distinguished by a tilde: for instance, if π denotes profit, then $\tilde{\pi}$ denotes profit in rescaled variables. The profit function in the original (non-rescaled) model is:

$$\pi = (A - q - c) q - bk^2$$
 (A.1)

Using the conversion rules given in Lemma 1, we obtain:

$$\begin{aligned} \pi &= (A - q - c) q - bk^2 \\ &= (A - A\tilde{q} - A\tilde{c}) A\tilde{q} - b \left(\frac{A}{\sqrt{b}}\tilde{k}\right)^2 \\ &= A^2 \left((1 - \tilde{q} - \tilde{c}) \tilde{q} - \tilde{k}^2\right) \\ &= A^2 \tilde{\pi} \end{aligned}$$

The equation for the evolution of marginal cost over time is in original variables given by:

$$\dot{c}(t) = c(t) (-k(t) + \delta).$$
 (A.2)

Write $c(t) = c\left(\frac{1}{\delta}\tilde{t}\right)$. Then

$$\frac{dc}{d\tilde{t}} = \frac{dc}{dt}\frac{dt}{d\tilde{t}}$$
$$= \frac{\dot{c}}{\delta}$$
$$= \left(1 - \frac{1}{\delta}k\right)c.$$

Setting $k = \frac{A}{\sqrt{b}}\tilde{k}$ and substituting it in the previous equation, we obtain:

$$\frac{dc}{d\tilde{t}} = \left(1 - \frac{A}{\delta\sqrt{b}}\tilde{k}\right)c.$$

It is now natural to introduce $\phi = \frac{A}{\delta\sqrt{b}}$. Note that if $\tilde{c} = c/A$, then $\dot{\tilde{c}} = \dot{c}/A$ and

$$\dot{\tilde{c}} = \left(1 - \phi \tilde{k}\right) \tilde{c}$$

Observe finally that if $t = \frac{\tilde{t}}{\delta}$, then $e^{-\tilde{\rho}\tilde{t}} = e^{-\rho t}$ if and only if $\tilde{\rho} = \frac{\rho}{\delta}$.

B Proof of Lemma 2

As $\rho > 0$, the equilibrium condition $\dot{k} = \rho k = 0$ implies that k = 0. But then $\dot{c} = c (1 - \phi k) = c$. This is equal to 0 if, and only if, c = 0. But this cannot be, as in the system with zero production we have $c \ge 1$.

C Proof of Proposition 1

Assume $\phi > 4\sqrt{\rho}$. The stability of the steady states can be analyzed by evaluating the trace and determinant of the following Jacobian matrix:⁴²

$$J^{M} = \begin{bmatrix} \frac{\partial \dot{c}}{\partial c} & \frac{\partial \dot{c}}{\partial k} \\ \\ \frac{\partial k}{\partial c} & \frac{\partial k}{\partial k} \end{bmatrix} = \begin{bmatrix} 1 - \phi k & -\phi c \\ \\ -\frac{\phi(1-2c)}{4} & \rho \end{bmatrix}.$$
 (C.1)

At c = k = 0, the trace τ of the matrix J^M is given as

$$\tau \stackrel{\text{def}}{=} \operatorname{tr} J^M = 1 + \rho > 0,$$

its determinant Δ is

$$\Delta \stackrel{\text{def}}{=} \det J^M = \rho > 0,$$

⁴²Note that the trace of the respective Jacobian matrix is equal to the sum of eigenvalues, while its determinant is equal to their product. If the real part of each eigenvalue is negative, then the steady state is asymptotically stable. If the real part of at least one of the eigenvalues is positive, then the steady state is unstable. More particular, if one eigenvalue is real and positive and the other one real and negative, the steady state is said to be a saddle. In this last situation, there are four special trajectories, the separatrices, two of which are forward asymptotic to the saddle, while the other two are backward asymptotic to it. The union of the former separatrices with the saddle form the stable manifold of the saddle; analogously, the union of the latter with the saddle form the unstable manifold.

and its discriminant D is

$$D \stackrel{\text{def}}{=} \tau^2 - 4\Delta = (1 - \rho)^2 > 0.$$

Hence, this steady state is an unstable node.

