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ON THE MITRA–WAN FOREST MANAGEMENT PROBLEM IN CONTINUOUS TIME¹

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The paper provides a continuous-time version of the discrete-time Mitra–Wan model of optimal forest management, where trees are harvested to maximize the utility of timber flow over an infinite time horizon. The available trees and the other parameters of the problem vary continuously with respect to both time and age of the trees, so that the system is ruled by a partial differential equation. The behavior of optimal or maximal couples is classified in the cases of linear, concave or strictly concave utility, and positive or null discount rate. All sets of data share the common feature that optimal controls need to be more general than functions, *i.e.* positive measures. Formulas are provided for *golden-rule* configurations (uniform density functions with cutting at the ages that solve a Faustmann problem) and for *Faustmann policies*, and their optimality/maximality is discussed. The results do not always confirm the corresponding ones in discrete time.

KEYWORDS: Optimal harvesting problems, Forest Management, Measure-valued Control.

JEL CLASSIFICATION: C61, C62, E22, D90, Q23.

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1. INTRODUCTION

Although forest economics has a centuries-long history (see, *e.g.* Samuelson, 1995), the first complete formulation of the forest management problem as a Ramsey-like optimal control model in discrete time is contained in two papers by T. Mitra and H. Wan of the early eighties (Mitra and Wan, 1985, 1986). The authors there discuss the long run structure of the cutting/replanting strategy that maximizes, over an infinite horizon, the sum of utilities generated by the flows of timber obtained by harvesting the trees of a forest. In the basic model, the forest comprises trees of different ages, up to a maximum age, that are cultivated on a unit piece of land that cannot be transferred to other uses, a productivity function gives the amount of wood that is obtained harvesting a tree of any given age, cutting and replanting costs are zero, and new saplings are immediately replanted on the cleared land. The main results are: 1) that the *Faustmann policy* (*i.e.*, cutting trees that reach that age maximizing the present value of bare forest land subject to an infinite sequence of planting cycles) is optimal when the utility function is linear, which implies a cycle in the configuration of the forest, 2) that optimal trajectories converge to the golden rule configuration (the uniform forest with the maximum sustainable yield) when the utility function is strictly concave and the discount factor is equal to 1, and 3) that cycles of the optimal trajectory reappear whenever future utility is discounted even if the utility function is strictly concave. Following this lead, almost the entire theory of optimal forest management has been developed in terms of discrete time (see Tahvonen, 2004 and Khan and Piazza, 2012 for recent lists of the extensions of the model) while, to our knowledge, a consistent continuous time version has never entered the literature.

The two usual justifications for the choice of discrete time – that, (i) transactions do not occur continuously, and (ii) sometimes a “natural period” can be found for agricultural products (Foley, 1975) – have some appeal in forestry, but on balance there seem not to be compelling reasons to assume that transactions in timber markets are synchronized. Moreover, forestry tasks (*e.g.*, tree felling, timber extraction, etc.) require time and therefore it is likely that these operations overlap not only for different agents but also for a single forestry firm. In addition, as it is the case with other natural resources, forests grow continuously and a single natural period for cutting and replanting activities can hardly be identified in general (*e.g.*, at the tropics forests are cut continuously). Hence, no specific economic reason dictates that the discrete time should be preferred to the continuous time framework. If so, it is the technical complexity continuous time models bear that explains the lack of contributions in optimal forest management. Indeed, in elaborating Wicksell’s classical model of natural aging process, Cass (1973) advised that continuous time leads to difficulties in the derivation of efficiency prices.

Some of the additional hurdles brought about by the formulation in continuous time of the Mitra-Wan model (as well as other vintage capital models) are the following: 1) the ages of capital goods (*i.e.* the ages of trees) vary continuously, so that the evolution in time of the state of the system is described by means of a partial differential equation; 2) the control appears also in the boundary condition (giving rise to a *boundary* control system); 3) candidate optimal controls are objects like Dirac’s deltas, hence more general than measurable functions, that is, *measures*. We underline the fact that – contrary to what happens in other vintage capital models – the control appearing in the evolutionary

equation (the so called distributed control) *need* be a measure as a consequence of 3), so that the equation itself cannot be interpreted (pointwise) in \mathbb{R} as usual, and calls for an extended formulation to make sense. All these features make the problem mathematically challenging and, differently from that in discrete time, not a straightforward multi-sectoral generalization of the Ramsey model (for a similar conclusion see Khan and Piazza, 2011).

There have been various attempts to gain insight into the solution of the continuous time Mitra-Wan model, either by studying stripped down versions of the full model or by adding further assumptions that somehow simplify the analysis. Heaps (1984) (see also Heaps, 2006) for example, presented a model in which only the oldest trees can be harvested, and proved in such case that the optimality conditions take the form of a delay differential equation¹. Tahvonen and Salo (1999) (see also Tahvonen et al., 2001) on the other hand, studied a model in which time is continuous, but trees are indivisible. This implies that the number of trees grown on the given land is finite, but also that harvesting cannot provide a continuous flow of timber, whereas a sequence of mass points in connection with the jumps in the state variables. In Tahvonen and Salo (1999) storage of wood is then allowed so that the planner can smooth consumption during the time which elapses between two successive jumps. Finally, Salant (2013) analyzed the equilibrium price paths of different vintages of trees, in a simple model in which the forest land may be used in an alternative way, but replanting is not allowed: irreversible deforestation allows to study optimal continuous harvesting/wood consumption without entering the complexities of distributed state variables.

In this paper we develop a new approach to handle the continuous-time Mitra-Wan model, that consists in reformulating the control problem for the partial differential equation as an equivalent problem for a ordinary differential equation in an infinite dimensional space, and in developing *ad hoc* techniques to perform the analysis. In our formulation we need neither to reduce the dimensionality of the problem (as in Tahvonen and Salo, 1999 or Salant, 2013), nor to constrain the controls (as in Heaps, 1984). Although we allow strategies to be measures rather than functions, with the consequence that instantaneous cutting for the forest of any given age is possible (the golden rule configuration is indeed of such type), we require the associated trajectories to be functions, so to avoid mass points. To this extent, it is enough to consider initial distributions of the forest which are square integrable functions and prove, as we do, that consequently the whole trajectory enjoys the same property.

Once the stage is set, two goals are mainly addressed: giving a joint classification of the behavior of optimal and maximal programs in the cases in which the utility is linear, convex, or strictly convex and the discount rate is positive or null; comparing the properties of continuous-time optimal paths with those known in the discrete-time framework (indeed it turns out that there are significant differences). More in detail, we show that:

- a) the analog of golden-rule and *modified* golden-rule configurations is available for the continuous-time model;
- b) modified golden rules are optimal stationary solutions for the discounted model (with

¹In Heaps (2006) it is claimed that cutting only the oldest trees is a property of the optimal policy function of the full fledged model, although no formal proof is provided. For the discrete-time two-age-classes model it is known that cutting the old trees *and* part of the young trees is optimal if the share of land on which old trees are planted is below a given threshold (see for example Tahvonen, 2004). On the other hand, for the continuous-time model discussed in this paper, constraining the control as suggested by Heaps would make unfeasible the shrinking in time of the length of the support of the state variable and thus a convergence result as that in Theorem 5.9 below would not hold.

optimal cutting age M and timber stationary consumption level monotonically not increasing in the rate of discount), while in the undiscounted case the golden rule is maximal when the utility function is linear, and optimal when the utility is strictly concave, provided it is unique;

- c) if the golden-rule configuration is unique, then undiscounted maximal (or optimal) paths exist from any given initial configuration and, provided the utility function is strictly concave, converge over time to the golden rule configuration;
- d) the Faustmann policy is optimal when the utility is linear and the discount positive, is maximal (and not optimal) when the utility is linear and the discount null, it is not optimal when the utility is strictly concave and the discount positive, for initial data in any neighborhood of the optimal steady state. In particular this result contradicts the analog in discrete time.

Thus we contribute extending to continuous time some of the classical results of the theory of optimal economic growth. We show for the undiscounted case (see (c) and (d)) the conclusions by Brock (1970) on the existence and “average” convergence of maximal paths hold also in our framework, and that the results can be strengthened to existence and asymptotic convergence of optimal paths as in Gale (1967) if the utility function is strictly concave (Mitra and Wan, 1986 and Khan and Piazza, 2010 have already shown that the same holds in discrete time). In addition we refine the above results by providing an example in the style of Brock (1970) and Peleg (1973), which proves that optimal paths do not exist in the linear case. On the other hand, comparative statics results under (b) are specific to the continuous-time setting (hints for such results are in Samuelson (1995) see Figure 2 on page 133 and the first paragraph on page 134) and have not counterpart in the discrete time forestry literature. It is interesting to note that monotonicity of the modified golden rule consumption may not hold in models with several capital goods like ours, while it holds in the one sector Ramsey discounted model (see for example Mas-Colell et al., 1995, pages 758–9). Regarding the Faustmann policy (see (d)) we prove that, similarly to what occurs in the discrete time setting (see Mitra and Wan, 1985), discounting does not affect the structure of the optimal policy when the utility function is linear. However, when the utility function is strictly concave, we show that the periodic optimal solution is not optimal, contrary to what is distinguishing of the discounted discrete-time model (Mitra and Wan, 1985 page 265 and Salo and Tahvonen, 2003 Proposition 1).

We already mentioned that the mathematics underneath the problem is challenging, due to the control in the boundary condition and to the fact that also the distributed control need be a distribution. While the literature on distributed control systems is wide and (also theoretically) well established, that on boundary control systems is not as much, even when described by a simple linear age-structured equation as that in our problem, as the presence of the control in the boundary condition yields discontinuities in the control operator and in the Hamiltonians which are difficult to handle. The technique of functional analysis of rephrasing the problem in a space of functions and of later extending it to a space of distributions (see Section 3) is performed in some theoretical and applied works. It was first introduced in the economic literature by Barucci and Gozzi (1998, 2001) for a problem of optimal investment with vintage capitals, and then studied in various works, under the point of view of theoretical Dynamic Programming (Faggian, 2005, 2008, with finite horizon; Faggian and Gozzi, 2010, with infinite horizon) and that of applications

(Feichtinger et al., 2003, 2006; Faggian and Gozzi, 2004). We mention also the papers by Barucci and Gozzi (1999) and Faggian and Grosset (2013) with application of such techniques to optimal advertising. Nevertheless, in none of these works the control space need be a space of distributions (more often it is the space of square integrable functions). Thus, also the fact that our stationary optimal control is a distribution constitutes a new development in the literature.

We strongly expect that the innovations here presented will find application in vintage capital models and in other models comprising age distributed state variables (*e.g.*, demographic models). For example, having controls that are (not functions but) positive measures may prove useful both in the analysis of the problem of endogenous scrapping of a machine and in the study of the investment profile (*i.e.*, whether the investment is spread over a set of vintages or not, see Feichtinger et al., 2006). Similarly, having controls that can be concentrated on single ages may help in developing theoretical models of the determinants of the optimal retirement age for not stationary populations endowed with a rich demographic structure (Chan and Guo, 1990; Heijdra and Romp, 2009). We also expect that our results and methods find useful applications in the analysis of other models with age distributed natural resources, for example age distributed fisheries (see Tahvonen, 2009) and the so called *orchard model* (Mitra et al., 1991).

Our results will have a more direct bearing on the analysis of the continuous-time Ramsey version of the clay-clay vintage capital model of Solow et al. (1966) developed by Boucekkine et al. (1997) (see also Boucekkine et al., 1998; Hritonenko and Yatsenko, 2008), where a Faustmann-like age shows up as the steady state age at which old machines became obsolete. There, like for forest management, the Faustmann-like policy that replaces all stocks of capital goods that reach the critical age, while preserving full employment of labor, turns out to be optimal in a neighborhood of the steady state for the case of a linear utility function. For this model, our control space will allow the handling of reversible investment, that is beyond the reach of current theory.

Moreover, we observe that in the Ramsey vintage capital model with labor augmenting technical progress an exact golden rule path exists if the utility function is isoelastic and is optimal - or maximal - when the discount rate equals the Mirrlees-Brock-Gale critical discount rate. For this case, our strong convergence result in the strictly concave case provides a formal proof of the fact that the cycles induced by replacement echoes are dampened in the long run by the force of consumption smoothing provided utility is not linear. In addition, the same result should pave the way to the general turnpike result for the discounted model that the literature envisages (see Boucekkine et al., 1998, 2011). In connection with this, a more complete integration of the two models can be expected, with the optimal cyclical paths that arise in the discrete-time discounted forestry model (with the radii of the cycles converging to zero with the discount rate, see Tahvonen et al., 2001; Salo and Tahvonen, 2003; Dasgupta and Mitra, 2011) turning to be optimal for the discrete-time version of the Solow et al. vintage capital model and the asymptotic convergence result of the vintage capital model extended to the strictly concave continuous-time discounted Mitra-Wan model.

The paper is organized as follows. In Section 2 we describe the model in continuous time, in Section 3 we rephrase it into abstract terms and introduce some useful notation, besides the formal definition of optimal and maximal strategies. In Section 4 first we build

the modified golden rules and Faustmann Policies. In Section 5, we classify in four different subsections the behavior of optimal trajectories, according to the fact that the utility is linear, strictly concave (or general concave, in some cases), and that the discount rate is positive or null. We contextually establish whether Golden Rules and Faustmann policies constitute optimal programs, maximal programs, or none of the two. In Section 6 we draw the conclusions. A rich Appendix, following the references section, completes the work with the proofs of the many theorems and all auxiliary technical results.

2. THE CONTINUOUS TIME MODEL

A forest of unitary extension is described in terms of a density $x(t, s)$ which represents the part of the forest covered at time t by trees of a certain age s , with $t \geq 0$, $s \geq 0$, and trees reaching a maximum (finite) age S . Starting from an initial density $x_0(s)$, trees grow in time and may be harvested with a certain cutting rate $c(t, s)$, performed at time t on trees of any age s , and chosen by the optimizer. Harvested trees are instantaneously replaced by new saplings. The evolution of the system is described by the following transport equation

$$(1) \quad \begin{cases} \frac{\partial x}{\partial t}(t, s) = -\frac{\partial x}{\partial s}(t, s) - c(t, s) & t > 0, \quad 0 \leq s \leq S \\ x(t, 0) = \int_0^S c(t, s) ds & t \geq 0 \\ x(0, s) = x_0(s) & 0 \leq s \leq S \end{cases}$$

where the variation of density $\frac{\partial x}{\partial t}(t, s)$ is due to ageing of trees $-\frac{\partial x}{\partial s}(t, s)$, and to harvesting $-c(t, s)$. Note that in the boundary condition the quantity $x(t, 0)$ of saplings of age zero at time t is assumed to coincide with the total amount of trees (of different ages) cut at time t , represented by $\int_0^S c(t, s) ds$.² In addition we require the strategy-trajectory couples (c, x) to satisfy some non negativity constraints, that is

$$(2) \quad c(t, s) \geq 0, \text{ and } x(t, s) \geq 0, \quad \forall t \geq 0, \quad 0 \leq s \leq S$$

implying that only positive quantities are cut, and that the quantity of trees of all ages remains positive in time.

Some remarks are here due. First of all, note that $x(t, s)$ does not represent a spatial density. As a consequence, it may be imagined that trees grow far from one another, and not reciprocally interfering. Moreover, since the size of the forest is normalized to 1 at initial time, that is $\int_0^S x_0(s) ds = 1$, $\int_{\sigma_1}^{\sigma_2} x(t, s) ds$ may be interpreted as the percentage of the total forest which is covered at time t by trees of age between σ_1 and σ_2 . Thus, as a consequence of the boundary condition, the total surface of the forest is covered in time by the constant amount 1 of trees of different ages (see Proposition 3.4), that is

$$\int_0^S x(t, s) ds \equiv \int_0^S x_0(s) ds = 1, \quad \forall t \geq 0.$$

²A dimensional interpretation of the boundary condition is the following. The quantity $x(t, 0)ds$ represents the infinitesimal portion occupied at time t by trees of age 0, which has to be equal to the amount of trees (of all ages) cut at time t , represented by the quantity $\left[\int_0^S c(t, s) ds \right] dt$. The condition then follows by observing that time and age vary jointly, that is $dt/ds = 1$.

Now let $f(s)$ represent the *productivity* of a tree of age s . Summing all the contributions $f(s)c(t, s)ds$ of wood of different ages s harvested at time t , we obtain the total wood $w(c(t))$ harvested at time t , that is

$$(3) \quad w(c(t)) = \int_0^S f(s)c(t, s)ds.$$

Note that f is the continuous version of the timber-content function f defined by Mitra and Wan (1985, 1986). We assume also that f has support contained in $(0, \bar{s})$, with $\bar{s} < S$, meaning in particular that trees of age greater than \bar{s} are considered unproductive. Note also that it would make a little difference to assume trees eternal (that is, $S = \infty$), since the action takes place in $[0, \bar{s}]$ as a consequence of the choice of the productivity function.

Example 2.1 An example of (regular) productivity function f is the following

$$f(s) = \begin{cases} \exp \left[\left(s - \frac{M}{2} \right)^{-1} \left(s - \frac{3M}{2} \right)^{-1} \right] & s \in (M/2, 3M/2) \\ 0 & s \in [0, M/2] \cup [3M/2, +\infty). \end{cases}$$

Note that trees younger than $M/2$ and older than $3M/2$ are considered unproductive, while the productivity increases towards M and decreases afterwards. \square

We anticipate that in the next section, where the problem is formalized, the analysis is restricted to a set of admissible controls $c(t, s)$ which are null, and leave trajectories $x(t, s)$ null, for all $s \geq \bar{s}$. To simplify the mathematical work, we take into account initial data $x_0(s)$ which are also null for all $s \geq \bar{s}$. Those assumptions are justified by the fact that, with zero productivity for $s \geq \bar{s}$, one expects optimal trajectories yielding zero trees older than \bar{s} .