Evaluating the Jacobian matrix at

$$k = \frac{\phi}{4\rho}c(1-c), \quad c = \frac{1}{2} + V = \frac{1}{2} + \frac{1}{2}\sqrt{1 - \frac{16\rho}{\phi^2}},$$
 (C.2)

we obtain $\tau=\rho>0$ and

$$\Delta = \frac{\sqrt{\phi^2 - 16\rho} \left[\phi + \sqrt{\phi^2 - 16\rho}\right]}{8} = \phi^2 V \left(\frac{1}{4} + \frac{1}{2}V\right), \quad (C.3)$$

which is clearly positive if $\phi > 4\sqrt{\rho}$. Hence, this steady state is also unstable. The discriminant takes the value

$$D = \rho(8+\rho) - \frac{1}{2}\phi\left(\phi + \sqrt{\phi^2 - 16\rho}\right),$$
 (C.4)

which is zero for $\phi = \phi_0 = \sqrt{\frac{\rho(8+\rho)^2}{(4+\rho)}}$, negative for $\phi > \phi_0$, and positive for $4\sqrt{\rho} \le \phi < \phi_0$. The steady state is an unstable node if D > 0 and an unstable focus if D < 0. In the latter case, the eigenvalues of J^M are complex conjugates with positive real parts.⁴³

Finally, evaluating the Jacobian matrix at

$$k = \frac{\phi}{4\rho}c(1-c), \quad c = \frac{1}{2} - V = \frac{1}{2} - \frac{1}{2}\sqrt{1 - \frac{16\rho}{\phi^2}},$$
 (C.5)

we obtain $\tau=\rho>0$ and

$$\Delta = \frac{\phi^2 - 16\rho - \phi\sqrt{\phi^2 - 16\rho}}{8}.$$
 (C.6)

If $\phi > 4\sqrt{\rho}$, then $\Delta < 0$, and the eigenvalues are real and have opposite sign. Therefore,

⁴³Note that the eigenvalues of J^M are $r_{1,2} = \frac{1}{2} \left(\tau \pm \sqrt{\tau^2 - 4\Delta} \right)$.

(C.5) is a saddle-point steady state of the system. Observe that for $\phi = 4\sqrt{\rho}$, the two steady states (C.2) and (C.5) coincide at c = 1/2 and $k = \frac{1}{\phi}$. A saddle-node bifurcation occurs at these parameter values, where the two equilibria collide and disappear.⁴⁴

Substituting the expression for the steady-state marginal cost of the steady states other than the origin into (32), the expression for the optimal investment in the steady state simplifies to $k^M = \frac{1}{\phi}$.

D Proof of Lemma 3

As we deal with both versions of the model, we use tildes in the notation of rescaled variables and parameters to avoid ambiguity. We first prove the stated relations between parameters and steady-state values for the rescaled model. Take $\phi > 4\sqrt{\tilde{\rho}}$, such that the saddle-point steady state exists. The saddle-point steady state values of the marginal cost, output, and profit are given in (36)-(38). Consumer surplus and total surplus are, respectively:

$$\widetilde{CS}^{M} = \frac{(1-\tilde{p})\tilde{q}}{2} = \frac{1}{2}(\tilde{q}^{M})^{2} = \frac{1}{32}(1+2V)^{2},$$
 (D.1)

$$\widetilde{TS}^{M} = \widetilde{\pi}^{M} + \widetilde{CS}^{M} = \frac{3}{32} (1+2V)^{2} - \frac{1}{\phi^{2}}.$$
 (D.2)

⁴⁴In Figure 2, this corresponds to the locus $\dot{c} = 0$, which is the horizontal line $k = \frac{1}{\dot{\phi}}$, becoming tangent to the locus $\dot{k} = 0$ at its peak.

Substituting the expression for V given in (35) in the steady state expressions (36)-(38) and (D.1)-(D.2) and taking appropriate derivatives, we obtain:

$$\begin{split} &\frac{\partial\tilde{c}^{M}}{\partial\phi}=-\frac{8\tilde{\rho}}{\phi^{2}\sqrt{\phi^{2}-16\tilde{\rho}}}<0,\\ &\frac{\partial\tilde{c}^{M}}{\partial\tilde{\rho}}=\frac{4}{\phi\sqrt{\phi^{2}-16\tilde{\rho}}}>0,\\ &\frac{\partial\tilde{q}^{M}}{\partial\phi}=\frac{4\tilde{\rho}}{\phi^{2}\sqrt{\phi^{2}-16\tilde{\rho}}}>0,\\ &\frac{\partial\tilde{q}^{M}}{\partial\tilde{\rho}}=-\frac{2}{\phi\sqrt{\phi^{2}-16\tilde{\rho}}}<0,\\ &\frac{\partial\tilde{\pi}^{M}}{\partial\phi}=\frac{2\left(1+\tilde{\rho}+\frac{\tilde{\rho}\phi}{\sqrt{\phi^{2}-16\tilde{\rho}}}\right)}{\phi^{3}}>0,\\ &\frac{\partial\tilde{\pi}^{S}}{\partial\phi}=-\frac{1+\frac{\phi}{\sqrt{\phi^{2}-16\tilde{\rho}}}}{\phi^{2}}<0,\\ &\frac{\partial\widetilde{CS}^{M}}{\partial\phi}=\frac{\tilde{\rho}+\frac{\tilde{\rho}\phi}{\sqrt{\phi^{2}-16\tilde{\rho}}}}{2\phi^{2}}<0,\\ &\frac{\partial\widetilde{CS}^{M}}{\partial\phi}=\frac{2+3\tilde{\rho}+\frac{3\tilde{\rho}\phi}{\sqrt{\phi^{2}-16\tilde{\rho}}}}{\phi^{3}}>0,\\ &\frac{\partial\widetilde{TS}^{M}}{\partial\phi}=-\frac{3+\frac{3\phi}{\sqrt{\phi^{2}-16\tilde{\rho}}}}{2\phi^{2}}<0. \end{split}$$

All inequalities follow straightforwardly form the admissible values of the parameters.

We now prove the stated relations for the original model. Applying the conversion

rules defined in Lemma 1, we obtain:

$$\begin{split} c^{M} &= \frac{A - \sqrt{A^{2} - 16b\delta\rho}}{2}, \\ q^{M} &= \frac{A + \sqrt{A^{2} - 16b\delta\rho}}{4}, \\ \pi^{M} &= \frac{A^{2} - 8b\delta(\delta + \rho) + A\sqrt{A^{2} - 16b\delta\rho}}{8}, \\ CS^{M} &= \frac{1}{2}(q^{M})^{2} = \frac{1}{32}\left(A + \sqrt{A^{2} - 16b\delta\rho}\right)^{2}, \\ TS^{M} &= \frac{3A^{2} - 8b\delta(2\delta + 3\rho) + 3A\sqrt{A^{2} - 16b\delta\rho}}{16}. \end{split}$$

Take $A > 4\sqrt{b\delta\rho}$, such that the saddle-point steady state exists. Then:

$$\frac{\partial c^M}{\partial b} = \frac{4\delta\rho}{\sqrt{A^2 - 16b\delta\rho}} > 0, \tag{D.3}$$

$$\frac{\partial c^M}{\partial \delta} = \frac{4b\rho}{\sqrt{A^2 - 16b\delta\rho}} > 0, \tag{D.4}$$

$$\frac{\partial q^M}{\partial b} = -\frac{2\delta\rho}{\sqrt{A^2 - 16b\delta\rho}} < 0, \tag{D.5}$$

$$\frac{\partial q^M}{\partial \delta} = -\frac{2b\rho}{\sqrt{A^2 - 16b\delta\rho}} < 0. \tag{D.6}$$

It follows from the definition of consumer surplus $(CS^M = \frac{1}{2} (q^M)^2)$ that the effect of b and δ on CS^M is qualitatively the same as their effect on q^M . Moreover, we have:

$$\begin{split} &\frac{\partial \pi^M}{\partial b} = -\delta(\delta + \rho) - \frac{A\delta\rho}{\sqrt{A^2 - 16b\delta\rho}} < 0, \\ &\frac{\partial \pi^M}{\partial \delta} = b\left(-2\delta + \rho\left(-1 - \frac{A}{\sqrt{A^2 - 16b\delta\rho}}\right)\right) < 0, \\ &\frac{\partial TS^M}{\partial b} = \frac{1}{2}\delta\left(-2\delta + \rho\left(-3 - \frac{3A}{\sqrt{A^2 - 16b\delta\rho}}\right)\right) < 0, \\ &\frac{\partial TS^M}{\partial \delta} = \frac{1}{2}b\left(-4\delta + \rho\left(-3 - \frac{3A}{\sqrt{A^2 - 16b\delta\rho}}\right)\right) < 0. \end{split}$$

Again, all inequalities follow straightforwardly from the relevant values of the parameters. $\hfill \square$