Next we introduce a *utility function* u , which is assumed bounded below, increasing and concave (possibly linear)³, and an overall utility $U(c)$ defined in terms of u as

$$U(c) = \int_0^{+\infty} e^{-\rho t} u(w(c(t))) dt.$$

The problem is maximizing in a suitable sense the overall utility $U(c)$, over the set of admissible strategies, with or without discount ($\rho > 0$ or $\rho = 0$, respectively). Note that when $\rho > 0$ the concavity of u implies the finiteness of $U(c)$, while when $\rho = 0$, $U(c)$ may indeed be infinite valued, and this has to be taken into account when choosing a suitable definition of optimality. Denoting with U_T the overall utility at a finite horizon T , that is

$$U_T(c) = \int_0^T e^{-\rho t} u(w(c(t))) dt,$$

we say that a control strategy \tilde{c} *catches up to* a control strategy c if

$$\forall n \in \mathbb{N}, \exists T_n > 0 : T \geq T_n \Rightarrow U_T(\tilde{c}) > U_T(c) - \frac{1}{n}.$$

For a given initial stock x_0 , an admissible control strategy c^* is said to be *optimal at* x_0 if it catches up to every control strategy c admissible at the same initial stock x_0 . Then, an

³For instance, $u(r)$ may be the identity function, or $\ln(r+1)$, or $r^{1-\sigma}$, with $0 < \sigma < 1$.

optimal control c^* yields definitively (namely, for a sufficiently large horizon T) a greater utility than any other control c starting at the same x_0 except for a (small) difference $\frac{1}{n}$. Note that the above definition of optimality is equivalent to require that

$$\liminf_{T \rightarrow \infty} (U_T(c^*) - U_T(c)) \geq 0,$$

for every control strategy c , admissible at the initial stock x_0 .

If the utility function fails to be strictly concave, then optimality defined this way will prove a too strong requirement, meaning that in some cases under study no control matching such definition will be available, and a weaker optimality property will have to be taken into account. Then, we say that an admissible control strategy c^* is *maximal at x_0* if, given any other control c admissible at x_0 , one has

$$\forall n \in \mathbb{N}, \forall T > 0, \exists T_n > T : U_{T_n}(c^*) > U_{T_n}(c) - \frac{1}{n},$$

meaning that any maximal control c^* yields repeatedly (at a $T_n > T$, for increasing values of T) a greater utility than any other control c starting at the same x_0 except for a difference $\frac{1}{n}$. Note that optimality of a control implies maximality, but the viceversa is false in general. Moreover a control c^* is maximal at x_0 if and only if, for every control c admissible at x_0

$$\limsup_{T \rightarrow \infty} (U_T(c^*) - U_T(c)) \geq 0.$$

Maximality is a more flexible idea of optimality, as it allows for controls engendering fluctuating behaviors of the overall utility U_T in time. Comparison of non convergent infinite horizon integrals (or sums) appear in the optimal growth literature since Ramsey (1928), but the terminology is not consistent across the different papers. We follow the terminology used by McKenzie (1986), page 1286, and recall that the notion of optimal (catching up) controls we use was formalized by Von Weizsäcker (1965) in continuous time to study the existence of optimal programs in the aggregative model of growth, while Gale (1967) and McKenzie (2009) extensively studied optimal paths for the multisector optimal growth model in discrete time. Maximality of controls in the acception cited above was introduced by Brock (1970) where existence of maximal programs for the n sectors optimal growth model in discrete time is established, while Halkin (1974) adapted the concept to continuous time. Extensions in continuous time of the results by Brock (1970) are also due to Brock and Haurie (1976), Carlson et al. (1987), and Zaslavski (2006).

Finally, we remark that, mathematically speaking, the challenging issues go beyond the presence of the control in the boundary condition. Here is an example: the state equation has a solution which can be written easily by means of the characteristic method, as long as the control is an integrable function, given by

$$(4) \quad x(t, s) = \begin{cases} x_0(s-t) - \int_0^t c(t-\tau, s-\tau) d\tau & s \geq t \\ \int_0^s c(t-s, r) dr - \int_0^s c(t-\tau, s-\tau) d\tau & 0 \leq s < t. \end{cases}$$

Unfortunately, the space of admissible controls cannot be a space of functions, but need be a larger space – a space of measures, where optimal controls are shaped like Dirac’s Deltas – in which a formula like the one above is not defined. Or needs to be redefined, as we explain in the next section.

3. THE ABSTRACT PROBLEM

We now rephrase the problem in Section 2 by means of semigroup theory, and formalize assumptions accordingly. The original problem for the partial differential equation is rewritten as a problem for an ordinary differential equation, set in an infinite dimensional space. Roughly speaking, rather than considering the state and the control as real functions of t and s , one sees them as functions of the only variable t , taking values in some space of functions of variable s (e.g. the space of square integrable real functions): $x(t)$ and $c(t)$ are interpreted as the functions of variable s defined by $x(t)(s) \equiv x(t, s)$ and $c(t)(s) \equiv c(t, s)$. The initial condition is written as $x(0) = x_0$, the state equation as $x'(t) = Ax(t) + Bc(t)$ where x' denotes the time derivative, the differential operator $A = -\partial/\partial s$ is a linear operator between space of functions, while the control operator B represents the joint action of the control appearing both in the partial differential equation and in the boundary condition, namely $Bc(t)(s) = -c(t)(s) + [\int_0^S c(t)(s)ds]\delta_0$, where δ_0 is the Dirac's delta at 0. We mention that usually the state/control spaces for such problems is $L^2(0, S)$, the set of real square integrable functions of variable s , considered as a Hilbert space by means of the scalar product $\langle \phi, \psi \rangle_{L^2} = \int_0^S \phi(s)\psi(s) ds$, for ϕ, ψ in $L^2(0, S)$. Nonetheless we need to generalize the problem to a larger space D' (the dual of some D space contained in $L^2(0, S)$, see 3.1) which contains objects like the Dirac's deltas, as mentioned at the end of Section 2, as optimal strategies will be of that kind (stationary programs - the so called *golden rules*, see Section 4 - in the first place).

We advise that a full understanding of the following Subsection 3.1 requires some familiarity with semigroup theory, and may be largely skipped at a first reading. We refer the reader to Engel and Nagel (1999) or Pazy (1983) for the general theory of strongly continuous semigroups, and to Bensoussan et al. (2007) for optimal control in infinite dimension.

3.1. *The extended framework*

We introduce some useful notation. Let X be a Banach space, X' its dual space, we denote by $\langle \cdot, \cdot \rangle_{X', X}$ or simply by $\langle \cdot, \cdot \rangle$ the duality pairing. If $-\infty \leq \sigma_1 < \sigma_2 \leq \infty$, we denote by $L^p(\sigma_1, \sigma_2; X)$, or simply $L^p(\sigma_1, \sigma_2)$ when $X = \mathbb{R}$, the space of function with integrable p -norm, from $[\sigma_1, \sigma_2]$ (or $[\sigma_1, +\infty)$, when $\sigma_2 = +\infty$) to X . We write $H^1(\sigma_1, \sigma_2)$ for the space of functions of $L^2(\sigma_1, \sigma_2)$ with (weak) derivative in $L^2(\sigma_1, \sigma_2)$. We also denote by $L^2_{loc}(\sigma_1, \sigma_2; X)$ X -valued functions from $[\sigma_1, \sigma_2]$ which are square integrable on every compact interval contained in $[\sigma_1, \sigma_2]$, and with $L^\infty(\sigma_1, \sigma_2; X)$ X -valued functions having bounded essential supremum in $[\sigma_1, \sigma_2]$. If $k \in \mathbb{N} \cup \{\infty\}$, then $C^k([\sigma_1, \sigma_2]; X)$ (or simply $C^k([\sigma_1, \sigma_2])$ when $X = \mathbb{R}$) is the space of functions of class C^k from $[\sigma_1, \sigma_2]$ to X .

When rephrasing the model in abstract terms, an intermediate step is formulating the problem in $L^2(0, S)$, making use of the translation semigroup $\{T(t)\}_{t \geq 0}$ on $L^2(0, S)$, namely linear operators $T(t) : L^2(0, S) \rightarrow L^2(0, S)$ such that $[T(t)f](s) = f(s - t)$, if $s \in [t, S]$, and $[T(t)f](s) = 0$ otherwise. The generator of $T(t)$ is the operator $A : D(A) \rightarrow L^2(0, S)$ where $D(A) = \{f \in H^1(0, S) : f(0) = 0\}$, given by $[Af](s) = -\partial f(s)/\partial s$. The adjoint of A is then $A^* : D(A^*) \rightarrow L^2(0, S)$ with $D(A^*) = \{f \in H^1(0, S) : f(S) = 0\}$ defined by $[A^*f](s) = \partial f(s)/\partial s$, generating itself a translation semigroup $T^*(t) : L^2(0, S) \rightarrow L^2(0, S)$ given by $T^*(t)f(s) = f(s + t)$, if $s \in [0, S - t]$, and $T^*(t)f(s) = 0$ otherwise.

The second step is generalization of all previous notions to a wider space. We set

$$D \equiv D(A^*), \quad D' \equiv D(A^*)',$$

and assume D' is both the control space and the state space of the abstract problem. Indeed by standard arguments, in particular by replacing the scalar product in L^2 with the duality pairing $\langle \phi, \psi \rangle_{D', D}$ with $\phi \in D'$, $\psi \in D$, the semigroup $\{T(t)\}_{t \geq 0}$ can be extended to a strongly continuous semigroup on D' (with respect to the operator norm on D'), while $T^*(t)$ can be restricted to a strongly continuous semigroup on D . The generators of such semigroups are respectively an extension and a restriction of the ones in $L^2(0, S)$. For simplicity we keep denoting the semigroups and their generators by $T(t)$ and A , and by $T^*(t)$ and A^* respectively. For details we refer the reader to Faggian (2005) and Faggian and Gozzi (2010). The role of $L^2(0, S)$ remains that of pivot space between D and D' , namely $D \subset L^2 \subset D'$, with the norm in D' dominating the L^2 -norm, as the duality pairing coincides with the scalar product when $\phi \in L^2(0, S)$, namely

$$\langle \phi, \psi \rangle_{D', D} = \langle \phi, \psi \rangle_{L^2}, \quad \psi \in D, \quad \phi \in L^2(0, S).$$

We use the notation $\langle \cdot, \cdot \rangle$ in both cases, unless it is ambiguous. It is very important to say that such formulation enables the possibility of choosing controls which are positive measures rather than functions.⁴ More precisely, we make use of a subset of D' , that of (positive) Radon measures \mathcal{R} on $[0, S]$, endowed with the norm $|c|_{\mathcal{R}}$ (the finite measure of the set $[0, S]$ with respect to c), for all $c \in \mathcal{R}$. The space \mathcal{R} contains in particular all Dirac's measures δ_{s_0} , with $s_0 \in [0, \bar{s}]$. Moreover, since $\mathcal{R} \subset D'$ with continuous inclusion, the D' -norm is dominated by the \mathcal{R} -norm.⁵

We also denote by $\text{supp}(g)$ the support of any function/measure $g \in \mathcal{R}$. We define the cut-off function $\psi \in C^\infty([0, S]; \mathbb{R}^+)$ such that for fixed s_1, s_2 , with $\bar{s} < s_1 < s_2 < S$,⁶

$$(5) \quad \psi \equiv 1 \text{ on } [0, s_1], \quad \psi \equiv 0 \text{ on } [s_2, S], \quad \psi \text{ decreasing on } [s_1, s_2].$$

Observe that $\psi \in D$. Then, when c has support in $[0, \bar{s}]$, by means of the linear continuous functional $L: D' \rightarrow \mathbb{R}$, $c \mapsto Lc := \langle c, \psi \rangle$ we may write the boundary condition as $x(t, 0) = Lc(t) = \langle c(t), \psi \rangle$, and moreover enclose the boundary condition in the control operator B as follows (the technique is standard)

$$(6) \quad B: D' \rightarrow D', \quad Bc := -c + (Lc) \delta_0 = -c + \langle c, \psi \rangle \delta_0$$

where δ_0 is the Dirac's delta at 0.

Remark 3.1 It is easy to verify that B^* , the adjoint operator of B , is given by

$$B^*: D \rightarrow D, \quad \text{with } B^*v := -v + \langle \delta_0, v \rangle \psi.$$

Note also that the cut-off function ψ belongs to the set

$$(7) \quad D^2 \equiv D(A^{*2}) = \{g \in D : g' \in D\} = \{g \in H^2(0, S) : g(S) = g'(S) = 0\},$$

namely, the domain of the generator of the adjoint semigroup $T^*(t)$ restricted to D . \square

⁴More precisely, as a function of t , the control $c(\cdot)$ is chosen (positively) measure-valued.

⁵If $C^0 \equiv C^0([0, S])$, then $\mathcal{R} = (C^0)'$ as, by Riesz Theorem, c may be represented as a positive linear functional on C^0 , with norm $|c|_{\mathcal{R}} = \sup_{\phi \in C^0} \langle c\phi \rangle / |\phi|_{C^0}$ (see Rudin, 1987, Theorem 2.14, p.40). Since $D \subset C^0([0, S])$, then $\mathcal{R} \subset D'$ (both continuous inclusions) and there exists $\alpha > 0$, $|c|_{D'} \leq \alpha |c|_{\mathcal{R}}$, for any $c \in \mathcal{R}$.

⁶For instance, $\psi \equiv 1$ on $[0, \bar{s} + \frac{S-\bar{s}}{4}]$, and ψ decreasing on $[\bar{s} + \frac{S-\bar{s}}{4}, \bar{s} + \frac{3}{4}(S-\bar{s})]$.

Summing up, the *state equation* (1) can be written as

$$(8) \quad \begin{cases} x'(t) = Ax(t) + Bc(t), & t > 0 \\ x(0) = x_0 \end{cases}$$

and rewritten in mild form (see *e.g.* Bensoussan et al., 2007 Section 3.II.1) as

$$(9) \quad x(t) = T(t)x_0 + \int_0^t T(t-\tau)Bc(\tau) d\tau.$$

Moreover we assume

$$(10) \quad f \in D, f \geq 0, \text{ and } \text{supp}(f) \subset (0, \bar{s})$$

implying in particular that f is continuous, null in $[0, \lambda]$ for a $\lambda > 0$, and

$$(11) \quad u \in C^1(\mathbb{R}^+, \mathbb{R}^+) \text{ and concave.}$$

Since (3) may be rewritten as $w(c(t)) = \langle c(t), f \rangle$, coinciding with (3) when c is in $L^2(0, S)$, the *objective functional* U_T , whenever finite, may be written as

$$(12) \quad U_T(c) = \int_0^T e^{-\rho t} u(\langle c(t), f \rangle) dt, \quad 0 \leq T \leq +\infty.$$

3.2. Admissible controls and initial data

We denote a trajectory starting at x_0 and driven by a control c as $x(\cdot; x_0, c)$ or $x_{x_0, c}(\cdot)$. Although the abstract problem is set in D' , we assume for technical reasons that x_0 is a function in $L^2(0, S)$ (and not a measure), aware of the fact that initial densities where the mass is concentrated at certain ages do not match the requirement. Moreover, we assume that x_0 is compactly supported in $[0, \bar{s}]$, where \bar{s} is the age above which trees will be considered unproductive (see (10)). Such assumptions allow a simplification of the mathematics of the problem, as one expects an optimal trajectory starting from such initial data to preserve the property of being null at ages greater than \bar{s} . Accordingly, the an admissible control $c(t)$ is a positive distributions in D' , which is null for $s \geq \bar{s}$ and yields trajectories which are null for $s \geq \bar{s}$. These requirements need to be formally stated, as we do next.

Initial data. Initial densities x_0 are chosen in the set

$$(13) \quad \Pi := \left\{ x \in L^2(0, S) : x \geq 0, \text{supp}(x) \subseteq [0, \bar{s}], \int_0^{\bar{s}} x(s) ds = 1 \right\}.$$

Admissible control strategies. The set \mathcal{U}_{x_0} of control strategies admissible at x_0 is

$$(14) \quad \mathcal{U}_{x_0} := \left\{ c \in L^2_{loc}(0, +\infty; D') : \begin{array}{l} \text{supp}(c(t)), \text{supp}(x(t)) \subseteq [0, \bar{s}] \quad \forall t \geq 0 \\ c(t) \text{ and } x(t; x_0, c) \text{ lie in } \mathcal{R}, \quad \forall t \geq 0 \end{array} \right\}$$

Remark 3.2 Note that the condition “ $c(t)$ and $x(t; x_0, c)$ lie in \mathcal{R} ” in (14) translates the non-negativity constraints (2) in terms of measures. It also allows the trajectory associated to an admissible control to be a measure rather than a function, although $x(t; x_0, c)$ is

proven to be a function *ex post*, in Proposition 3.3, for initial data x_0 in $L^2(0, S)$ and controls in \mathcal{U}_{x_0} . Moreover $c \in L^2_{loc}(0, +\infty; D')$ implies that (and this is a mathematical technicality) although controls are \mathcal{R} -valued, their integrability is required with respect to the D' -norm, which is less restrictive than integrability with respect to the \mathcal{R} -norm, as the second dominates the first. \square

Proposition 3.3 *Consider an initial datum x_0 in Π and a control $c \in \mathcal{U}_{x_0}$. Then there exists a unique solution $x(\cdot; x_0, c)$ of (9) and it belongs to $C^0([0, +\infty); D')$. Moreover, for any $t \in [0, +\infty)$, $x(t)$ belongs to $L^2(0, S)$.⁷*

We can also formally prove that the surface of the land covered with trees stays constant in time, as stated in the next proposition.

Proposition 3.4 *Assume $x_0 \in \Pi$. Then the trajectory $x = x(\cdot; x_0, c)$ of (9) satisfies*

$$\langle x(t), \psi \rangle \equiv \langle x_0, \psi \rangle \text{ for all } t \geq 0, \text{ for all } c \in \mathcal{U}_{x_0},$$

that is

$$(15) \quad \int_0^{\bar{s}} x(t, s) \, ds = \int_0^{\bar{s}} x_0(s) \, ds = 1.$$

In some cases we need to consider a restricted class of admissible controls. If $\lambda > 0$, we set

$$(16) \quad \mathcal{U}_{x_0}^\lambda := \left\{ c \in L^\infty(0, +\infty; \mathcal{R}) : \begin{array}{l} \text{supp}(c(t)) \subseteq [\lambda, \bar{s}], \text{ supp}(x(t)) \subseteq [0, \bar{s}], \forall t \geq 0 \\ c(t), x(t; x_0, c) \in \mathcal{R}, \forall t \geq 0 \end{array} \right\}$$

Note that here the controls are bounded in the \mathcal{R} -norm. Note also that assuming $\text{supp}(c^*(t)) \subseteq [\lambda, \bar{s}]$ seems natural, as by (10) any optimal control $c^*(t)$ is expected to satisfy $\text{supp}(c^*(t)) \subseteq [\lambda, \bar{s}]$ for almost all $t \geq 0$, for a suitable $\lambda > 0$, even if not required explicitly in the definition of the admissible class \mathcal{U}_{x_0} . A further restriction will be

$$(17) \quad \mathcal{U}_{x_0}^{\lambda, K} := \left\{ c \in L^\infty(0, +\infty; \mathcal{R}) : \begin{array}{l} \text{supp}(c(t)) \subseteq [\lambda, \bar{s}], \text{ supp}(x(t)) \subseteq [0, \bar{s}] \forall t \geq 0 \\ c(t), x(t; x_0, c) \in \mathcal{R}, \quad |c(t)|_{\mathcal{R}} \leq K, \quad \forall t \geq 0 \end{array} \right\}$$

meaning that admissible controls are not only bounded in the \mathcal{R} -norm, but also by means of the same constant K .