E Proof of Remark 1

The Hessian matrix of the Pontryagin function with respect to control and state variables is

$$D_{(q,k,c)}^{2}P = \begin{bmatrix} \frac{\partial^{2}P(\cdot)}{\partial q^{2}} & \frac{\partial^{2}P(\cdot)}{\partial q\partial k} & \frac{\partial^{2}P(\cdot)}{\partial q\partial c} \\ \frac{\partial^{2}P(\cdot)}{\partial k\partial q} & \frac{\partial^{2}P(\cdot)}{\partial k^{2}} & \frac{\partial^{2}P(\cdot)}{\partial k\partial c} \\ \frac{\partial^{2}P(\cdot)}{\partial c\partial q} & \frac{\partial^{2}P(\cdot)}{\partial c\partial k} & \frac{\partial^{2}P(\cdot)}{\partial c^{2}} \end{bmatrix} = \begin{bmatrix} -2 & 0 & -1 \\ 0 & -2 & -\phi\lambda \\ -1 & -\phi\lambda & 0 \end{bmatrix}.$$
(E.1)

The leading principal minor of the first order of this matrix is -2. The leading principal minor of the third order, which is also the determinant, of the above matrix is equal to $2 + 2\phi^2\lambda^2$. As the leading principal minors of odd order do not have the same sign, the matrix is indefinite. Consequently, the Pontryagin function is nowhere jointly concave in state and control variables. Hence, the Arrow-Mangasarian sufficiency conditions are not satisfied.⁴⁵

F Proof of Lemma 4

To prove the lemma, we consider the state-co-state form of the solution, given by:

$$\dot{c} = \begin{cases} c \left(1 + \frac{1}{2}\phi^{2}\lambda c\right), & 0 \le c < 1, \lambda \le 0; \\ c, & 0 \le c < 1, \lambda > 0; \\ c \left(1 + \frac{1}{2}\phi^{2}\lambda c\right), & c \ge 1, \lambda \le 0; \\ c, & c \ge 1, \lambda > 0; \end{cases}$$
(F.1)

⁴⁵For details of sufficiency conditions, see, for instance, Grass et al. (2008).

$$\dot{\lambda} = \begin{cases} \frac{1}{2}(1-c) + \left(\rho - 1 - \frac{1}{2}\phi^{2}\lambda c\right)\lambda, & 0 \le c < 1, \lambda \le 0; \\ \frac{1}{2}(1-c) + (\rho - 1)\lambda, & 0 \le c < 1, \lambda > 0; \\ \left(\rho - 1 - \frac{1}{2}\phi^{2}\lambda c\right)\lambda, & c \ge 1, \lambda \le 0; \\ (\rho - 1)\lambda, & c \ge 1, \lambda > 0; \end{cases}$$
(F.2)

$$H(c,\lambda) = \begin{cases} \frac{1}{4}(1-c)^2 + \frac{1}{4}\phi^2\lambda^2c^2 + \lambda c, & 0 \le c < 1, \lambda \le 0; \\ \frac{1}{4}(1-c)^2 + \lambda c, & 0 \le c < 1, \lambda > 0; \\ \frac{1}{4}\phi^2\lambda^2c^2 + \lambda c, & c \ge 1, \lambda \le 0; \\ \lambda c, & c \ge 1, \lambda > 0. \end{cases}$$
(F.3)

Introduce the characteristic function χ_S of a set S by

$$\chi_S(x) = \begin{cases} 1 & \text{if } x \in S \\ 0 & \text{if } x \notin S, \end{cases}$$
(F.4)

differential equations (F.1) and (F.2) can be rewritten more compactly as:

$$\dot{c} = c + \chi_{(-\infty,0]}(\lambda) \frac{1}{2} \phi^2 \lambda c^2 \stackrel{\text{def}}{=} F(c,\lambda)$$
(F.5)

and

$$\dot{\lambda} = (\rho - 1)\lambda - \chi_{(-\infty,0]}(\lambda)\frac{1}{2}\phi^2 c\lambda^2 + \chi_{[0,1)}(c)\frac{1}{2}(1-c) \stackrel{\text{def}}{=} G(c,\lambda).$$
(F.6)

The state–co-state space is presented in Figure 15. We first consider the trajectories in the region where $c \ge 1$ and $\lambda > 0$, which are the solution to the following canonical