3.3. Optimal, Maximal and Stationary programs

Here we formalize the definition of optimal and maximal control.

Definition 3.5 *Given $x_0 \in \Pi$ we say that $c^* \in \mathcal{U}_{x_0}$ is optimal at x_0 if, given any other control $c \in \mathcal{U}_{x_0}$, one has*

$$\liminf_{T \rightarrow \infty} (U_T(c^*) - U_T(c)) \geq 0.$$

and that $c^* \in \mathcal{U}_{x_0}$ is maximal at x_0 if, given any other control $c \in \mathcal{U}_{x_0}$, one has

$$\limsup_{T \rightarrow \infty} (U_T(c^*) - U_T(c)) \geq 0.$$

⁷Note that the solution coincides with the simplified formula (1) given by (4) only if in addition $c \in L^2_{loc}(0, +\infty; L^2(0, \bar{s}))$. Unfortunately, meaningful controls never fall into that class.

The trajectory x^* associated to an optimal (respectively, maximal) control c^* is said to be an optimal (maximal) trajectory, while the couple (x^*, c^*) is said to an optimal (maximal) couple or program.

A last definition completes the abstract framework, that of stationary program. When the control is a function, a stationary program may be defined as a couple (x, c) such that the control is time independent, that is $c(t) \equiv c \in L^2(0, S)$, and the trajectory satisfies

$$(18) \quad x(s) = \begin{cases} x(s-t) - \int_{s-t}^s c(r)dr & s \geq t \\ \int_s^{\bar{s}} c(r)dr & 0 \leq s < t. \end{cases}$$

namely, (x, c) solves (1) with a null derivative of x with respect to t , so that x itself does not depend on time. When c is instead a measure, (18) may be interpreted in the following abstract way.

Definition 3.6 We say that $(\tilde{x}, \tilde{c}) \in \Pi \times \mathcal{R}$ is a stationary couple if, for all $t \geq 0$,

$$(19) \quad \tilde{x} = T(t)\tilde{x} + \int_0^t T(t-s)B\tilde{c}ds.$$

A stationary couple (\tilde{x}, \tilde{c}) is optimal if $c(t) \equiv \tilde{c}$ is optimal at \tilde{x} .

In the following Lemma we characterize stationary couples.

Lemma 3.7 A couple $(\tilde{x}, \tilde{c}) \in \Pi \times \mathcal{R}$ is stationary if and only if $-\tilde{x}$ is weakly increasing and \tilde{c} is its Stieltjes derivative. In this case \tilde{c} is also the derivative of $-\tilde{x}$ in the distributional sense $\langle \tilde{x}, \phi \rangle_{L^2} = \langle \tilde{c}, \phi' \rangle_{D', D}$, for any $\phi \in C^\infty([0, S], \mathbb{R})$ compactly supported in $(0, S)$.

4. THE FAUSTMANN PROBLEM AND CANDIDATE OPTIMAL PROGRAMS

In this section we identify candidates optimal and maximal programs consistently with Mitra and Wan (1985, 1986), only in continuous time. There and here, all candidates are characterized by a cutting age that is obtained solving the *Faustmann problem*, which consists in identifying critical ages maximizing “the present discounted value of all net cash receipts [...] calculated over the *infinite chain* of cycles of planting on the given acre of land from now until Kingdom Come” (Samuelson, 1995 p. 122). The rule “cutting any tree that reaches the critical age” is called the *Faustmann policy* and, clearly, candidates generated by that policy are cyclical. Stationary candidates are also prices supported and are called here *golden rules* or *modified golden rules*, depending on the fact that the discount rate is zero or positive. In the discrete time formulation, Mitra and Wan (1985, 1986) did not use the term modified golden rules for the optimal stationary states of the discounted model, although they called golden rules the stationary states of the undiscounted model, and pointed out that, analogously to what occurs in the Ramsey model, golden rules achieve maximum sustained forest yields.

In continuous time and for $\rho > 0$, the Faustmann problem is identifying maximizers for the function

$$g_\rho(s) = \sum_{n=1}^{\infty} e^{-\rho ns} f(s) = \frac{f(s)}{e^{\rho s} - 1}$$

which can be described as the value of an infinite sequence of planting cycles with harvesting at age s . Note that for f satisfying (10) the following sets coincide and are nonempty

$$\operatorname{argmax} \{g_\rho(s) : s \in [0, \bar{s}]\} = \operatorname{argmax} \{g_\rho(s)(1 - e^{-\rho}) : s \in [0, \bar{s}]\},$$

with $\lim_{\rho \rightarrow 0^+} g_\rho(s)(1 - e^{-\rho}) = f(s)/s$. Therefore the analysis may be extended to the case $\rho = 0$ by maximizing, rather than g_ρ , the function

$$G_\rho(s) = \begin{cases} \frac{1-e^{-\rho}}{e^{\rho s}-1} f(s) & \rho > 0 \\ f(s)/s & \rho = 0. \end{cases}$$

Hence the Faustmann problem becomes identifying

$$\mathcal{A}_\rho \equiv \operatorname{argmax} \{G_\rho(s) : s \in [0, \bar{s}]\}, \quad \forall \rho \geq 0.$$

Maximizers enjoy some interesting properties. In order to study them, once chosen a particular selection M_ρ in \mathcal{A}_ρ , it is useful to define the following *support function*

$$(20) \quad h_\rho(s) = \begin{cases} g_\rho(M_\rho)(e^{\rho s} - 1) & \rho > 0 \\ f(M_0)M_0^{-1}s & \rho = 0. \end{cases}$$

Remark 4.1 Indeed, since $h_\rho(s) \geq f(s)$, for all s in $(0, S]$ and $h_\rho(M_\rho) = f(M_\rho)$, one has

$$(21) \quad \mathcal{A}_\rho = \{s \in (0, S] : h_\rho(s) = f(s)\}.$$

With reference to Figure 1, \mathcal{A}_ρ is the set where the graph of f touches (from below) that of h_ρ . \square

Proposition 4.2 Assume f satisfies (10). Then $\mathcal{A}_\rho \subset (0, \bar{s}]$, and $\mathcal{A}_\rho \neq \emptyset$, for all $\rho \geq 0$. Moreover, if $0 < \rho_B < \rho_A$, then:

- (i) There exists $\tilde{s} \in (0, S]$ such that $\mathcal{A}_{\rho_A} \subseteq (0, \tilde{s}]$ and $\mathcal{A}_{\rho_B} \subseteq [\tilde{s}, S]$. Moreover, \mathcal{A}_{ρ_A} and \mathcal{A}_{ρ_B} may be non-disjoint only if f is not differentiable at \tilde{s} .
- (ii) For any chosen $M_\rho \in \mathcal{A}_\rho$, the selections $\rho \mapsto M_\rho$ and $\rho \mapsto \frac{f(M_\rho)}{M_\rho}$ are nonincreasing. Moreover \mathcal{A}_ρ is not a singleton for at most countable set of values of ρ .
- (iii) For every selection M_ρ of \mathcal{A}_ρ , there exists $\lim_{\rho \downarrow 0^+} M_\rho = m_0$. Moreover $m_0 = \min \mathcal{A}_0$.

Remark 4.3 With reference to Figure 1, if \mathcal{A}_{ρ_B} has a minimum M_{ρ_B} , then B has coordinates $(M_{\rho_B}, f(M_{\rho_B}))$ and, for all $\rho > \rho_B$, $(M_\rho, f(M_\rho))$ lies on the portion of the graph of f delimited by A and B . \square

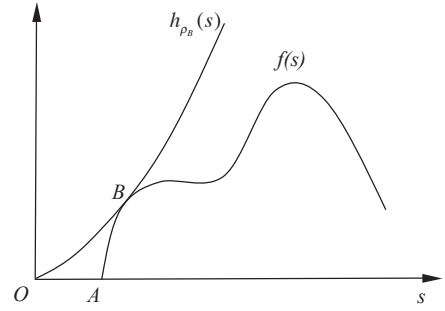


FIGURE 1.— The support function when \mathcal{A}_ρ is singleton. The coordinates of B are $(M_\rho, f(M_\rho))$.

4.1. The Golden Rule

A modified golden rule (x_ρ, c_ρ) (or golden rule, when $\rho = 0$) is a couple in $\Pi \times \mathcal{R}$ so defined

$$(22) \quad x_\rho(s) := \frac{1}{M_\rho} \chi_{[0, M_\rho]}(s),$$

where $M_\rho \in \mathcal{A}_\rho$, meaning that all ages in the range $[0, M_\rho]$ are uniformly distributed and equal to $1/M_\rho$, while those in the range $[M_\rho, S]$ are null, and

$$(23) \quad c_\rho(t, s) \equiv \frac{1}{M_\rho} \delta_{M_\rho},$$

where δ_{M_ρ} is the Dirac Delta at point M_ρ , that is, the action undertaken by c_ρ is cutting exactly the trees reaching age M_ρ . Note that x_ρ is a function in Π (see (13)), and that c_ρ is not a function of s but a positive measure. The choice of the control c_ρ in the problem affects the quantity of wood, by definition of Dirac's Delta, as follows

$$(24) \quad w(c_\rho) = \langle c_\rho, f \rangle = \frac{1}{M_\rho} \langle \delta_{M_\rho}, f \rangle = \frac{1}{M_\rho} f(M_\rho).$$

Note that the set \mathcal{A}_ρ may be not singleton. We will take such fact into account and derive different results accordingly.

It is not difficult to guess that any golden rule is a stationary couple, as the amount of trees cut at age M_ρ is instantaneously replanted at age 0, preserving the distribution among different ages unaltered. Nonetheless the following result identifies completely the shape of stationary couples of the problem.

Proposition 4.4 *Assume $\rho \geq 0$, and f and u satisfying (10) (11) respectively. Consider the trajectory of system (8) starting at x_ρ . Then (x_ρ, c_ρ) is a stationary couple in the sense of Definition 3.6.*

Define now

$$(25) \quad \beta_\rho := \langle c_\rho, f \rangle = \frac{f(M_\rho)}{M_\rho}, \quad \eta_\rho := \frac{f(M_\rho)}{e^{\rho M_\rho} - 1}.$$

and $p_\rho: [0, S] \rightarrow \mathbb{R}^+$ as

$$(26) \quad p_\rho(s) := \begin{cases} \eta_\rho (e^{\rho s} - 1) \psi(s) & \rho > 0 \\ \beta_0 s \psi(s) & \rho = 0. \end{cases}$$

Note that $p_0(s) = \lim_{\rho \rightarrow 0^+} p_\rho(s)$ and that, for any $\rho \geq 0$, p_ρ is twice differentiable with $p_\rho(S) = p'_\rho(S) = 0$, which implies p_ρ is in D^2 .

The dual variables in (26) have a straightforward interpretation as stationary competitive prices associated with a golden rule path (see Cass and Shell, 1976). Indeed, assume we interpret $p_\rho(s)$ as the (infinite dimensional) vector of the prices of capital goods (i.e, the prices of the different vintages s of trees) and set $R = \rho \eta_\rho$ the rent rate of the land on which the trees are planted (when $\rho = 0$, define $R = \lim_{\rho \rightarrow 0^+} \rho \eta_\rho = \beta_0$). Then by definition (26)

$$f(s) \leq p_\rho(s), \quad s \in [0, \bar{s}], \quad f(M_\rho) = p_\rho(M_\rho),$$

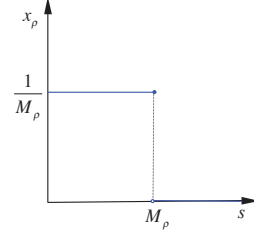


FIGURE 2.— The golden rule stationary path.

where the first inequality means that no cutting process yields a positive profit, while the equality says that the only cutting processes that do not generate losses are those that operates at the Faustmann ages. Thus, the golden rule controls maximizes the short run profits. In addition, since, for all $s \in [0, \bar{s}]$, $p'_\rho(s) = \rho p_\rho(s) + R$ for $M_\rho \geq s \geq 0$ and $p'_\rho(s) \leq \rho p_\rho(s) + R$ for $s \geq M_\rho$, then the asset-market-clearing conditions that hold under competitive arbitrage are satisfied. Clearly, the arbitrage condition in a golden rule takes the form of a “modified Hotelling rule” because a piece of land needs to be rented in order to hold a tree of a given age *in situ*.

4.1.1. Modified golden rules

In the following sections we will classify the behavior of candidate optimal or maximal programs, assuming either the discount ρ is positive or null, the utility function u is linear or strictly concave, \mathcal{A}_ρ is singleton or multivalued. Nonetheless, when $\rho > 0$ is strictly positive, optimality of the golden rule is a general property (holding for a general concave utility u and a possibly multivalued \mathcal{A}_ρ), as stated in the next theorem.

Theorem 4.5 *Assume $\rho > 0$, $M_\rho \in \mathcal{A}_\rho$, and f and u satisfying (10) (11) respectively. Then c_ρ is optimal at x_ρ in the sense of Definition 3.5. Moreover, if $\mathcal{A}_\rho = \{M_\rho\}$, then the unique optimal stationary couple is (x_ρ, c_ρ) .*

Remark 4.6 Note that the first assertion of the previous proposition holds for any choice of M_ρ in \mathcal{A}_ρ . It is easy to prove (see the Appendix for a formal proof) that the same property holds for any convex linear combination of golden rules, that is, if $\mathcal{A}_\rho = \{M_\rho^1, \dots, M_\rho^n\}$, and (x_ρ^i, c_ρ^i) is the golden rule associated to M_ρ^i , then

$$\tilde{x} = \sum_{i=1}^n \lambda_i x_\rho^i, \quad \tilde{c} = \sum_{i=1}^n \lambda_i c_\rho^i,$$

where $\lambda_i \geq 0$, $\sum_{i=1}^n \lambda_i = 1$, is also an optimal stationary program. \square

Remark 4.7 Note that by (24) it is $U_T(c_\rho) = \rho^{-1}(1 - e^{-\rho T})u(\beta_\rho)$, when $\rho > 0$ the golden rule is optimal when starting at x_ρ , with maximal overall utility given by

$$\max_{c \in \mathcal{U}_{x_\rho}} U(c) = U(c_\rho) = \lim_{T \rightarrow +\infty} U_T(c_\rho) = \frac{u(\beta_\rho)}{\rho}.$$

The proofs of Theorem 4.5 and of other theorems in the following sections rely on the construction of the *value-loss function*

$$(27) \quad \theta_\rho(c(t), x(t)) = u(\beta_\rho) - u(\langle c(t), f \rangle) + u'(\beta_\rho) [\rho \langle x(t) - x_\rho, p_\rho \rangle - \langle x(t), A^* p_\rho \rangle + \langle c(t), p_\rho \rangle],$$

which is formally introduced in the Appendix, in the statement of Corollary A.6. This function, which gives the value-loss of any admissible couple at the steady state competitive prices, is the analogous of the value-loss function commonly used for finite dimensional optimal growth problem (see McKenzie, 1986 for the discrete time case and Magill, 1977 for continuous time). The only aspect that is specific to our infinite dimensional setting is that the unit rental costs function contains an element accounting for the ageing process of capital goods. Note that the input-output prices in the value-loss function are expressed in

terms of marginal utilities, while capital goods prices in (26) are given in terms of timber. However, since there is a single final good, expressing the prices in the new numeraire amounts to rescaling them by means of the factor $u'(\langle c_\rho, f \rangle)$. An important consequence of this fact is that golden rules are independent from the instantaneous utility function (see Mitra and Wan, 1985). On the contrary, the form of the instantaneous utility function matters in the analysis of the stationary states of the undiscounted model.

4.2. The Faustmann solution

Besides the golden rule, other controls are candidates to be optimal or maximal when starting at a general initial datum x_0 . Indeed, the golden rule may fail even to be admissible at x_0 . Nonetheless, given an initial datum $x_0 \in \Pi$, if M_ρ represents a preferable cutting age providing a maximal harvesting, one may attempt to use the feedback strategy (\hat{x}, \hat{c}) , where

$$(28) \quad \hat{c}(t) = \hat{x}(t, M_\rho) \delta_{M_\rho}, \quad \forall t \geq 0,$$

that is, \hat{c} cuts existing trees reaching age M_ρ . Such trees vary in time depending on the initial distribution x_0 . We remark that a trajectory of the system starting from an initial datum in Π is in L^2 , as a function of s , so that $\hat{x}(t, M_\rho)$ is not well defined, as well as the control \hat{c} . Nonetheless, in the following lemma we are able to give meaning to both.

Lemma 4.8 *Assume x_0 satisfies $\text{supp}(x_0) \subset [0, M_\rho]$. Set*

$$(29) \quad \hat{x}(t, s) = \hat{x}(t)(s) = x_0(s - \sigma(t)) \chi_{[\sigma(t), M_\rho]}(s) + x_0(s + M_\rho - \sigma(t)) \chi_{[0, \sigma(t)]}(s)$$

where $\sigma(t) = \left\{ \frac{t}{M_\rho} \right\} M_\rho = t - \left[\frac{t}{M_\rho} \right] M_\rho$, $[a]$ and $\{a\}$ denote respectively the integer and the fractional part of the real number a . Then \hat{x} is M_ρ -periodic, that is $\hat{x}(t + M_\rho) = \hat{x}(t)$, for all $t \geq 0$, the control $\hat{c}(t) = \hat{x}(t, M_\rho) \delta_{M_\rho}$ is admissible at x_0 and \hat{x} solves the closed loop equation

$$(30) \quad \hat{x}(t) = T(t)x_0 + \int_0^t T(t - \tau) B \hat{x}(\tau, M_\rho) \delta_{M_\rho} d\tau.$$

We will refer to (\hat{x}, \hat{c}) as to the *Faustmann solution* or *Faustmann Policy*.