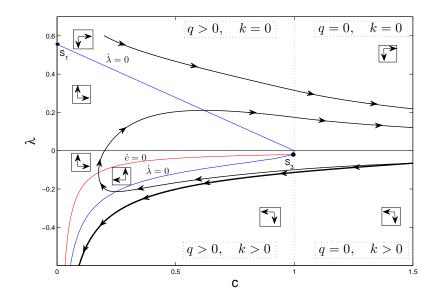


Figure 15: Illustrative sketch of the state-co-state space (for $\rho < 1$). The dotted vertical line c = 1 separates the region with zero production from the region with a positive level of production, whereas the horizontal line $\lambda = 0$ separates the region with positive investment from the region with zero investment. The loci $\dot{c} = 0$ and $\dot{\lambda} = 0$ intersect in the two unstable steady states (S_1 and S_3) and the saddle-point steady state S_2 (not indicated). A number of trajectories is indicated by black curves: the arrows point in the direction of the flow. The thick curve indicates the stable path leading to the saddle-point steady state.

system:

$$\begin{cases} \dot{c} = c, \\ \dot{\lambda} = (\rho - 1)\lambda, \end{cases}$$
(F.7)

given by $\lambda(t) = C_1 e^{(\rho-1)t}$ and $c(t) = C_2 e^t$, where C_1 and C_2 are positive constants. As mentioned in the main text, every candidate for an optimal solution must necessarily satisfy the transversality condition $\lim_{t\to\infty} e^{-\rho t}\lambda(t)c(t) = 0$. As we now show, the trajectories in the region considered violate this condition. Moreover, they also violate the following transversality condition:⁴⁶

$$\lim_{t \to \infty} e^{-\rho t} H(c, \lambda) = 0, \tag{F.8}$$

⁴⁶In words, the above condition means that the present value of the maximum of the Pontryagin function (the present value of the Hamiltonian) converges to zero when time goes to infinity.

which in this case coincides with the first transversality condition, as shown below. It follows from Michel (1982) that condition (F.8) is also a necessary condition, and hence that it allows exclusion of trajectories which verify the other necessary conditions in an infinite horizon optimization problem. The value of the Hamiltonian function evaluated along the considered trajectories is $H(c, \lambda) = \lambda(t)c(t) = C_1C_2e^{\rho t}$; it follows that

$$\lim_{t \to \infty} e^{-\rho t} H(c, \lambda) = \lim_{t \to \infty} e^{-\rho t} \lambda(t) c(t) = C_1 C_2 \neq 0.$$
(F.9)

Hence, no trajectory in the region given by the restrictions $c \ge 1$ and $\lambda > 0$ can be optimal.

Consider now trajectories in the region with 0 < c < 1 and $\lambda > 0$. These trajectories are the solution to

$$\begin{cases} \dot{c} = c, \\ \dot{\lambda} = \frac{1}{2}(1-c) + (\rho - 1)\lambda. \end{cases}$$
(F.10)

We show that any trajectory in this region sooner or later enters the region with $c \ge 1$ and $\lambda > 0$.

Observe that the equation $\dot{c} = c$ has as solution $c(t) = C_3 e^t$, where C_3 is a positive constant. Hence, along any trajectory in this region, c is increasing. From (F.2) follows that if $\lambda = 0$, then

$$\dot{\lambda} = \frac{1}{2}(1-c) > 0$$

for all $c \in (0, 1)$. Hence, trajectories in the region with 0 < c < 1 and $\lambda > 0$ cannot exit this region through the line segment $\{(c, \lambda) : c \in (0, 1), \lambda = 0\}$.

We now show that they also cannot exit through the point $(c, \lambda) = (1, 0)$.⁴⁷ Let $\mathbf{x} = (c, \lambda)$ and $\mathbf{x}_0 = (c_0, \lambda_0) = (1, 0)$. Furthermore, let $\mathfrak{F} : D \to \mathbb{R}^2$ be a vector function defined as

$$\mathfrak{F}(c,\lambda) = \Big(F(c,\lambda), G(c,\lambda)\Big),$$

where F is defined in (F.5) and G in (F.6).⁴⁸ Its domain is $D = \mathbb{R}_+ \times \mathbb{R} \subset \mathbb{R}^2$. We are

⁴⁷If they exited through this point, they would satisfy the transversality conditions as $\dot{\lambda} = 0$ for $\lambda = 0$ and $c \ge 1$.

⁴⁸Though continuous, functions F and G are not differentiable: F is not differentiable with respect to λ at $\lambda = 0$, whereas G is not differentiable with respect to c at c = 1. Using Peano's existence theorem,

then looking for a solution to a 2-dimensional nonlinear autonomous dynamical system of the form

$$\dot{\mathbf{x}} = \mathfrak{F}(\mathbf{x}), \quad \mathbf{x}(0) = \mathbf{x}_0. \tag{F.11}$$

Take 0 < r < 1. Then the set

$$D_0 = \overline{B_r(\mathbf{x}_0)} = \{ \mathbf{x} \in D : \|\mathbf{x} - \mathbf{x}_0\| \le r \},$$
(F.12)

is a neighborhood of \mathbf{x}_0 . First, we show that the restriction of the functions F and G on D_0 (a compact subset of D) is *Lipschitz*.⁴⁹ Consider first the function

$$F(c,\lambda) = c + \lambda \chi_{(-\infty,0]}(\lambda) \frac{\phi^2 c^2}{2}.$$

Write for brevity $\chi = \chi_{(-\infty,0]}$ and set $h(\lambda) = \chi(\lambda)\lambda$. Then,

$$|h(\lambda) - h(0)| = |\chi(\lambda)\lambda - 0| = |\chi(\lambda)\lambda| \le |\lambda| = 1 \cdot |\lambda - 0|.$$
(F.13)

Hence, h is Lipschitz on D_0 . As on compact sets continuously differentiable functions are Lipschitz, as well as sums and products of Lipschitz functions, it follows that F is Lipschitz on D_0 . Consider now on D_0 the function

$$G = (\rho - 1)\lambda - \lambda^2 \chi_{(-\infty,0]}(\lambda) \frac{\phi^2 c}{2} + \frac{1 - c}{2} \chi_{[0,1)}(c).$$

The first term of this expression is a linear function; the second term is the function h introduced above times a differentiable function. That the final term is Lipschitz at c = 1 is demonstrated in the same way as for the function h. It follows that G is Lipschitz on D_0 .

We have proved that the functions F and G are locally Lipschitz at \mathbf{x}_0 . Consequently, so is the vector function \mathfrak{F} . By the Picard-Lindelöf Theorem, the system (F.11) has a

the continuity of \mathfrak{F} implies that at least one solution to (F.11) exists. However, continuity is not enough to guarantee uniqueness. An additional condition that needs to be fulfilled to guarantee uniqueness of solutions, at least in some neighborhood of \mathbf{x}_0 , is that \mathfrak{F} is locally *Lipschitz* in \mathbf{x} at \mathbf{x}_0 .

⁴⁹Simply put, a function $f: D \to \mathbb{R}^n$ is said to be *Lipschitz* on $B \subseteq D$ if there exists a constant K > 0 such that $||f(x) - f(y)|| \le K ||x - y||$, for all $x, y \in B$. See Kelley and Peterson (2010), ch. 8, and Sohrab (2003), ch. 4, for the introduction to concepts and precise definitions of terms used in this section.

unique solution in a neighborhood of \mathbf{x}_0 . As this point is on the exit trajectory, no other trajectory can pass through it, in particular no trajectory from the region with $\lambda > 0$. As $\dot{c} > 0$ in the region with 0 < c < 1 and $\lambda > 0$, as shown above, all trajectories in this region exit through $\{c = 1, \lambda > 0\}$ and enter the region with $c \ge 1$ and $\lambda > 0$ as $t \to \infty$. We have already shown that none of these can be optimal.

Moreover, trajectories through points in the set $\{(c, \lambda) : c \in (0, 1), \lambda = 0\}$ satisfy $\dot{\lambda} > 0$; they move in the region for which $\lambda > 0$ and hence cannot be a part of any optimal trajectory. Due to a one-to-one correspondence between λ and k as given in the first part of (21), this set corresponds to the set $\{(c, k) : c \in (0, 1), k = 0\}$ in the state-control representation of the solution. As we have shown, all trajectories leading to these points violate the transversality conditions which every optimal trajectory must necessarily satisfy.

G Proof of Lemma 5

Assume that there is an optimal investment schedule for which $c \to 0$ and $k \to \infty$ as $t \to \infty$.

The instantaneous profit function of the firm reads as

$$\pi = (1 - q - c)q - k^2.$$
(G.1)

As $c \to 0$ along the trajectories considered, they sooner or later enter the region with positive production, where we have q = (1 - c)/2. Substituting this expression into (G.1), we obtain:

$$\pi = \frac{(1-c)^2}{4} - k^2. \tag{G.2}$$

As $c \to 0$, $(1-c)^2/4$ approaches its upper bound of 1/4. The second term in the above equation, $-k^2$, decreases beyond all bounds as $k \to \infty$. As $t \to \infty$, there is therefore a time t_0 such that $\pi(t) = 0$ for $t = t_0$, and $\pi(t) < 0$ for all $t > t_0$.