Remark 4.9 Note that the golden rule is the Faustmann solution associated to the initial datum x_ρ . \square

Remark 4.10 Note that we require that the support of the initial datum x_0 lies in $[0, M_\rho]$ in order to prove the Lemma. With such an assumption the defined feedback control is admissible, in particular it implies $\int_0^S \hat{x}(t, s) ds = 1$. Indeed, since the solution is M_ρ periodic

$$\begin{aligned} \int_0^S \hat{x}(t, s) ds &= \int_0^{\sigma(t)} \hat{x}(t, s) ds + \int_{\sigma(t)}^{M_\rho} \hat{x}(t, s) ds \\ &= \int_0^t x_0(s - t) ds + \int_t^{M_\rho} x_0(s + M_\rho - t) ds = \int_0^{M_\rho} x_0(r) dr. \end{aligned}$$

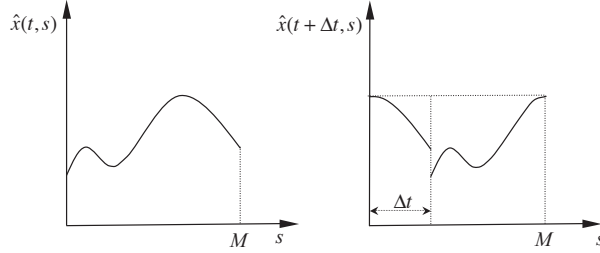


FIGURE 3.— The Faustmann Solution. The effect of cutting at age M_ρ and replanting at age 0 induces cycling of the trajectory in a time length of M_ρ . After a time length $\Delta t < M_\rho$, the graph of the trajectory is translated forward of Δt and the portion exceeding age M_ρ reappears for s in $[0, \Delta t]$.

Lemma 4.11 Assume $\text{supp}(x_0) \subset [0, M_\rho]$, \hat{c} the Faustmann policy, $T \geq 0$, and set $n \equiv \lceil T/M_\rho \rceil$, $\sigma(T) = \{T/M_\rho\} M_\rho$ and

$$U_1^\rho := \int_0^{M_\rho} e^{\rho\tau} u(f(M_\rho)x_0(\tau)) d\tau, \quad U_2^\rho(T) := \int_{M_\rho - \sigma(T)}^{M_\rho} e^{\rho\tau} u(f(M_\rho)x_0(\tau)) d\tau$$

Then

$$(31) \quad U_T(\hat{c}) = \begin{cases} \frac{1 - e^{-\rho n M_\rho}}{e^{\rho M_\rho} - 1} \chi_{[M_\rho, \infty)}(T) U_1^\rho + e^{-\rho(n+1)M_\rho} U_2^\rho(T), & \rho > 0 \\ nU_1^0 + U_2^0(T), & \rho = 0 \end{cases}$$

Remark 4.12 Note that when $\rho > 0$ the overall utility is finite

$$U(\hat{c}) = \lim_{T \rightarrow \infty} U_T(\hat{c}) = U_1^\rho (e^{\rho M_\rho} - 1)^{-1},$$

contrary to the case $\rho = 0$ where it is not. The formula is consistent with those contained in Remark 4.7 when $x_0 = x_\rho$. \square

4.3. Null discounts and Good Controls

Assume $\rho = 0$, $M \in \mathcal{A}_0$ and denote by (\bar{x}, \bar{c}) the associated golden rule, that is

$$\bar{x} = \frac{1}{M} \chi_{[0, M]}, \quad \bar{c} = \frac{1}{M} \delta_M.$$

The case when $\rho = 0$ appears immediately as more complicated than the case of positive discount. For example, the utility over a finite horizon T associated to the golden rule is

$$U_T(\bar{c}) = T u(\beta_0)$$

so that for null discount the utility over an infinite horizon fails to be finite. With null discount the notion of *good controls*, which we give next, will prove useful.

Definition 4.13 Assume $\rho = 0$. A control $c \in \mathcal{U}_{x_0}$ is good if there exists $\theta \in \mathbb{R}$ s.t.

$$\inf_{T \geq 0} (U_T(c) - U_T(\bar{c})) \geq -\theta.$$

We recall that the notion of good controls was introduced by Gale (1967) for the undiscounted n sector optimal growth model in discrete time. Note that a control is defined “good” in comparison to the golden rule \bar{c} , and that such comparison is performed although the golden rule may fail to be admissible at some arbitrary initial datum x_0 . Note also that an equivalent way of giving the definition is to say that a control c is good if

$$\exists \theta \in \mathbb{R} : \forall T \geq 0, U_T(c) \geq U_T(\bar{c}) - \theta.$$

meaning that the utility (over an arbitrary finite horizon T) achieved by means of a good control is dominated by that obtained at \bar{c} by at most a finite quantity θ .

The following proposition compares good and optimal controls.

Proposition 4.14 If $c^* \in \mathcal{U}_{x_0}$ is maximal (and, in particular, optimal) control then it is good.

The previous result allows to seek for optimal or maximal programs in the class of good controls, as no control which is not good may be optimal or maximal.

5. CLASSIFICATION OF OPTIMAL PROGRAMS

5.1. Linear utility, positive discount

In Theorem 4.5 we already established that, when $\rho > 0$, the modified golden rules are optimal in all sets of assumptions. In particular this holds true for u linear, say $u(r) = r$. In the following theorem we establish that, in the particular case of $\rho > 0$ and u linear, the Faustmann solution is an optimal program, that is, the optimal policy is cutting trees reaching age M_ρ , regardless the initial distribution x_0 , as long as x_0 does not contain trees older than the optimal age M_ρ . This is consistent with Theorem 4.5, as the Faustmann solution coincides with the golden rule when the initial datum is x_ρ .

Theorem 5.1 Assume $\rho > 0$, f satisfying (10), and $u(r) = r$, $r \geq 0$. Consider an initial datum x_0 in Π with $\text{supp}(x_0) \subseteq [0, M_\rho]$. Then the Faustmann Solution (\hat{x}, \hat{c}) , given by (29) (28) is optimal at x_0 .

Remark 5.2 The proof may be easily adapted to the case of affine utility $u(r) = ar + b$.

Remark 5.3 Note that, for a wide class of initial data, all those supported in $[0, M_\rho]$, the optimal trajectory is cyclic. As a counterpart, the modified golden rule is a stationary solution – an equilibrium – but not an asymptotic equilibrium. Optimal trajectories do not tend to any stationary solution, except when starting already at it.

Remark 5.4 In proving Theorem 5.1 we establish that the optimal value function is linear. The linearity of the function (*i.e.*, all differences in value from the steady state reduces to the difference in value of the initial forest from the stationary forest) explains the lack of convergence of optimal trajectories to the modified golden rule. It is well known indeed that the clustering of solutions in optimal growth models is driven by second order differences due to strict concavity of the value function. \square

5.2. Linear utility, null discount

In this subsection we are going to show that, with null discount and linear u , the Faustmann solution is maximal but not optimal. Anyway, this is the best one may expect, as one shows that optimal programs do not exist.

Moreover through this and the following subsections we always require the assumption

$$(32) \quad \mathcal{A}_0 \text{ is singleton, } \mathcal{A}_0 \equiv \{M\}$$

and only in this case we discuss optimality and/or maximality of steady states and of the Faustmann Solution. The case of multivalued \mathcal{A}_0 remains unsolved, nonetheless multiplicity of maxima is a fragile phenomenon that vanishes under small perturbations of the productivity function.

Theorem 5.5 *Assume that (32) is satisfied, $\rho = 0$ and $u(r) = r$ for all $r \geq 0$. Consider an initial datum $x_0 \in \Pi$ with $\text{supp}(x_0) \subseteq [0, M]$. Then the Faustmann Solution (\hat{x}, \hat{c}) , given by (29)–(28) is maximal, although it is not optimal. Indeed no optimal control exists for the problem in this set of data.*

The fact applies to the particular case of the golden rule.

Corollary 5.6 *In the assumptions of Theorem 5.5, the golden rule (\bar{x}, \bar{c}) is a maximal, but not optimal, program at \bar{x} . Moreover no admissible control at \bar{x} may be optimal.*

As a direct proof of the assertion that \bar{c} is not optimal, nor an optimal control exists, one may build the following example (the proof of Theorem 5.5 is based on a similar construction) where the control \bar{c} is not catching up to c_1 defined by means of (33), admissible at \bar{x} . The control c_1 behaves *on average* like \bar{c} but delayed of some initial time interval: the difference in utilities yielded by \bar{c} and c_1 coincide repeatedly with their difference in the initial time interval, precisely because $\rho = 0$ and u is linear.

Example 5.7 Let N be a natural number greater than 1. Define $s_j := jM/N$, for $j = 1, \dots, N$ and consider a control c_1 and associated trajectory x_1 so defined: when $t \leq M/N$, c_1 cuts the quantity $x_1(t, s_j)$ of available trees of age s_j , subsequently when $t \geq M/N$, c_1 cuts the quantity $x_1(t, M)$ of trees reaching age M , that is

$$(33) \quad c_1(t) = \begin{cases} \sum_{j=1}^N x_1(t, s_j) \delta_{s_j}, & 0 \leq t < \frac{M}{N} \\ x_1(t, M) \delta_M, & t \geq \frac{M}{N} \end{cases}$$

It is easy to check that c_1 is admissible at any x_0 with $\text{supp}(x_0) \subset [0, M]$, in particular for $x_0 = \bar{x}$. In the latter case, the associated trajectory $x_1(t, s; c_1, \bar{x}) \equiv x_1(t, s)$ is (a.e.) given by the explicit formula

$$x_1(t, s) = \frac{N}{M} \chi_{[0, t]}(s) + \frac{1}{M} \sum_{j=1}^N \chi_{[s_{j-1} + t, s_j]}(s)$$

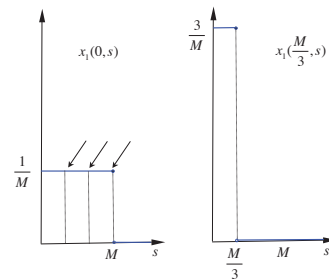


FIGURE 4.— The control c_1 with $N = 3$. From time 0 to time $M/3$ the trees of age $M/3$, $(2M)/3$, and M are cut, yielding a resulting distribution, at time $M/3$, $x_1(M/3, s) = \frac{3}{M} \chi_{[0, M/3]}$.

when $t \in [0, \frac{M}{N}]$, $s \geq 0$, while in the following interval of length M it is equal to

$$(34) \quad x_1(t, s) = \begin{cases} \frac{N}{M} \chi_{[t-\frac{M}{N}, t]}(s) & t \in [\frac{M}{N}, M] \\ \frac{N}{M} \left[\chi_{[0, t-M]}(s) + \chi_{[t-\frac{M}{N}, M]}(s) \right] & t \in [M, M + \frac{M}{N}]. \end{cases}$$

It is easily understood that, from $t = M/N$ on, the trajectory is periodic with period M , and attains the values described in (34) in all intervals of type $[T_i, T_{i+1}]$ where $T_i = M/N + iM$, with $i \in \mathbb{N}$.

Note that for during every time interval $[T_i, T_{i+1}]$, the control c_1 cuts an amount N/M for a time length M/N , while \bar{c} cuts the amount $1/M$ for a time length M . Then, except for the quantity obtained at the initial time interval, the utilities yielded by \bar{c} and c_1 on a period length interval are both equal to $f(M)$, as no discount is applied and u is linear. As a result, the difference between such utilities is periodically equal to the difference yielded on $[0, M/N]$, that is

$$U_{\frac{M}{N}}(c_1) - U_{\frac{M}{N}}(\bar{c}) = \frac{1}{N} \sum_{j=1}^{N-1} f(s_j)$$

and which may be assumed strictly positive, provided f is not null everywhere, as for a suitable choice of N one may infer that $f(s_j) > 0$ for at least one j . Such fact is equivalent by definition to stating that the control \bar{c} cannot be (definitively) catching up to c_1 , and by means of the same idea one is also able to contradict the existence of an optimal control. For details we refer the reader to the proof of the general case, Theorem 5.5 in the Appendix. \square

Remark 5.8 As it is shown later in Theorem 5.12 (ii), for a utility u which is concave but not necessarily strictly concave, one may prove existence of a maximal control when the admissible set is $\mathcal{U}_{x_0}^{\lambda, K}$ defined in (17). In particular, the result applies when u is linear (and $\rho = 0$). \square

5.3. Strictly concave utility, null discount

As in the previous, in this subsection we assume that (32) is satisfied. Moreover all statements are proved not for \mathcal{U}_{x_0} as admissible class of strategies, but on the subsets $\mathcal{U}_{x_0}^\lambda$ and $\mathcal{U}_{x_0}^{K, \lambda}$ defined in (16) (17).

Theorem 5.9 *Assume $\rho = 0$, and that (10)(11)(32) are satisfied. Assume moreover that u is strictly concave. Then, along the trajectory x starting from some $x_0 \in \Pi$ and driven by a good control $c \in \mathcal{U}_{x_0}^\lambda$ one has*

$$x(t) \xrightarrow[L^2(0, S)]{t \rightarrow \infty} \bar{x},$$

that is, the trajectory x converges to the golden rule \bar{x} in $L^2(0, S)$ norm.

A straightforward consequence of Theorem 5.9 and of its assertion holds also for optimal trajectories, *i.e.* trajectories driven by an optimal control, which is good by means of Proposition 4.14.

Theorem 5.10 *Assume $\rho = 0$, and that (10) (11) (32) hold. Assume moreover that u is strictly concave. Then the golden rule (\bar{x}, \bar{c}) is an optimal stationary couple.*

As a consequence of Theorem 5.9 the Faustmann solutions (except for the Golden Rule itself), which were maximal for linear u , are not maximal anymore for strictly concave u , as they are definitely caught up by the convergent solution. As a consequence, Theorem 5.9 and Proposition 4.14 imply the following Corollary.

Corollary 5.11 *In the assumptions of Theorem 5.9, the Faustmann solution can be neither an optimal nor a maximal program, except for the particular case of the Golden Rule.*

Theorem 5.12 *Let $x_0 \in \Pi$, $\rho = 0$, and assume (32) is satisfied. Let $\mathcal{U}_{x_0}^{K,\lambda}$ be the space of admissible control defined in (17). Then:*

- (i) *if u is strictly concave, then there exists an optimal control in $\mathcal{U}_{x_0}^{K,\lambda}$;*
- (ii) *if u is concave (but not necessarily strictly concave), then there exists a maximal control in $\mathcal{U}_{x_0}^{K,\lambda}$.*

Remark 5.13 The existence result yielded by Theorem 5.12 is not as general as one may hope. Indeed the proof works based on the fact that admissible controls are uniformly bounded by a common constant K , as that set enjoys the compactness properties needed to derive existence of a maximum. Nonetheless, for bounded initial data and for K chosen big enough, the Faustmann policy and the golden rule fall into the described set of controls. \square

5.4. Strictly concave utility, positive discount

As observed in Corollary 5.11 the Faustmann Policy is not optimal for the case of a strictly concave utility function and null discount. However, this result does not preclude the possibility that the Faustmann policy turns out optimal for the discounted model. Indeed, for the discrete time model with a strictly concave utility and discounted future utilities, Mitra and Wan (1985) provided a couple of examples in which the Faustmann Policy was in fact optimal, and Wan (1994); Salo and Tahvonen (2002, 2003) have taken the issue further (see also Mitra et al., 1991 for similar results in a different vintage capital model) by showing that optimal Faustmann cycles persist in a neighborhood of the (modified) golden rule even if the discount factor approaches unity. Proposition 1 in Salo and Tahvonen (2003), in particular, states that for any discount factor less than one the Faustmann Policy is optimal for all initial forests that are sufficiently close to the uniform steady state forest.

On the contrary, for the strictly concave continuous-time discounted model the issue of the optimality of cyclical Faustmann solutions is still open: neither a convergence result is available, nor a case in which the Faustmann Policy is optimal has been found. However, we can establish a partial result by proving that Proposition 1 in Salo and Tahvonen (2003) does not carry over to our continuous-time formulation and, hence, that the model behaves differently from the discrete-time model. To illustrate the point consider the following simple example in which a Most Rapid Approach Path to the steady state dominated the path generated by the Faustmann Policy for a set of initial distributions that contains elements arbitrary close to the (modified) golden rule. Assume that $M_\rho = 1$ is the unique Faustmann maturity age, and that $f(1) = 1$. Consider the following initial density of forest

$$x_a(s) = \begin{cases} 1 & s \in [0, 1 - 2a) \\ 1 + a & s \in [1 - 2a, 1 - a) \\ 1 - a & s \in [1 - a, 1] \\ 0 & s \in]1, +\infty), \end{cases}$$

where $0 \leq a \leq \frac{1}{2}$ (note that for $a = 0$ we obtain the golden rule forest). We intend to show that for any a in a right neighborhood of 0 the Faustmann Policy is not optimal starting at $x_a(s)$. As a consequence, the analogous of Proposition 1 in Salo and Tahvonen (2003) does not hold in continuous time.

We first compute the utility associated to the Faustmann Policy $\hat{c}_a(t) = \hat{x}_a(t, 1)\delta$. Note that each Faustmann cycle comprises three phases of constant timber consumption: an initial phase of length a during which $1 - a$ units of timber are consumed followed by a phase of length a during which timber consumption rises to $1 + a$ and a final phase of length $1 - 2a$ during which consumption is constant at the modified golden rule level. Then on the first cycle, that is for $t \in [0, 1]$, one has

$$\langle \hat{c}_a(t), f \rangle = f(1)x_a(t, 1) = x_a(t, 1) = (1 - a)\chi_{[0, a]}(t) + (1 + a)\chi_{[a, 2a]}(t) + \chi_{[2a, 1]}(t),$$

so that the utility at horizon $T = 1$ is given by

$$\begin{aligned} U_1(\hat{c}_a) &= \int_0^a u(1 - a)e^{-\rho s} ds + e^{-\rho a} \int_0^a u(1 + a)e^{-\rho s} ds + e^{-2\rho a} \int_0^{1-2a} u(1)e^{-\rho s} ds \\ &= \frac{u(1 - a) + u(1 + a)e^{-\rho a}}{\rho} (1 - e^{-\rho a}) + \frac{u(1)}{\rho} (e^{-2\rho a} - e^{-\rho}). \end{aligned}$$

Next we note that, starting from any forest $x_a(s)$, a feasible Most Rapid Approach Path to the steady state, namely

$$c_{mra}(t) = [(1 - a)\delta_1 + a\delta_{1-a}] \chi_{[0, a]}(t) + \delta_1 \chi_{[a, \infty)}(t),$$

reaches the golden rule after a units of time by continuously clearing the $1 - a$ units of land on which mature trees are planted and the a units of land in excess on which trees of age $1 - a$ are grown. The utility associated to this MRA policy at horizon $T = 1$ is then

$$\begin{aligned} U_1(c_{mra}) &= \int_0^a u((1 - a) + af(1 - a))e^{-\rho s} ds + e^{-\rho a} \int_0^a u(1)e^{-\rho s} ds + e^{-2\rho a} \int_0^{1-2a} u(1)e^{-\rho s} ds \\ &= u(1 - a + af(1 - a)) \frac{(1 - e^{-\rho a})}{\rho} + \frac{u(1)}{\rho} (e^{-\rho a} - e^{-\rho}). \end{aligned}$$

Instead, when $T = n \geq 2$, one has

$$U_n(\hat{c}_a) = \sum_{i=0}^{n-1} e^{-i\rho} U_1(\hat{c}_a) = U_1(\hat{c}_a) + U_1(\hat{c}_a) \frac{e^{-\rho} - e^{-n\rho}}{1 - e^{-\rho}}$$

$$U_n(c_{mra}) = U_1(c_{mra}) + \sum_{i=1}^n e^{-i\rho} \frac{u(1)}{\rho} (1 - e^{-\rho}) = U_1(c_{mra}) + \frac{u(1)}{\rho} (e^{-\rho} - e^{-n\rho}).$$

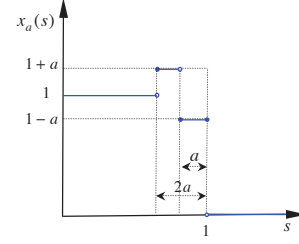


FIGURE 5.— The initial state $x_a(s)$.