Changing the investment schedule to k(t) = 0 for all $t \ge t_0$ would yield a higher value of total discounted profits Π , contradicting the assumption.

H Proof of Lemma 6

We know that the state-control system with zero production is:

$$\dot{k}(t) = \rho k(t),$$

$$\dot{c}(t) = c(t) (1 - \phi k(t))$$

At k = 0, this reduces to: $\dot{k}(t) = 0$, $\dot{c}(t) = c$. Hence, marginal costs increase to infinity along the exit trajectory as $t \to \infty$. However, as $\lambda = 0$ for $c \ge 1$ along the exit trajectory, the transversality condition (17) is satisfied. Observe that along the exit trajectory k > 0 for 0 < c < 1 and k = 0 for $c \ge 1$. Consequently, it follows from (21) that $\lambda = 0$ at c = 1. Then, from (F.2) we have that indeed $\lambda = 0$ for all $c \ge 1$. Increasing marginal costs along the exit trajectory sooner or later exceed the value of 1, for which the value of $H(c, \lambda)$ becomes 0 (see (F.3)). Hence, the transversality condition (16) is satisfied as well.

I Proof of Corollary 1

In this appendix, we prove that the only candidates for an optimal solution curve are the stable path of the saddle-point steady state and the exit trajectory. In particular, we show that any solution curve of the state-control system, given in (26) and (28) and depicted in Figure 2, starting at a point (c_0, k_0) with $c_0, k_0 > 0$ either (i) ends on the stable path, (ii) ends on the exit trajectory, (iii) gives rise to a control k(t) that goes to infinity, and then satisfies the condition of Lemma 5, or (iv) passes through the line segment $\{(c, k) : c \in (0, 1), k = 0\}$, and is then excluded as an optimal solution by Lemma 4. We note that the vector field defined by the state-control system is always locally Lipschitz, and that therefore the theorem of existence and uniqueness of trajectories through a given initial point holds.

There are two situations, determined by the location of the maximum $(\frac{1}{2}, k^*)$ of the $\dot{k} = 0$ isocline which is the quadratic function $c \mapsto \frac{\phi}{4\rho}c(1-c)$. In the first situation $k^* \geq 1/\phi$, in the second $k^* < 1/\phi$.

If $k^* \ge 1/\phi$, the state-control space $S = \{(c, k) : c > 0, k > 0\}$ can be partitioned in three regions S_1, S_2 , and S_3 . The first set is defined as follows:

$$S_1 = \{ (c,k) : 0 < c < 1, \, k > k^* \},\$$

where $k^* = \phi/(16\rho)$ is the maximum of $c \mapsto \frac{\phi}{4\rho}c(1-c)$. To define the second set, we note that the trajectory γ of the state-control space that passes through the point $(c,k) = (1, \phi/(16\rho))$, when continued backwards in time, necessarily has a second intersection with the line c = 1. The first of these intersections, as time decreases, is denoted $(c,k) = (1,k_*)$, where $0 < k_* < 1/\phi$. Let D be the region bounded by γ and the line c = 1. Then,

$$S_2 = \{(c,k) : c \ge 1, k > 0\} \setminus D.$$

Finally,

$$S_3 = S \setminus (S_1 \cup S_2).$$

From the state-control equations

$$\dot{c} = c(1 - \phi k), \quad \dot{k} = \rho k - \frac{\phi}{4}c(1 - c),$$

and the fact that $k^* > 1/\phi$, it follows that everywhere in S_1 we have $\dot{c} < 0$ and $\dot{k} > 0$. It follows by Lemma 5 that no trajectory in this region can be optimal.

In region S_2 , the state-control equations read as

$$\dot{c} = c(1 - \phi k), \qquad \dot{k} = \rho k.$$

We claim that every trajectory in this region must leave it through the half-line $\ell_1 = \{(c,k) : c = 1, k \ge k^*\}$. Note first that the boundary of S_2 consists of ℓ_1 , the curve γ , the line segment $\ell_2 = \{(c,k) : c = 1, 0 < k \le k_*\}$, and the half-line $\ell_3 = \{(c,k) : c \ge 1, k = 0\}$. As γ and ℓ_3 are parts of trajectories of the state-control system, no trajectories can leave S_2 through them. Moreover, we have $\dot{c} \ge 0$ on ℓ_2 , so exit through

this part of the boundary is also impossible. The remaining possibilities are to leave through ℓ_1 , as claimed, or to remain in S_2 indefinitely.