As a consequence one has

$$\begin{aligned} & \lim_{n \rightarrow \infty} (U_n(\hat{c}_a) - U_n(c_{mra})) = \\ & = \frac{1 - e^{-\rho a}}{\rho} \left[\frac{u(1-a) - u(1)}{1 - e^{-\rho}} + \frac{u(1+a) - u(1)}{1 - e^{-\rho}} e^{-\rho a} - u(1-a + af(1-a)) + u(1) \right]. \end{aligned}$$

The sign of the above expression is the sign of the sum in the square brackets, which is null at $a = 0$, hence its sign for $a > 0$ in a neighborhood of 0 is given by the sign of its lowest order non-zero derivative evaluated at the steady state. Simple calculations show that the first derivative is null, while the second one is given by $2(u''(1) - \rho u'(1)) + 2f'(1)u'(1)(1 - e^{-\rho}) = 2u''(1)$, which is strictly negative, as $f'(1)(1 - e^{-\rho}) = \rho$ because $M_\rho = 1$ is the solution of the Faustmann problem. One can therefore conclude that the Faustmann Policy is not optimal starting at $x_a(s)$ in a neighborhood of the steady state.

6. CONCLUSIONS

In this paper we developed and analyzed a continuous time version of the Mitra and Wan (1985) model of optimal forest management. Following the methodological precept that: "No substantive prediction or explanation in a well-defined macroeconomic period model should depend on the real time length of the period" (Foley, 1975, p. 301), our main purpose was to isolate the set of phenomena that in the optimal management of a forest are independent of the way time is modeled. Table 1 gives an overview of the results we have established for the continuous-time model and that can be compared with the results that have been obtained in the discrete time model.

	$\rho = 0$	$\rho > 0$
u linear	<ul style="list-style-type: none"> • If \mathcal{A}_0 is singleton the GR is maximal, but not optimal • If \mathcal{A}_0 is singleton the FS is maximal at any admissible x_0 satisfying $\text{supp}(x_0) \subset [0, M_\rho]$ • There do not exist optimal controls 	<ul style="list-style-type: none"> • Any MGR is optimal • FS is optimal at any admissible x_0 satisfying $\text{supp}(x_0) \subset [0, M_\rho]$
u strictly concave	Assume \mathcal{A}_0 is singleton: <ul style="list-style-type: none"> • GR is the unique optimal stationary couple • There exists an optimal control • Any optimal trajectory converges (in L^2 norm) to the GR 	<ul style="list-style-type: none"> • Any MGR is optimal • It is not true (as in discrete time) that FS is optimal for all initial forests close to the MGR. There is a counterexample
u concave _{u}	Assume \mathcal{A}_0 is singleton: <ul style="list-style-type: none"> • GR is the unique maximal stationary couple • There exists a maximal (admissible) control 	<ul style="list-style-type: none"> • Any MGR is optimal

TABLE I

RESULTS AT ONE GLANCE: FS STANDS FOR FAUSTMANN SOLUTION, GR FOR GOLDEN RULE, MGR FOR MODIFIED GOLDEN RULE.

It turned out that many of the discrete time results carry over to the continuous time version of the model with an important exception: the cyclical optimal solutions that are characteristics of the discounted strictly-concave discrete model disappear in continuous time. We have also established for the continuous time model a set of results that the literature in discrete time has not yet proved: that the Faustmann solution is maximal but not optimal for the undiscounted model with a linear utility function and that in steady states both the Faustmann age and timber production decrease monotonically with the increase of the discount rate.

Unlike in discrete time, modeling timber production in continuous time required a quantum leap from the received vintage capital theory. Indeed, in the typical vintage capital model in continuous time, only irreversible investment in new machines is possible, so that distributed controls can be avoided altogether. Moreover, in the few instances in which investment in older machines is considered (*e.g.* Feichtinger et al., 2006) it turned out that optimal investment is spread over a continuum of ages, so that the controls can be functions. In continuous time, however, timber production cannot be modeled this way, because the Faustmann condition implies that generically it is optimal to fell down only trees of a single age. Therefore, to handle the case of forest management in continuous time we had to develop an entirely new class of vintage models in which measure-valued controls are allowed. Since this is the first attempt to formulate the Mitra-Wan model in continuous time, we have been concentrating on the basic features of the model, without attempting to use minimal assumptions and without taking into account recent refinements of the theory (Khan and Piazza, 2012). Discussion of these issues is left for future work. One may also work on generalizations of the model in several directions, for example considering positive cutting costs, adding environmental well-being variables into the objective, and allowing for alternative use of the forest land.

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APPENDIX A: PROOFS

In this appendix we complete our exposition with the proofs of the aforementioned results.

A.1. Proofs for Section 3

The subsection contains the proofs of the results which were stated in Section 3 and of some other useful ones. Those results are mostly well known in semigroup theory applied to control in infinite dimensions. In this respect, our main reference is Bensoussan et al. (2007).

Proposition A.1 *Given $T > 0$, the operator $\mathcal{C}: L^2(0, T; D') \rightarrow C([0, T]; D')$ given by $\mathcal{C}(c)(t) := \int_0^t e^{(t-s)A} Bc(s) ds$ is continuous. As a consequence $\mathcal{S}: D' \times L^2(0, T; D') \rightarrow C([0, T]; D')$ defined by $\mathcal{S}(x_0, c)(t) := T(t)x_0 + \mathcal{C}(c)(t)$ is continuous. In particular, for any $x_0 \in D'$, and for any $c \in \mathcal{U}_{x_0}$, the function $[0, T] \rightarrow D'$, $t \mapsto T(t)x_0 + \mathcal{C}(c)(t)$ is also continuous.*

PROOF: See *e.g.* Bensoussan et al. (2007) Section II.1.3.

Q.E.D.

Proof of Proposition 3.3. The proof of the first fact is a consequence of Proposition A.1. Next we need to show that, for any $t \in [0, +\infty)$, $x(t)$ lies in $L^2(0, S)$. Expanding in (9) the definition of B we have

$$(35) \quad x(t) = T(t)x_0 + \int_0^t \langle c(\tau), \psi \rangle T(t - \tau) \delta_0 \, d\tau - \int_0^t T(t - \tau) c(\tau) \, d\tau.$$

- (i) Since $x_0 \in L^2(0, S)$ then $T(t)x_0 \in L^2(0, S)$, as $T(t)$ coincides on $L^2(0, S)$ with the translation semi-group. By definition, $T(\tau)x_0$ is then a positive function;
- (ii) The term $\int_0^t \langle c(\tau), \psi \rangle T(t - \tau) \delta_0 \, d\tau$ is a positive function, as all integrands are positive. Moreover it belongs to $L^2(0, S)$ as a consequence of Proposition 3.1 in Bensoussan et al. (2007), page 212, once we have proven that Hypothesis 3.1 page 212 in Bensoussan et al. (2007) is satisfied. To this extent, we denote with B_1 the operator $B_1 : D' \rightarrow D'$ given by $B_1 \phi := \langle \phi, \psi \rangle \delta_0$ so that $B_1^* : D \rightarrow D$ is given by $B_1^* h := \langle \delta_0, h \rangle \psi$. As a consequence

$$|B_1^* T^*(\tau) h|_D \leq |\langle \delta_0, T^*(\tau) h \rangle|_{\mathbb{R}} |\psi|_D \leq |h(\tau)|_{\mathbb{R}} |\psi|_D$$

so that

$$\int_0^T |B_1^* T^*(\tau) h|_D^2 \, d\tau \leq |\psi|_D^2 \int_0^S |h(\tau)|_{\mathbb{R}}^2 \, d\tau = |\psi|_D^2 |h|_{L^2(0, S)}^2,$$

and the statement is proven.

- (iii) Last we observe that $-\int_0^t T(t - \tau) c(s) \, d\tau$ is a negative distribution in D' and by Proposition 2.3 page 270 of Hirsch and Lacombe (1999) is a negative measure on $[0, S]$. Then, by means of Lebesgue decomposition theorem (see *e.g.* Rudin, 1987, Theorem 6.10 page 121), it may be decomposed into a (negative) part, absolutely continuous w.r.t. the Lebesgue measure on $[0, S]$, and a (negative) singular part.

As a result of the previous analysis, the measure defined by the right side of (35) has singular part coinciding with that described in (iii). At the same time, the singular part of the measure defined by the left side of (35) need be positive, then the singular part on both sides is null. Then for any $t \geq 0$, the term $-\int_0^t T(\tau) c(\tau) \, d\tau$ is a function in $L^2(0, S)$, as it is a measure, absolutely continuous w.r.t. the Lebesgue measure on $[0, S]$, and whose density (by positivity of $x(t)$ and of the first two terms in the right hand side of (35)), is dominated (in absolute value) by a function in L^2 (the sum of such positive terms). The explicit formula solution when $c(\cdot) \in L_{loc}^2(0, +\infty; L^2(0, S))$ can be computed by standard calculations. *Q.E.D.*

Proof of Proposition 3.4. Consider first c in $L_{loc}^2(0, +\infty; L^2(0, S))$. By means of (4) one has

$$\langle x(t), \psi \rangle = \langle x_0, \psi \rangle - \int_0^{\bar{s}} \int_0^{t \wedge s} c(t - \tau, s - \tau) \, d\tau \, ds + \int_0^{t \wedge s} \int_0^{\bar{s}} c(t - s, \tau) \, d\tau \, ds.$$

It then suffices a change of variables in the first integral above to derive

$$\int_0^{\bar{s}} \int_0^{t \wedge s} c(t - \tau, s - \tau) \, d\tau \, ds = \int_0^t \int_0^{\bar{s}} c(s, \tau) \, d\tau \, ds = \int_0^t \int_0^{\bar{s}} c(t - s, \tau) \, d\tau \, ds$$

and hence $\langle x(t), \psi \rangle = \langle x_0, \psi \rangle$. The claim for a general c in \mathcal{U}_{x_0} follows by density and by continuity of the operator \mathcal{S} defined in Proposition A.1. *Q.E.D.*

It is a well known fact (see Bensoussan et al., 2007 Section II.3.1, pages 201-204)⁸ that if $x(t)$ is a solution of (9) then it is also a *weak solution* of the same equation, where by weak we mean that the the left and right hand sides of (8) are equal when evaluated at any test function $p \in D^2$, with D^2 defined in (7). Hence the following proposition holds true.

⁸More precisely, as indicated at page 204, one has to repeat the construction of the weak solution at page 203, with $k \in D^2$.

Proposition A.2 *Let x be the solution to (9) when $x_0 \in \Pi$, $c \in \mathcal{U}_{x_0}$. Let also T be any finite horizon and p be any function in D^2 . Then*

$$\begin{cases} \frac{d}{dt} \langle x(t), p \rangle = \langle x(t), A^* p \rangle - \langle c(t), p \rangle + \langle \delta_0, p \rangle \langle \psi, c(t) \rangle, & \forall t \in (0, T] \\ \langle x(0), p \rangle = \langle x_0, p \rangle = 1. \end{cases}$$

Proof of Lemma 3.7 Let $\phi \in C^1(0, S; \mathbb{R})$ be compactly supported in $(0, S)$ (so that $\phi(S) = 0$ and $\phi \in D$). By definition, a stationary couple (\tilde{x}, \tilde{c}) satisfies

$$\langle \tilde{x}, \phi \rangle_{L^2} = \left\langle T(t)\tilde{x} + \int_0^t T(t-s)B\tilde{c}ds, \phi \right\rangle_{L^2}.$$

By deriving with respect to t both sides and evaluating at $t = 0$ one has

$$0 = \langle \tilde{x}, A^* \phi \rangle_{L^2} + \langle B\tilde{c}, \phi \rangle = \langle \tilde{x}, A^* \phi \rangle - \langle \tilde{c}, \phi \rangle = \langle \tilde{x}, \phi' \rangle - \langle \tilde{c}, \phi \rangle$$

as $\phi(0) = 0$ and $A^* \phi = \phi'$, which implies that \tilde{c} is the distributional derivative of $-\tilde{x}$. Since in addition \tilde{c} is a positive measure, $-\tilde{x}$ is an almost everywhere increasing function and \tilde{c} is its Stieltjes derivative (see Theorem 346.1 and the related proof in Ziemer, 2004).

On the other hand, if $x \in \Pi$, with $-\tilde{x}$ increasing and with Stieltjes derivative \tilde{c} , it suffices to show that, for any ϕ in D and any $t \geq 0$,

$$(36) \quad \frac{d}{dt} \left\langle T(t)\tilde{x} + \int_0^t T(t-s)B\tilde{c}ds, \phi \right\rangle_{L^2} = 0.$$

We denote by $\partial\tilde{x}$ the Stieltjes derivative of \tilde{x} . Since $s \mapsto [T^*(t)\phi](s)$ is differentiable, the derivative in (36) equals

$$\langle \tilde{x}, A^* T^*(t)\phi \rangle_{L^2} + \langle B\tilde{c}, T^*(t)\phi \rangle = \int_0^S [T^*(t)\phi]'(r)\tilde{x}(r)dr - [T^*(t)\phi](0) \int_0^S \partial\tilde{x}(r) + \int_0^S [T^*(t)\phi](r)\partial\tilde{x}(r).$$

Using the integration by part formula for Stieltjes integrals (see *e.g.* Hewitt, 1960), and denoting by $\tilde{x}(0) = \lim_{s \rightarrow 0^+} \tilde{x}(s)$ and $\tilde{x}(S) = \lim_{s \rightarrow S^-} \tilde{x}(s)$ (we recall that \tilde{x} is monotone), the previous expression is equal to

$$\begin{aligned} & \int_0^S [T^*(t)\phi]'(r)\tilde{x}(r)dr - [T^*(t)\phi](0)(\tilde{x}(S) - \tilde{x}(0)) \\ & \quad + [T^*(t)\phi](S)\tilde{x}(S) - [T^*(t)\phi](0)\tilde{x}(0) - \int_0^S [T^*(t)\phi]'(r)\tilde{x}(r)dr = 0 \end{aligned}$$

since $\tilde{x}(S) = 0$ and $[T^*(t)\phi](S) = 0$ by definition of $T^*(t)$ in Section 3.1.

Q.E.D.

A.2. Proofs for Section 4

Proof of Proposition 4.2. Note that $\mathcal{A}_\rho \neq \emptyset$ is a consequence of the fact that $g_\rho(\cdot)$ is continuous and possibly non-zero only on the compact set $[\lambda, \bar{s}]$, while (10) implies $0 \notin \mathcal{A}_\rho$ for any $\rho \geq 0$. Now let $M_{\rho_A} \in \mathcal{A}_{\rho_A}$ and $M_{\rho_B} \in \mathcal{A}_{\rho_B}$ be arbitrarily chosen.

Now, note that the support function h_ρ defined in (20) is an exponential function, increasing and convex, so that $\rho_B < \rho_A$ implies h_{ρ_A} is definitively greater than h_{ρ_B} for increasing values of ρ . Hence, $h_{\rho_A}(0) = h_{\rho_B}(0) = 0$ and: either 1) $h_{\rho_A}(s) > h_{\rho_B}(s)$ for all $s \in (0, S]$, or 2) there exists $\tilde{s} \in (0, +\infty)$ such that $h_{\rho_A}(\tilde{s}) = h_{\rho_B}(\tilde{s})$, $h_{\rho_A}(s) < h_{\rho_B}(s)$ for all $s \in (0, \tilde{s})$, and $h_{\rho_A}(s) > h_{\rho_B}(s)$ for all $s \in (\tilde{s}, +\infty)$. Nonetheless, the former never takes place, as g_{ρ_B} is maximal at M_{ρ_B} implies

$$h_{\rho_A}(M_{\rho_A}) - h_{\rho_B}(M_{\rho_A}) = (e^{\rho_B M_{\rho_B}} - 1)(g_{\rho_B}(M_{\rho_A}) - g_{\rho_B}(M_{\rho_B})) \leq 0$$

so that

$$h_{\rho_A}(M_{\rho_A}) \leq h_{\rho_B}(M_{\rho_A}), \text{ with } M_{\rho_A} > 0.$$

As a consequence, $M_{\rho_A} \in (0, \tilde{s}]$.

Similarly, from the maximality of g_{ρ_A} at M_{ρ_A} , one derives $h_{\rho_A}(M_{\rho_B}) \geq h_{\rho_B}(M_{\rho_B})$, which implies $\tilde{s} \leq S$ and $M_{\rho_B} \in [\tilde{s}, S]$. Since the selections M_{ρ_A} and M_{ρ_B} were arbitrarily chosen in \mathcal{A}_{ρ_A} and \mathcal{A}_{ρ_B} respectively, the first assertion in (i) is proven. If in addition f is differentiable at \tilde{s} , assume by contradiction that $\mathcal{A}_{\rho_A} \cap \mathcal{A}_{\rho_B} = \{\tilde{s}\}$. Then by Remark 4.3 $f(\tilde{s}) = h_{\rho_A}(\tilde{s}) = h_{\rho_B}(\tilde{s})$, and $h'_{\rho_A}(\tilde{s}) > h'_{\rho_B}(\tilde{s})$ which contradicts the fact that the graph of f lies underneath the graph of both support functions.

Next we prove (ii). The fact that the selection $\rho \mapsto M_\rho$ is nonincreasing is a direct consequence of (i). Now note that from $M_{\rho_A} \leq M_{\rho_B}$ and the convexity of $h_{\rho_B}(s)$ (see also Figure 1) follows

$$(37) \quad \frac{h_{\rho_B}(M_{\rho_B})}{M_{\rho_B}} \geq \frac{h_{\rho_B}(M_{\rho_A})}{M_{\rho_A}}$$

while Remark 4.1 implies

$$(38) \quad f(M_{\rho_B}) = h_{\rho_B}(M_{\rho_B}) \text{ and } h_{\rho_B}(M_{\rho_A}) \geq f(M_{\rho_A})$$

so that (37) and (38) give

$$\frac{f(M_{\rho_B})}{M_{\rho_B}} \geq \frac{f(M_{\rho_A})}{M_{\rho_A}}.$$

The last claim in (ii) is a consequence of (i) and of countability of the discontinuities of a decreasing function (see *e.g.* Chung, 2001 page 4). The limit m_0 in (iii) exists and is contained in $[0, S]$, as any selection M_ρ in \mathcal{A}_ρ is nonincreasing and there contained. Note that by continuity of f any \mathcal{A}_ρ necessarily has a positive minimum, and that $\mathcal{A}_\rho \cap \mathcal{A}_0$ contains at most one element, which implies $m_0 \leq \min \mathcal{A}_0$. Suppose by contradiction that $m_0 < \min \mathcal{A}_0$. Then $h_0(s) - f(s)$ is always strictly positive on $[\lambda, m_0]$. Moreover, if we define

$$(39) \quad k_\rho(s) = \frac{f(\min \mathcal{A}_0)}{e^{\rho \min \mathcal{A}_0} - 1} (e^{\rho s} - 1).$$

we may observe that: (i) $h_\rho(s) \geq k_\rho(s)$, and (ii) $k_\rho(s) - h_0(s)$ converges uniformly to 0 on $[\lambda, m_0]$, when $\rho \rightarrow 0$. Hence there exists $\hat{\rho}$ small enough such that,

$$(40) \quad h_{\hat{\rho}}(s) \geq k_{\hat{\rho}}(s) > f(s)$$

for any $s \in [\lambda, m_0]$. This implies $\mathcal{A}_{\hat{\rho}} \subset (m_0, \min \mathcal{A}_0]$, a contradiction. Q.E.D.

Prior to demonstrate some of the theorems in Section 4 we need a series of preliminary results, which we state and prove hereby.

Lemma A.3 *Given $a, b \in [0, S]$, $a \leq b$, one has:*

$$(41) \quad T(t)\chi_{[a,b]} = \chi_{[a+t, (b+t) \wedge S]}, \quad \text{and} \quad \int_0^t T(\tau)\delta_b \, d\tau = \chi_{[b, (b+t) \wedge S]}$$

PROOF: The first assertion follows from

$$T(t)\chi_{[a,b]}(s) = \chi_{[a,b]}(s-t)\chi_{[t,S]}(s) = \chi_{[a+t, b+t]}(s)\chi_{[t,S]}(s) = \chi_{[a+t, (b+t) \wedge S]}(s).$$

For the second, note that, if ϕ is any test function in D , one has $\langle \delta_b, T^*(\tau)\phi \rangle = \phi(b+\tau)$ if $b+\tau \leq S$, and 0 otherwise, so that, by changing the variable in the integral with $\sigma = \tau + b$, one obtains

$$\left\langle \int_0^t T(\tau)\delta_b \, d\tau, \phi \right\rangle = \int_0^t \langle \delta_b, T^*(\tau)\phi \rangle \, d\tau = \int_b^{(b+t) \wedge S} \phi(\sigma) \, d\sigma = \langle \chi_{[b, (b+t) \wedge S]}, \phi \rangle$$

which implies the claim. Q.E.D.

Lemma A.4 *Given $g \in L^2_{loc}(0, +\infty; \mathbb{R})$, $a \in [0, S]$, one has, for any $t \in [0, S-a]$, $s \in [0, S]$:*

$$(42) \quad \left(\int_0^t g(\tau)T(t-\tau)\delta_a \, d\tau \right) (s) = \chi_{[a, t+a]}(s)g(t+a-s).$$

PROOF: The proof is a consequence of the previous Lemma and arguing by density after applying (42) to approximating step-functions. Q.E.D.

Lemma A.5 *Assume $\rho \geq 0$, and x_ρ , c_ρ , p_ρ are defined by means of (22) (23) (26) respectively. Assume (10) (11). Consider any trajectory x of system (8) starting at $x_0 \in \Pi$, and driven by a control $c \in \mathcal{U}_{x_0}$. Then, for all $t \geq 0$*

$$(43) \quad \langle c(t) - c_\rho, f - p \rangle \leq \rho \langle x(t) - x_\rho, p_\rho \rangle - \langle x(t) - x_\rho, A^* p_\rho \rangle.$$

PROOF: From Proposition 4.2 we derive that there exists M_ρ in the nonempty set \mathcal{A}_ρ . Moreover, by definition of p_ρ one has that $f(s) \leq p_\rho(s)$ for all $s \in [0, \bar{s}]$, and $f(M_\rho) = p_\rho(M_\rho)$, so so that for all positive measure $c \in D'$, one derives $\langle f, c \rangle \leq \langle p_\rho, c \rangle$ with the equality holding at $c = \gamma \delta_{M_\rho}$, with γ any nonnegative constant. In particular, for $\gamma = 1$, one has

$$(44) \quad \langle c(t), f - p_\rho \rangle \leq 0 = \langle c_\rho, f - p_\rho \rangle, \quad \forall t \geq 0$$

Now we recall that by means of Proposition 3.3, $x(t)$ lies in $L^2(0, S)$, so that the duality pairing with $x(t)$ coincides with scalar product with $x(t)$ in $L^2(0, S)$, moreover $A^* p_\rho = p_\rho'$ so that

$$(45) \quad -\rho \langle x(t), p_\rho \rangle + \langle x(t), A^* p_\rho \rangle = -\rho \int_0^{\bar{s}} p_\rho(s) x(t, s) ds + \int_0^{\bar{s}} p_\rho'(s) x(t, s) ds =: \Delta(\rho).$$

When $\rho > 0$, we have, $\text{supp } x(t) \subset [0, \bar{s}]$, $p_\rho'(s) = \rho \eta_\rho e^{\rho s}$ on $[0, \bar{s}]$, and Proposition 3.4 holds, so that

$$(46) \quad \Delta(\rho) = \rho \eta_\rho \int_0^{\bar{s}} x(t, s) ds = \rho \eta_\rho \int_0^{\bar{s}} x_0(s) ds = \rho \eta_\rho$$

that is, the quantity on the left of (45) hand side is constant for all trajectories (*i.e.*, for all admissible controls c , and all initial data x_0 covering a unitary area). In particular the property holds true when x is set equal to x_ρ . Then the claim follows by means of (44), (45) and (46). We proceed similarly for the case $\rho = 0$, as this time $p_0'(s) = \beta_0$ on $[0, \bar{s}]$, so that

$$\Delta(0) = \int_0^{\bar{s}} p_0'(s) x(t, s) ds = \beta_0 \int_0^{\bar{s}} x(t, s) ds = \beta_0,$$

which leads to the same conclusion. Q.E.D.

Corollary A.6 *In the assumption of Lemma A.5, and for u satisfying (11), set $\beta_\rho := f(M_\rho)/M_\rho$, and $\alpha_\rho := u'(\beta_\rho)$. Then for all $t \geq 0$, the value-loss function defined in (27) satisfies*

$$\theta_\rho(c(t), x(t)) \equiv u(\beta_\rho) - u(\langle c(t), f \rangle) + \alpha_\rho [\rho \langle x(t) - x_\rho, p_\rho \rangle - \langle x(t), A^* p_\rho \rangle + \langle c(t), p_\rho \rangle] \geq 0$$

PROOF: For all c in D' , define

$$h: D' \rightarrow \mathbb{R}, \quad h(c) := u(\langle c, f \rangle)$$

so that $h(c_\rho) = u(\langle c_\rho, f \rangle) = u(\beta_\rho)$, moreover h is differentiable with $h'(c) = u'(\langle c, f \rangle) f \in D$, $h'(c_\rho) = u'(\beta_\rho) f = \alpha_\rho f$. Since h is concave, we have $u(\langle c, f \rangle) - u(\beta_\rho) \leq \alpha_\rho \langle c - c_\rho, f \rangle$, for all $c \in D'$. Then Lemma A.5 implies

$$u(\langle c(t), f \rangle) \leq u(\beta_\rho) + \alpha_\rho [\langle c(t) - c_\rho, p_\rho \rangle + \rho \langle x(t) - x_\rho, p_\rho \rangle - \langle x(t), A^* p_\rho \rangle],$$

and, to complete the proof, we need to show that $-\langle c_\rho, p_\rho \rangle + \langle x_\rho, A^* p_\rho \rangle = 0$. This holds true when $\rho > 0$ as

$$-\langle c_\rho, p_\rho \rangle + \langle x_\rho, A^* p_\rho \rangle = -\frac{\eta_\rho}{M_\rho} (e^{\rho M_\rho} - 1) + \frac{\rho \eta_\rho}{M_\rho} \int_0^{M_\rho} e^{\rho s} ds = 0.$$

When instead $\rho = 0$, the claim follows from

$$(47) \quad \langle x_\rho, A^* p_0 \rangle = \int_0^{\bar{s}} x(t, s) \beta_0 ds = \beta_0 = \langle c_\rho, p_0 \rangle.$$

Q.E.D.

Remark A.7 Note that the proofs of the previous results remain true when c_ρ is replaced by a positive multiple $\gamma\delta_{M_\rho}$ of the Dirac's delta at M_ρ . If moreover $\rho = 0$, (43) holds true for a general initial datum x_0 in place of x_ρ . Indeed $\langle x, A^*p_0 \rangle = \beta_0$ for all $x \in L^2(0, S)$ having support in $[0, \bar{s}]$ and unitary extension $\int_0^{\bar{s}} x(s)ds = 1$. We summarize these facts in the following generalized version of Lemma A.5 for the case $\rho = 0$. \square

Corollary A.8 *Let be $\rho = 0$. Consider any trajectory $x(t)$ starting at $x_0 \in \Pi$ and driven by a control $c \in \mathcal{U}_{x_0}$. Then, for all $t \geq 0$, $\gamma \geq 0$,*

$$\langle c(t) - \gamma\delta_{M_0}, f - p_0 \rangle \leq -\langle x(t) - x_0, A^*p_0 \rangle.$$

Corollary A.9 *In the assumption of Corollary A.6*

$$(48) \quad U_T(c_\rho) - U_T(c) \geq \alpha_\rho \left(\langle x_\rho - x_0, p_\rho \rangle - e^{-\rho T} \langle x_\rho - x(T), p_\rho \rangle \right),$$

PROOF: From Corollary A.6 we derive

$$u(\beta_\rho) - u(\langle c(t), f \rangle) \geq \alpha_\rho [\rho \langle x_\rho - x(t), p_\rho \rangle - \langle c(t), p_\rho \rangle + \langle x(t), A^*p_\rho \rangle] = e^{\rho t} \frac{d}{dt} \langle x(t) - x_\rho, e^{-\rho t} p_\rho \rangle.$$

which promptly implies the thesis. Q.E.D.

Lemma A.10 *Assume x_ρ, p_ρ are defined by means of (22)(26) respectively, and let $\rho > 0$. Consider any trajectory x of system (8) starting at x_ρ , and driven by a control $c \in \mathcal{U}_{x_\rho}$. Then*

$$\lim_{T \rightarrow +\infty} \int_0^T \frac{d}{dt} [\langle x_\rho - x(t), e^{-\rho t} p \rangle] dt = 0.$$

PROOF: Note that

$$\int_0^T \frac{d}{dt} [\langle x_\rho - x(t), e^{-\rho t} p \rangle] dt = e^{-\rho T} \langle x_\rho - x(T), p_\rho \rangle$$

so that recalling that $x(T)$ and x_ρ are supported in $[0, \bar{s}]$ and that (15) holds, one gets

$$\left| e^{-\rho T} \langle x_\rho - x(T), p_\rho \rangle \right| \leq e^{-\rho T} \int_0^S |x_\rho(s) - x(T, s)| |p_\rho(s)| ds \leq 2e^{-\rho T} \eta_\rho (e^{\rho S} - 1) \xrightarrow{T \rightarrow +\infty} 0$$

which implies the claim. Q.E.D.

Proof of Proposition 4.4. We need to show that (c_ρ, x_ρ) satisfies Definition 3.6. Note that by means of (6) we have

$$Bc_\rho = -\frac{1}{M_\rho} \delta_{M_\rho} + \left\langle \frac{1}{M_\rho} \delta_{M_\rho}, \psi \right\rangle \delta_0 = \frac{1}{M_\rho} (\delta_0 - \delta_{M_\rho})$$

so that, by making use of (41) one obtains

$$\begin{aligned} T(t)x_\rho + \int_0^t T(t-\tau)Bc_\rho d\tau &= \frac{1}{M_\rho} \chi_{[t, (t+M_\rho) \wedge S]} + \frac{1}{M_\rho} \int_0^t T(\sigma)(\delta_0 - \delta_{M_\rho}) d\sigma \\ &= \frac{1}{M_\rho} \chi_{[t, (t+M_\rho) \wedge S]} + \frac{1}{M_\rho} (\chi_{[0, t \wedge S]} - \chi_{[M_\rho, (M_\rho+t) \wedge S]}) = \frac{1}{M_\rho} \chi_{[0, M_\rho]} = x_\rho \end{aligned}$$

which implies the thesis. Q.E.D.

Proof of Theorem 4.5. We make use of Definition 3.5 to show that c_ρ is optimal. Let U_T be defined by means of (12) and let c be any control admissible at the initial datum x_ρ , and let $x(t) = x_{x_\rho, c}(t)$ be the associated trajectory. Define

$$\begin{aligned} (49) \quad \Gamma(T) &:= U_T(c_\rho) - U_T(c) + \alpha_\rho \int_0^T \frac{d}{dt} [\langle x_\rho - x(t), e^{-\rho t} p_\rho \rangle] dt = \\ &= \int_0^T e^{-\rho t} [u(\beta_\rho) - u(\langle c(t), f \rangle) + \alpha_\rho (\rho \langle x_\rho - x(t), p_\rho \rangle - \langle x(t), A^*p_\rho \rangle + \langle c(t), p_\rho \rangle)] dt \\ &\quad - \int_0^T e^{-\rho t} \langle c(t), \psi \rangle \langle \delta_0, p_\rho \rangle dt. \end{aligned}$$

where the last equality is obtained by applying Proposition A.2. By means of Corollary A.6 the first addendum in the right hand side is positive, while the second is null, due to the fact that $\langle p, \delta_0 \rangle = 0$. Hence $\Gamma(T) \geq 0$, which by means of Lemma A.10 implies

$$\liminf_{T \rightarrow +\infty} (U_T(c_\rho) - U_T(c)) \geq 0,$$

as required.

Now we assume that \mathcal{A}_ρ is singleton and that (\tilde{x}, \tilde{c}) is an optimal stationary couple, and we show that (\tilde{x}, \tilde{c}) necessarily coincides with the golden rule. We start by showing that $\text{supp}(\tilde{c}) = \{M_\rho\}$. To do so, we prove that $\text{supp}(\tilde{c}) \cap [0, M_\rho) = \emptyset$ and $\text{supp}(\tilde{c}) \cap (M_\rho, \bar{s}] = \emptyset$. Assume by contradiction that $\text{supp}(\tilde{c}) \cap [0, M_\rho) \neq \emptyset$ and define, for $\epsilon > 0$, the control

$$(50) \quad c_\epsilon(t)(s) := \begin{cases} (1 - \epsilon)\chi_{[0, M_\rho)}(s)\tilde{c}(s) + \chi_{(M_\rho, \bar{s}]}(s)\tilde{c}(s) + \epsilon\delta_{M_\rho} \int_0^t \tilde{c}(M_\rho - s) ds & t \in [0, M_\rho) \\ (1 - \epsilon)\chi_{[0, M_\rho)}(s)\tilde{c}(s) + \chi_{(M_\rho, \bar{s}]}(s)\tilde{c}(s) + \epsilon\delta_{M_\rho} \int_0^{M_\rho} \tilde{c}(s) ds & t \in [M_\rho, +\infty) \end{cases}$$

coinciding with \tilde{c} when $\epsilon = 0$.⁹ One can easily see that c_ϵ is admissible at \tilde{x} . If we show that $(d/d\epsilon)U(c_\epsilon)$ is strictly positive at $\epsilon = 0$ then, for ϵ small enough, $U(c_\epsilon) > U(\tilde{c})$ and \tilde{c} is not optimal:

$$\begin{aligned} \left. \frac{d}{d\epsilon} U(c_\epsilon) \right|_{\epsilon=0} &= \left. \frac{d}{d\epsilon} \left(\int_0^{+\infty} e^{-\rho t} u(\langle c_\epsilon(t), f \rangle) dt \right) \right|_{\epsilon=0} \\ &= u'(\langle \tilde{c}, f \rangle) \left(\int_0^{+\infty} e^{-\rho t} \left\langle -\tilde{c}\chi_{[0, M_\rho]} + \delta_{M_\rho} \left[\chi_{[0, M_\rho)}(t) \int_0^t \tilde{c}(M_\rho - s) ds \right. \right. \right. \\ &\quad \left. \left. \left. + \chi_{[M_\rho, +\infty)}(t) \int_0^{M_\rho} \tilde{c}(s) ds \right], f \right\rangle dt \right) = u'(\langle \tilde{c}, f \rangle) \left[\int_0^{+\infty} e^{-\rho t} \langle -\tilde{c}\chi_{[0, M_\rho]}, f \rangle dt + \right. \\ &\quad \left. + \int_0^{M_\rho} e^{-\rho t} f(M_\rho) \left(\int_0^t \tilde{c}(M_\rho - s) ds \right) dt + \left(\int_{M_\rho}^{+\infty} e^{-\rho t} f(M_\rho) dt \right) \left(\int_0^{M_\rho} \tilde{c}(s) ds \right) \right] \end{aligned}$$