To show that the latter alternative is impossible, note that since $\dot{k} = \rho k$, any trajectory in S_2 will eventually satisfy $k > 2/\phi$. But then $\dot{c} < -c$, and this implies that eventually c should satisfy c = 1, leaving the region S_2 towards S_1 . As no trajectory in S_1 can be optimal, this now extends to all trajectories in S_2 .

It remains to analyze the trajectories in S_3 . They can leave that region through the line segments ℓ_2 , $\ell_4 = \{(c, k) : 0 < c < 1, k = 0\}$ or $\ell_5 = \{(c, k) : 0 < c < 1, k = k^*\}$, as the remaining parts of the boundary are trajectories of the state-control system, or through the point (1, 0) on the exit trajectory.

Trajectories leaving through ℓ_2 enter S_2 and therefore cannot be optimal. As noted above, trajectories leaving through ℓ_4 are excluded by Lemma 4 from optimality. Trajectories leaving through ℓ_5 enter S_1 and again cannot be optimal. Of all the trajectories leaving S_3 , only those on the exit trajectory are thus candidates for optimal solutions.

It remains to discuss the trajectories that remain in S_3 for all time. By the Poincaré-Bendixon theorem, since S_3 is bounded, the limit set of such a trajectory is either a steady-state point or a closed curve. The latter possibility can be ruled out as the area enclosed by the curve would be invariant (cf. Wagener, 2003). The only steady state in S_3 that can be approached by a trajectory is the saddle, if $k^* > 1/\phi$, or the semi-stable steady state, if $k^* = 1/\phi$, and this shows the result.

If $k^* < 1/\phi$, the situation is much simpler. Define in that case

$$\begin{split} S_1 &= \{(c,k) \,:\, 0 < c < 1, \, k > 1/\phi\},\\ S_2 &= \{(c,k) \,:\, c \ge 1, \, k > 0\},\\ S_3 &= S \backslash (S_1 \cup S_2). \end{split}$$

As above, points in S_1 cannot be optimal as a consequence of Lemma 5; points in S_2 eventually end up in S_1 and are excluded by the same reasoning; and as there are no saddle points in S_3 and the trajectories leaving through ℓ_4 cannot be optimal, the only remaining candidate is the exit trajectory.

J Proof of Lemma 8

From (31), we know that the Hamiltonian for $c \ge 1$ is given by:

$$H(c,k) = k\left(k - \frac{2}{\phi}\right); \tag{J.1}$$

and from Lemma 7, we know that the comparison of the total discounted profits of each two candidate optimal paths amounts to comparing the values of the respective Hamiltonians in the initial point of each respective path. As the Hamiltonian in the case of zero investment and zero production is zero, the indifference point between the stable path and the exit trajectory in the region with zero production must be the point at which the Hamiltonian evaluated along the stable path obtains the value of zero.

$$H(c,k) = k\left(k - \frac{2}{\phi}\right) = 0$$

$$\Rightarrow k = 0 \quad \text{or} \quad k = \frac{2}{\phi}$$

The solution structure (39)-(40) tells us that the stable path in the region with $c \ge 1$ decreases as $t \to -\infty$, assuming that the stable path covers the respective region. Observe that the derivative of the Hamiltonian with respect to k is $\frac{\partial H(c,k)}{\partial k} = 2\left(k - \frac{1}{\phi}\right)$, which is positive for $k > 1/\phi$ and negative for $k < 1/\phi$.

If the stable path enters the region with zero production at all, it enters this region at the point where $k > \frac{1}{\phi}$; this follows directly from equation (28).

The same equation implies that k is decreasing along the stable path as $t \to -\infty$, assuming the stable path enters the zero-production region. The above conclusions lead us to distinguish three cases. First, if the stable path crosses the boundary line c = 1 at $k > \frac{2}{\phi}$, then the value of the Hamiltonian is decreasing along the stable path in the region with $c \ge 1$ as $t \to -\infty$, passes zero at $k = \frac{2}{\phi}$ and is negative afterwards. In this case, the indifference point is the value of the marginal cost $\hat{c} > 1$ that corresponds to the point $(\hat{c}, \frac{2}{\phi})$ on the stable path. Second, if the stable path reaches c = 1 at

exactly $k = \frac{2}{\phi}$, then the indifference point is c = 1. Third, for lower values of k on the stable path at c = 1, the value of the Hamiltonian evaluated along the stable path in the zero-production region is negative, such that the point of indifference (if at all) must be in the region with positive production.