Integrating with respect to t the first and the third addenda and integrating by parts (see *e.g.* Hewitt, 1960) the second, we obtain

$$(51) \quad \begin{aligned} &= u'(\langle \tilde{c}, f \rangle) \left[-\frac{1}{\rho} \langle f, \tilde{c}\chi_{[0, M_\rho]} \rangle + \left(-\frac{f(M_\rho)}{\rho} e^{-\rho M_\rho} \int_0^{M_\rho} \tilde{c}(s) ds \right. \right. \\ &\quad \left. \left. + \frac{f(M_\rho)}{\rho} \int_0^{M_\rho} e^{-\rho t} \tilde{c}(M_\rho - t) dt \right) + \frac{f(M_\rho)}{\rho} e^{-\rho M_\rho} \int_0^{M_\rho} \tilde{c}(s) ds \right] \\ &= u'(\langle \tilde{c}, f \rangle) \left[-\frac{1}{\rho} \langle f, \tilde{c}\chi_{[0, M_\rho]} \rangle + \frac{f(M_\rho)}{\rho} \int_0^{M_\rho} e^{-\rho t} \tilde{c}(M_\rho - t) dt \right] \\ &= u'(\langle \tilde{c}, f \rangle) \left[-\frac{1}{\rho} \langle f, \tilde{c}\chi_{[0, M_\rho]} \rangle + \frac{f(M_\rho)}{\rho} e^{-\rho M_\rho} \int_0^{M_\rho} e^{\rho s} \tilde{c}(s) ds \right] \end{aligned}$$

that, recalling (25) and (26), can be rewritten as

$$\begin{aligned} &= u'(\langle \tilde{c}, f \rangle) \left[-\frac{1}{\rho} \langle f, \tilde{c}\chi_{[0, M_\rho]} \rangle + \frac{f(M_\rho)e^{-\rho M_\rho}}{\rho} \int_0^{M_\rho} \left(\frac{p_\rho(s)}{\eta_\rho(s)} + 1 \right) \tilde{c}(s) ds \right] \\ &= \frac{u'(\langle \tilde{c}, f \rangle)}{\rho} \left[-\langle f, \tilde{c}\chi_{[0, M_\rho]} \rangle + \left(1 - e^{-\rho M_\rho} \right) \langle p_\rho, \tilde{c}\chi_{[0, M_\rho]} \rangle + f(M_\rho)e^{-\rho M_\rho} \langle \psi, \tilde{c}\chi_{[0, M_\rho]} \rangle \right] \\ &= \frac{u'(\langle \tilde{c}, f \rangle)}{\rho} \left[\langle p_\rho - f, \tilde{c}\chi_{[0, M_\rho]} \rangle + e^{-\rho M_\rho} \langle f(M_\rho)\psi - p_\rho, \tilde{c}\chi_{[0, M_\rho]} \rangle \right]. \end{aligned}$$

Since $u' > 0$ and $p_\rho \geq f$, the previous expression is greater than

$$\frac{u'(\langle \tilde{c}, f \rangle)}{\rho} \left[e^{-\rho M_\rho} \langle f(M_\rho)\psi - p_\rho, \tilde{c} \rangle \right] > 0$$

⁹ Since \tilde{x} is decreasing (and of bounded variation), then $\tilde{c} = -\partial x$ is a (positive) Radon measure. For this reason the integrals appearing in (50) need to be interpreted as Lebesgue-Stieltjes integrals (see *e.g.* Ash, 2000 Section 1.5 page 35), more precisely $\int_0^{M_\rho} \tilde{c}(s) ds = \int_{M_\rho-t}^{M_\rho} \partial \tilde{x}(s)$ and $\int_0^t \tilde{c}(M_\rho - s) ds = \int_0^{M_\rho} \partial \tilde{x}(s)$.

where the strict positivity of the last inequality follows from

$$p_\rho(s) \geq f(M_\rho), \forall s \in [0, M_\rho]; \quad p_\rho(s) < f(M_\rho), \forall s \in [0, M_\rho]; \quad \text{supp}(\tilde{c}) \cap [0, M_\rho] \neq \emptyset.$$

By a similar argument one may prove that, if $\text{supp}(\tilde{c}) \cap (M_\rho, \bar{s}] \neq \emptyset$, then (\tilde{x}, \tilde{c}) cannot be optimal as well, so that necessarily $\text{supp}(\tilde{c}) = \{M_\rho\}$. Since the only probability measures whose support is $\{M_\rho\}$ are of type $\tilde{c} = \alpha \delta_{M_\rho}$, for some real number $\alpha \geq 0$, \tilde{x} is in Π , if and only if $\alpha = 1/M_\rho$. Then $\tilde{c} = c_\rho$ defined in (23) and then that $\tilde{x} = x_\rho$ defined in (22), as we were meant to prove. *Q.E.D.*

Proof of Remark 4.6. By linearity, \tilde{x} satisfies Definition 3.6 and hence is a stationary program. What is left to show is that (\tilde{c}, \tilde{x}) is optimal. Let c be any control admissible at \tilde{x} and let x be the associated trajectory. Then using concavity, optimality of c_ρ^i and (49), one gets

$$\liminf_{T \rightarrow +\infty} (U_T(\tilde{c}) - U_T(c)) \geq \sum_{i=1}^n \lambda_i \liminf_{T \rightarrow +\infty} (U(c_\rho^i) - U_T(c)) \geq 0.$$

Q.E.D.

Proof of Lemma 4.8. It is straightforward from definition that \hat{x} is M_ρ -periodic. Then we need to show that \hat{x} solves (30), more precisely that

$$(52) \quad \langle \hat{x}(t) - T(t)x_0, \varphi \rangle = \left\langle \int_0^t T(t-\tau) B \hat{x}(\tau, M_\rho) \delta_{M_\rho} d\tau, \varphi \right\rangle, \quad \forall \varphi \in D.$$

Since \hat{x} is M_ρ -periodic, we may assume $t \in (0, M_\rho)$. We note that $\langle T(t)x_0, \varphi \rangle_{D', D} = \int_t^S x_0(s-t) \varphi(s) ds$ while

$$\langle \hat{x}(t), \varphi \rangle_{D', D} = \int_0^S \hat{x}(t, s) \varphi(s) ds = \int_t^M x_0(s-t) \varphi(s) ds + \int_0^t x_0(s+M-t) \varphi(s) ds$$

so that the left hand side in (52) may be rewritten as follows

$$(53) \quad \langle \hat{x}(t) - T(t)x_0, \varphi \rangle_{D', D} = - \int_M^S x_0(s-t) \varphi(s) ds + \int_0^t x_0(M-\tau) \varphi(t-\tau) d\tau$$

On the other hand $\hat{x}(t, M) = x_0(M-t)$, $\hat{c}(t) = x_0(M-t) \delta_{M_\rho}$, and $\langle \delta_{M_\rho}, \psi \rangle = \psi(M) = 1$ so that

$$B \hat{x}(\tau, M_\rho) \delta_{M_\rho} = \hat{x}(\tau, M_\rho) \delta_{M_\rho} + \langle \hat{x}(\tau, M_\rho) \delta_{M_\rho}, \psi \rangle \delta_0 = x_0(M-t) (\delta_0 - \delta_{M_\rho})$$

and the right hand side in (52) is

$$\begin{aligned} & \int_0^t \langle B \hat{x}(\tau, M_\rho) \delta_{M_\rho}, T^*(t-\tau) \varphi \rangle d\tau = \int_0^t x_0(M-t) \langle (\delta_0 - \delta_{M_\rho}), T^*(t-\tau) \varphi \rangle d\tau \\ & = \int_0^t x_0(M-t) ([T^*(t-\tau) \varphi](0) - [T^*(t-\tau) \varphi](M_\rho)) d\tau \\ & = \int_0^t x_0(M-\tau) \varphi(t-\tau) d\tau - \int_{M+t-S}^t x_0(M-\tau) \varphi(t-\tau+M) d\tau \end{aligned}$$

which is equal, by means of a change of variables, to the right hand side in (53). *Q.E.D.*

Proof of Lemma 4.11. Recalling that \hat{x} is periodic of period M_ρ , and that (Lemma 4.8) $\hat{x}(t, M_\rho) = x_0(M_\rho - \sigma(t))$, one has $\langle \hat{c}(t), f \rangle = \hat{x}(t, M_\rho) f(M_\rho) = f(M_\rho) x_0(M_\rho - \sigma(t))$, so that, once set $n = [T/M_\rho]$, by suitable changes of variables we have (note that if $T < M_\rho$ then the first sum is null)

$$\begin{aligned} U_T(\hat{c}) &= \sum_{i=0}^{n-1} e^{-\rho i M_\rho} \int_0^{M_\rho} e^{-\rho t} u(f(M_\rho) x_0(M_\rho - t)) dt + e^{-\rho n M_\rho} \int_0^{T-nM_\rho} e^{-\rho t} u(f(M_\rho) x_0(M_\rho - t)) dt \\ &= \frac{1 - e^{-\rho n M_\rho}}{e^{\rho M_\rho} - 1} U_1^\rho + e^{-\rho(n+1)M_\rho} U_2^\rho(T) \end{aligned}$$

When $\rho = 0$, similarly

$$U_T(\hat{c}) = \sum_{i=0}^{n-1} \int_0^{M_\rho} u(f(M_\rho) x_0(M_\rho - t)) dt + \int_0^{T-nM_\rho} u(f(M_\rho) x_0(M_\rho - t)) dt = nU_1^0 + e^{-\rho M_\rho} U_2^0(T)$$

Q.E.D.

A.2.1. *Optimality and good controls*

In order to prove Proposition 4.14, we need some preliminary results, contained in the following lemmata.

Lemma A.11 *Assume (10) and (11) are satisfied. Let $T > 0$. Then there exists a positive constant B_T depending on T (and not depending on c and x_0), such that*

$$U_{T_0}(c) \leq B_T, \quad \forall T_0 \leq T, \quad \forall x_0 \in \Pi, \quad \forall c \in \mathcal{U}_{x_0}$$

PROOF: We prove the assertion for $\rho = 0$, as for $\rho > 0$ the result holds *a fortiori*. Let $x(t)$ be the trajectory starting at an initial datum $x_0 \in \Pi$ and driven by a control $c \in \mathcal{U}_{x_0}$. Set $\varepsilon := \min\{\lambda, S - s_1\}$, where s_1 is that in (5), and consider a function $\phi \in C^1([0, S]; \mathbb{R})$ such that $\phi(s) = 1$ for all $s \in [0, \varepsilon/2]$, $\phi(s) = 0$ for all $s \in [\varepsilon, S]$, and $\phi'(s) \leq 0$ for all $s \in [0, S]$. Note that $\psi \geq \phi$ and $\phi \in D$, so that making us of Proposition 3.4 and of (9) one has

$$(54) \quad \langle T(\varepsilon/2)x_0, \phi \rangle + \int_0^{\varepsilon/2} \langle c(\tau), B^*T^*(\varepsilon/2 - \tau)\phi \rangle d\tau = \langle x(\varepsilon/2), \phi \rangle \leq \langle x(\varepsilon/2), \psi \rangle = 1.$$

For all $\tau \in [0, \varepsilon/2]$, one defines

$$\phi_\tau(s) \equiv [T^*(\varepsilon/2 - \tau)\phi](s) = \phi(s + \varepsilon/2 - \tau)\chi_{[0, S - (\varepsilon/2 - \tau)]}(s)$$

so that $B^*T^*(\varepsilon/2 - \tau)\phi = -\phi_\tau + \langle \delta_0, \phi_\tau \rangle \psi = -\phi_\tau + \psi$, which implies

$$\langle c(\tau), B^*T^*(\varepsilon/2 - \tau)\phi \rangle = \langle c(\tau), \psi - \phi_\tau \rangle \geq \langle c(\tau), \psi - \phi \rangle, \quad \forall \tau \leq \varepsilon/2.$$

Since $\langle T(\varepsilon/2)x_0, \phi \rangle \geq 0$, from the latter and (54) one derives

$$\int_0^{\varepsilon/2} \langle c(\tau), \psi - \phi \rangle d\tau \leq 1.$$

Now the argument is iterated. A consequence of (9) is

$$x(t) = T(t - r)x(r) + \int_r^t T(t - \tau)Bc(\tau) d\tau, \quad 0 \leq r \leq t$$

which is applied with $t = r + \frac{\varepsilon}{2}$, $r = n\frac{\varepsilon}{2}$ and $n \in \{0, 1, \dots, [2T/\varepsilon]\}$, deriving

$$\int_{n\frac{\varepsilon}{2}}^{(n+1)\frac{\varepsilon}{2}} \langle c(\tau), \psi - \phi \rangle d\tau \leq \int_{n\frac{\varepsilon}{2}}^{(n+1)\frac{\varepsilon}{2}} \langle c(\tau), \psi - \phi \rangle d\tau \leq 1$$

so that

$$(55) \quad \int_0^T \langle c(t), \psi - \phi \rangle dt \leq \sum_{n=0}^{[2T/\varepsilon]} \int_{n\frac{\varepsilon}{2}}^{(n+1)\frac{\varepsilon}{2}} \langle c(\tau), \chi_{[\lambda, \bar{s}]} \rangle d\tau \leq \frac{2T}{\varepsilon} + 1.$$

Since u is concave, there exist real constants a and b such that $u(q) \leq a + bq$ for all $q \in \mathbb{R}^+$. Moreover one can choose b_1 such that $b \geq \max_{s \in [\lambda, \bar{s}]} f(s)$ so that $b(\psi - \phi) \geq f$. Then one has

$$u(\langle c(t), f \rangle) \leq a + b \langle c(t), f \rangle \leq a + b_1 b \langle c(t), \psi - \phi \rangle$$

and hence by (55)

$$U_T(c) = \int_0^T u(\langle c(t), f \rangle) dt \leq aT + b_1 b \left(1 + \frac{2T}{\varepsilon}\right) =: B_T,$$

with B_T is independent of the chosen control and on the initial datum. Since (11) implies $U_T(c)$ is increasing in T , one has $U_{T_0}(c) \leq U_T(c) \leq B_T$ and the claim. *Q.E.D.*

Remark A.12 Note that for $\rho = 0$ the value-loss function defined in (27) and in Corollary A.6 satisfies

$$(56) \quad \begin{aligned} \theta(c) \equiv \theta_0(c, x) &= -[u(\langle c, f \rangle) - u(\langle \bar{c}, f \rangle) - u'(\langle \bar{c}, f \rangle) \langle c - \bar{c}, f \rangle] \\ &= u(\beta_0) - u(\langle c, f \rangle) + \alpha_0 \langle c - \bar{c}, p_0 \rangle. \end{aligned}$$

The concavity of u implies $\theta(c) \geq 0$, with $\theta(\bar{c}) = 0$. Note also that, although evaluated in (27) along the trajectories of the system, θ is a well defined real function on D' . \square

Remark A.13 We say that the continuous function $\omega: \mathbb{R}^+ \times \mathbb{R}^+ \rightarrow \mathbb{R}^+$ is a *local modulus* if for all $b > 0$, $\lim_{a \rightarrow 0^+} \omega(a; b) = 0$. Throughout this section we will denote by $\omega(\cdot; \cdot)$ any function having such properties, or by $\omega(\cdot)$, if there is no explicit dependence from a parameter b . Then, given $x \in \Pi$, (47) implies $\alpha_0 \langle x, A^* p_0 \rangle = \alpha_0 \beta_0$ so that

$$\theta(c) = u(\beta_0) - u(\langle c, f \rangle) + \alpha_0 \langle c, p_0 \rangle - \alpha_0 \langle x, A^* p_0 \rangle.$$

In particular, given is $x_0 \in \Pi$ and a control $c \in \mathcal{U}_{x_0}$, along the associated trajectory we have

$$(57) \quad \theta(c(t)) = u(\beta_0) - u(\langle c(t), f \rangle) + \alpha_0 \langle c(t), p_0 \rangle - \alpha_0 \langle x(t), A^* p_0 \rangle.$$

It is also straightforward that $\theta: D' \rightarrow \mathbb{R}$ is a continuous function, indeed

$$|\theta(c) - \theta(c_1)| \leq |u(\langle c, f \rangle) - u(\langle c_1, f \rangle)| + \alpha_0 |c - c_1|_{D'} |p_0|_D \leq \omega_\theta(|c - c_1|_{D'})$$

for some modulus ω_θ , for any $c, c_1 \in D'$ (note indeed that u is a uniformly continuous function). \square

Lemma A.14 *Given a good control $c(\cdot) \in \mathcal{U}_{x_0}$ then the following limit exists and is finite*

$$L_c := \lim_{T \rightarrow \infty} \int_0^T \theta(c(t)) dt \in [0, +\infty).$$

PROOF: Since $\theta(c(t)) \geq 0$, the function $T \mapsto \Delta_T := \int_0^T \theta(c(t)) dt$ is positive and increasing and therefore the limit exists, and is positive (possibly equal to $+\infty$). Now observe that (57) and Proposition A.2 imply

$$(58) \quad \begin{aligned} \Delta_T &= U_T(\bar{c}) - U_T(c) + \alpha_0 \int_0^T \langle c(t), p_0 \rangle - \langle x(t), A^* p_0 \rangle dt \\ &= U_T(\bar{c}) - U_T(c) - \alpha_0 \int_0^T \frac{d}{dt} [\langle x(t), p_0 \rangle] dt = U_T(\bar{c}) - U_T(c) - \alpha_0 \langle x(T) - x_0, p_0 \rangle. \end{aligned}$$

Recalling that c is good, by definition there exists $\theta \in \mathbb{R}$ such that for all T is $U_T(\bar{c}) - U_T(c) \leq \theta$. Moreover, since $x(T) - x_0 \in L^2(0, S)$, in (58) duality pairing coincides with the scalar product in $L^2(0, S)$, and observing that p_0 lies in $L^\infty(0, S)$ (i.e. it is a bounded function), with $|p_0|_{L^\infty} \leq \beta_0 S$, by means of Hölder inequality one gets

$$|\langle x(T) - x_0, p_0 \rangle| \leq |p_0|_{L^\infty} (|x(T)|_{L^1} + |x_0|_{L^1}) = 2|p_0|_{L^\infty},$$

so that

$$\Delta_T \leq \theta + 2\alpha_0 \beta_0 S$$

and the proof is complete. Q.E.D.

Remark A.15 As a consequence of the previous lemma, for any fixed positive constant A , we have

$$\int_{t-A}^t \theta(c(\tau)) d\tau \leq \omega(1/t)$$

for a suitable modulus ω , that is, the integral is infinitesimal as t tends to $+\infty$. \square

Lemma A.16 *For any given x_0 and x_1 in Π , there exists a control $\hat{c} \in \mathcal{U}_{x_0}$, denoted with $\hat{c}(\cdot) = \hat{c}(\cdot; x_0, x_1, \bar{s})$ that drives the system from x_0 to x_1 in a time length less or equal to \bar{s} .*

PROOF: We define

$$d^+(s) := (x_0(s) - x_1(s)) \vee 0, \quad d^-(s) := (x_1(s) - x_0(s)) \vee 0, \quad \text{for } s \in [0, S],$$

so that $d^+(s)$ (respectively, $d^-(s)$) is strictly positive at those points where x_0 is strictly bigger (respectively, smaller) than x_1 . In some sense d^+ and d^- represent the values that need to be compensated by the choice of a suitable control. Since $\int_0^S x_0(s) ds = \int_0^S x_1(s) ds = 1$ we have that

$$J := \int_0^S d^+(s) ds = \int_0^S d^-(s) ds = \int_0^{\bar{s}} d^+(s) ds = \int_0^{\bar{s}} d^-(s) ds.$$

where the last equalities derive from the fact that both x_0 and x_1 are supported in $[0, \bar{s}]$.

If $J = 0$, then $x_0 = x_1$ and there is nothing to prove. Now assume $J > 0$. We define, for $t \in [0, \bar{s}]$

$$(59) \quad e^-(t) := \int_{\bar{s}-t}^{\bar{s}} d^-(\tau) d\tau.$$

where e^- measures the mass of trees to be compensated, with age in the interval $[\bar{s}-t, \bar{s}]$. Note that $e^-(0) = 0$, $e^-(\bar{s}) = J$, and e^- is an increasing function. We define also, for $t \in [0, \bar{s}]$

$$(60) \quad D^+(t, s) := \begin{cases} 0, & s > \bar{s} \\ d^+(s-t), & s \in [t, \bar{s}] \\ d^+(s-t+\bar{s}), & s \in [0, t). \end{cases}$$

This function represents the translation with replanting of the exceeding part of the initial forest x_0 . We want to prove that the following control satisfies the claim of the lemma in a time length of \bar{s} :

$$\hat{c}(t, \cdot) = \left(x_0(\bar{s}-t) - \frac{d^+(\bar{s}-t)e^-(t)}{J} \right) \delta_{\bar{s}} + d^-(\bar{s}-t) \frac{D^+(t, \cdot)}{J}, \quad t \in [0, \bar{s}].$$

Note that $x_0, x_1 \in L^2(0, S)$ imply $d^+, d^- \in L^2(0, S)$, $D^+(t) \in L^2(0, S)$ for any $t \in [0, \bar{s}]$, and $t \mapsto D(T)$ belongs to $C(0, \bar{s}; L^2(0, S))$. Moreover $e^- \in C(0, \bar{s})$, implies $\hat{c} \in L^2(0, \bar{s}, D')$ (in fact, $c \in L^2(0, \bar{s}, \mathcal{R})$).¹⁰

One needs to prove that the trajectory $\hat{x}(t)$ associated to the control $\hat{c}(t)$ satisfies $\hat{x}(\bar{s}, s) = x_1(s)$ for all s :

$$\begin{aligned} \hat{x}(t) &= I_1(t) + I_2(t) + I_3(t) := \left[T(t)x_0 + \int_0^t T(t-\tau)Bx_0(t-\tau)\delta_{\bar{s}} d\tau \right] \\ &\quad - \frac{1}{J} \left[\int_0^t d^+(\bar{s}-\tau)e^-(\tau)T(t-\tau)B\delta_{\bar{s}} d\tau \right] + \frac{1}{J} \left[\int_0^t d^-(\bar{s}-\tau)T(t-\tau)BD^+(\tau) d\tau \right]. \end{aligned}$$

Recalling the Faustmann solution (see 29), we have $I_1(\bar{s}) = x_0$. Regarding $I_3(t)$, note that

$$T(t-\tau)BD^+(\tau) = T(t-\tau) \langle D^+(\tau), \psi \rangle \delta_0 - T(t-\tau)D^+(\tau) = J\delta_0 - T(t-\tau)D^+(\tau)$$

so that $I_3(t) = I_{31}(t) + I_{32}(t)$ with

$$I_{31}(t, s) = \int_0^t d^-(\bar{s}-\tau)T(t-\tau)\delta_0 d\tau = d^-(\bar{s}-t+s)\chi_{[0,t]}(s)$$

where the last equality is derived by means of (42), while

$$I_{32}(t) = - \int_0^t d^-(\bar{s}-\tau)T(t-\tau) \frac{D^+(\tau)}{J} d\tau.$$

Now note that $T(t-\tau)D^+(\tau)(s) = D^+(\tau)(s-t+\tau)$ if $s-t+\tau \geq 0$ and 0 if $s-t+\tau < 0$, so that the last expression, evaluated at s , gives

$$I_{32}(t, s) = -\frac{1}{J} \int_{(t-s) \vee 0}^t d^-(\bar{s}-\tau)D^+(\tau)(s-t+\tau) d\tau = -\frac{1}{J} \int_{(t-s) \vee 0}^{t \wedge (\bar{s}+t-s)} d^-(\bar{s}-\tau)D^+(\tau)(s-t+\tau) d\tau.$$

By means of the definition of e^- and $D^+(\tau)$ given in (59) (60) respectively, the latter is explicitated as follow

$$I_{32}(t)(s) = \begin{cases} -\frac{1}{J} \int_{t-s}^t d^-(\bar{s}-\tau)d^+(s-t+\bar{s}) d\tau = -\frac{1}{J}d^+(s-t+\bar{s})(e^-(t) - e^-(t-s)) & s \in [0, t) \\ -\frac{1}{J} \int_0^t d^-(\bar{s}-\tau)d^+(s-t) d\tau = -\frac{1}{J}d^+(s-t)e^-(t) & s \in (t, \bar{s}] \\ -\frac{1}{J} \int_0^{(\bar{s}+t-s)} d^-(\bar{s}-\tau)d^+(s-t) d\tau = -\frac{1}{J}d^+(s-t)e^-(\bar{s}+t-s) & s > \bar{s} \end{cases}$$

¹⁰The control \hat{c} , roughly speaking, acts as follows: (i) the term $x_0(\bar{s}-t)\delta_{\bar{s}}$ cyclicly cuts and replants at age 0 the trees reaching age \bar{s} (it is indeed the Faustmann policy for $M = \bar{s}$); (ii) Note that if at time t the trees aged \bar{s} are not enough (*i.e.* if $d^-(\bar{s}-t) > 0$), one needs to cut more trees in order to reach the level $x_1(\bar{s}-t)$. The term $d^-(\bar{s}-t)D^+(t, \cdot)/J$ does the job, taking from those ages where trees are exceeding the final target x_1 and represented by $D^+(t)$ (note that the trees (re)planted at time t are those having age $\bar{s}-t$ at time \bar{s} , the final time in which we want the forest to have the configuration x_1); (iv) The term $-d^+(\bar{s}-t)e^-(t)\delta_{\bar{s}}/J$ takes into account the fact that part of exceeding trees described in (ii) (for which d^+ is strictly positive) were already cut in $[0, t)$, so that one needs only to cut (by means of $\delta_{\bar{s}}$) the remaining ones.

Regarding $I_2(t)$, since $-B\delta_{\bar{s}} = \delta_{\bar{s}} - \delta_0$, we may apply again (42) and derive

$$I_2(t, s) = \frac{1}{J} [d^+(s-t)e^-(t+\bar{s}-s)\chi_{([\bar{s}, \bar{s}+t])}(s) - d^+(\bar{s}+s-t)e^-(t-s)\chi_{(0,t)}(s)]$$

As a whole

$$\hat{x}(t, s) = \begin{cases} x_0(s) + d^-(\bar{s}-t+s) - d^+(\bar{s}-t+s)\frac{e^-(t)}{J} & \text{if } s \in [0, t) \\ x_0(s) - d^+(s-t)\frac{e^-(t)}{J} & \text{if } s \in [t, \bar{s}] \\ 0 & \text{if } s > \bar{s}. \end{cases}$$

so that at any time t the support of $x(t)$ is in $[0, \bar{s}]$, moreover $e^-(\bar{s}) = J$ implies at $t = \bar{s}$ that

$$\hat{x}(\bar{s}, s) = \begin{cases} x_0(s) + d^-(s) - d^+(s) & \text{if } s \in [0, \bar{s}] \\ 0 & \text{if } s > \bar{s} \end{cases}$$

which equals $x_1(s)$ by means of the definitions of d^+ and d^- .

Q.E.D.

Now we are ready to prove that maximal (and, in particular, optimal) controls are good controls.

Proof of Proposition 4.14: Assume by contradiction that the maximal control c^* is not good, and denote by x^* the associated trajectory. Then, given any $\theta \in \mathbb{R}$, there exists $T_\theta \geq 0$ with

$$U_{T_\theta}(c^*) - U_{T_\theta}(\bar{c}) < -\theta$$

Next we show that T_θ may be chosen arbitrarily large, for instance $T_\theta > 2\bar{s}$, if θ is chosen sufficiently large. Indeed by means of Lemma A.11 one has $\sup_{t \in [0, 2\bar{s}]} |U_t(c^*) - U_t(\bar{c})| \leq B_{2\bar{s}}$ so that for $\theta > B_{2\bar{s}}$, we have $U_T(c^*) - U_T(\bar{c}) < -\theta$ only for values of T which are greater than $2\bar{s}$.

Hence we select $\theta > 2B_{\bar{s}} > B_{2\bar{s}}$ and $T_\theta > 2\bar{s}$ and define, with the notation of the previous Lemma, the following controls: $c_1(t) = c(t; x_0, \bar{x})$ stirring the system from x_0 to \bar{x} in time \bar{s} , and $c_2(t) = c(t; \bar{x}, x^*(T_\theta))$, stirring the system from \bar{x} to $x^*(T)$ in time \bar{s} and moreover

$$\tilde{c}(t) = \begin{cases} c_1(t) & \text{if } t \in [0, \bar{s}) \\ \bar{c} & \text{if } t \in [\bar{s}, T_\theta - \bar{s}) \\ c_2(t) & \text{if } t \in [T_\theta - \bar{s}, T_\theta) \\ c^*(t) & \text{if } t \geq T_\theta \end{cases}$$

We show that \tilde{c} catches up to c^* , so that c^* cannot be maximal, yielding a contradiction. To do so it is enough to observe that, for any $T \geq T_\theta$, one has

$$\begin{aligned} U_T(\tilde{c}) - U_T(c^*) &= U_{T_\theta}(\tilde{c}) - U_{T_\theta}(c^*) = U_{\bar{s}}(\tilde{c}) - U_{\bar{s}}(\bar{c}) + U_{T_\theta}(\tilde{c}) - U_{T_\theta}(c^*) + \int_{T-\bar{s}}^T [u(\langle \tilde{c}(t), f \rangle) - u(\langle c^*(t), f \rangle)] dt \\ &\geq U_{\bar{s}}(\tilde{c}) - U_{\bar{s}}(\bar{c}) + \theta + U_{\bar{s}}(\tilde{c}(\cdot + T - \bar{s})) - U_{\bar{s}}(\langle c^*(\cdot), f \rangle) \geq \theta - 2B_{\bar{s}} > 0. \end{aligned}$$

Q.E.D.

A.3. Proofs for Section 5

A.3.1. Linear utility, positive discount

Proof of Theorem 5.1. Consider (31) when $u(r) = r$. Note that $\tilde{c}(t)$ coincides with c_ρ when $x_0 = x_\rho$, so that (31) applies also with (x_ρ, c_ρ) in place of (x^*, \tilde{c}) . If $n = \lceil T/M_\rho \rceil$, $\sigma(T) = \{T/M_\rho\}M_\rho$ then

$$U_T(\tilde{c}) - U_T(c_\rho) = \eta_\rho(1 - e^{-\rho n M_\rho}) \int_0^{M_\rho} e^{\rho\tau} (x_0(\tau) - 1/M_\rho) d\tau + e^{-\rho(n+1)M_\rho} \int_{M_\rho - \sigma(T)}^{M_\rho} e^{\rho\tau} (x_0(\tau) - 1/M_\rho) d\tau.$$

Hence when $T \rightarrow +\infty$, and once set $\phi(t) = e^{\rho t}$, we derive

$$(61) \quad U(\tilde{c}) - U(c_\rho) = \lim_{T \rightarrow +\infty} (U_T(\tilde{c}) - U_T(c_\rho)) = \eta_\rho \langle x_0 - x_\rho, \phi \rangle = \langle x_0 - x_\rho, p_\rho \rangle.$$

Now let c be arbitrarily chosen in \mathcal{U}_{x_0} , with $x(t) = x(t; c, x_0)$ the associated trajectory. Let $T > 0$, and note that Corollary A.9 implies (for $u(r) = r$, is $\alpha_\rho = 1$)

$$(62) \quad U_T(c_\rho) - U_T(c) \geq -e^{-\rho T} \langle x_\rho - x(T), p_\rho \rangle + \langle x_\rho - x_0, p_\rho \rangle,$$

Coupling the previous relation with (61) we derive

$$U_T(\hat{c}) - U_T(c) \geq \omega(T)$$

for a suitable function ω , $\omega(T) \rightarrow 0$ as $T \rightarrow +\infty$, which implies the thesis. Q.E.D.

A.3.2. Linear Utility, Null discount

Average of a trajectory. Assume x is the trajectory associated to some initial datum x_0 and driven by an admissible control c . We denote by means of $x^A(t)$ the *average* of the trajectory over a time interval $[0, t]$ that is

$$(63) \quad x^A(t) := \frac{1}{t} \int_0^t x(s; x_0, c) \, ds.$$

Lemma A.17 Assume $\rho = 0$, $c \in \mathcal{U}_{x_0} \cap L^\infty(0, +\infty; D')$ a good control, and $x_0 \in \Pi$. Then

$$(64) \quad \left\langle x^A(t), h \right\rangle \rightarrow \langle \bar{x}, h \rangle, \text{ as } t \rightarrow +\infty \quad \text{for all } h \in D.$$

PROOF: The proof is obtained by applying Theorem 9.1.3 in Zaslavski (2006) to the modified objective functional $\tilde{U}_T(c) = \int_0^T u(\langle c(t), f \rangle) - u(\langle c_\rho, f \rangle) \, dt$, so that controls which are good for $\tilde{U}_T(c)$ satisfy Hypothesis (b) of Theorem 9.1.3. Note that the weak convergence derived in Zaslavski (2006) is in our case the weak convergence described in (64). Note also that, since the control is bounded in D' and so is the trajectory, we can modify the functional $\tilde{U}_T(c)$ outside a ball in D' where both are contained in order to verify the coercivity assumption (1.9) page 260 (although not recalled in the statement of the theorem, it is indeed needed). Eventually, using the arguments of Corollary A.6 we can see that Assumption 1 page 259 of Zaslavski (2006) is satisfied and the maximum (there cited as a minimum) is attained by the golden rule couple (\bar{x}, \bar{c}) (which is true *a fortiori* for the modified functional). Q.E.D.

Proof of Theorem 5.5. We divide the long proof into several steps.

Claim 1: \hat{c} is a maximal control. We consider the trajectory $\hat{x} = x(\cdot; x_0, \hat{c})$, starting at x_0 and driven by the control \hat{c} . The control \hat{c} is good. Indeed, we let $T > 0$ be arbitrarily fixed, and we apply Lemma 4.11 with $\rho = 0$ and $u(r) = r$ both to \hat{c} and \bar{c} deriving

$$U_T(\bar{c}) - U_T(\hat{c}) = f(M) \int_{M-\sigma(T)}^M \left(x_0(\tau) - \frac{1}{M} \right) d\tau = f(M) \left[\int_{M-\sigma(T)}^M x_0(\tau) d\tau - \left\{ \frac{T}{M} \right\} \right] \geq -f(M)$$

which implies \hat{c} is good. If by contradiction \hat{c} is not maximal, then there exists a control \bar{c} in \mathcal{U}_{x_0} and some $\hat{T}, a > 0$ such that for all $T \geq \hat{T}$

$$(65) \quad U_T(\hat{c}) - U_T(\bar{c}) < -a.$$

Now assume $R \geq 3\hat{T}$. We integrate on $[0, R]$ and divide by R the left hand side, obtaining

$$(66) \quad \frac{1}{R} \int_0^R (U_T(\hat{c}) - U_T(\bar{c})) \, dT = \frac{1}{R} \int_0^{\hat{T}} (U_T(\hat{c}) - U_T(\bar{c})) \, dT + \frac{1}{R} \int_{\hat{T}}^R (U_T(\hat{c}) - U_T(\bar{c})) \, dT$$

where the first integral converges to 0 for $R \rightarrow \infty$ while, for R large enough, the second is smaller than $-\frac{5}{6}a$ as a consequence of (65). Then for a sufficiently large R one has

$$(67) \quad \frac{1}{R} \int_0^R (U_T(\hat{c}) - U_T(\bar{c})) \, dT < -\frac{2}{3}a.$$

On the other hand, if \tilde{x} is the trajectory starting at x_0 and driven by control \bar{c} , and $\tilde{x}^A(t)$ the respective average, from Corollary A.9, one has

$$(68) \quad U_T(\hat{c}) - U_T(\bar{c}) \geq \int_0^T \frac{d}{dt} \langle \hat{x}(t) - \tilde{x}(t), p \rangle \, dt = \langle \hat{x}(T) - \tilde{x}(T), p \rangle$$

so that, integrating on $[0, R]$ and dividing by R , one gets

$$\frac{1}{R} \int_0^R (U_T(\hat{c}) - U_T(\bar{c})) \, dT \geq \langle \hat{x}^A(R) - \bar{x}^A(R), p_0 \rangle \geq -\frac{1}{3}a$$

for a sufficiently large R , as $\langle \hat{x}^A(R) - \bar{x}^A(R), p_0 \rangle \rightarrow 0$, as $R \rightarrow +\infty$, in view of (64): a contradiction.

Claim 2: the control \hat{c} is not optimal. Assume c_1 is the control defined in (33), and $x_1(t, s) \equiv x_1(t, s; c_1, x_0)$, the associated trajectory. With reference to the notation there introduced, we assume also (and the reasons will be clear in a short while) that N is big enough so that $f(s_{N-1}) > 0$, which is true as f is continuous and $f(M) > 0$. Moreover we assume that x_0 satisfies the condition

$$(69) \quad \int_{s_{N-2}}^{s_{N-1}} x_0(r) \, dr > 0.$$

In order to prove \hat{c} not optimal, it is sufficient to show that there exists $a > 0$ such that

$$U_{T_n}(\hat{c}) - U_{T_n}(c_1) = -a, \quad \forall T_n = \frac{M}{N} + nM, \text{ with } n \in \mathbb{N}.$$

At some initial time interval, that is, for t in $[0, M/N]$, we have

$$x_1(t, s) = \sum_{j=0}^N x_0(s_j + s - t) \chi_{[0, t]}(s) + x_0(s - t) \sum_{j=1}^N \chi_{[s_j, s_j + t]}(s)$$

Afterwards, the solution becomes periodic of period M , and repeatedly equal to

$$x_1(t, s) = \begin{cases} \chi_{[t - \frac{M}{N}, t]}(s) \sum_{j=1}^N x_0(s_{j-1} + s + \frac{M}{N} - t) & t \in [\frac{M}{N}, M] \\ \chi_{[0, t - M]}(s) \sum_{j=1}^N x_0(s_{j-1} + s + \frac{N+1}{N}M - t) + \\ \quad + \chi_{[t - \frac{M}{N}, M]}(s) \sum_{j=1}^N x_0(s_{j-1} + s + \frac{M}{N} - t) & t \in [M, M + \frac{M}{N}] \end{cases}$$

(the general formula is obtained by replacing t with $\xi(t) = t - [\frac{M}{N}]M - \frac{M}{N}$ in the right hand side). Recalling that for any t in $[0, M/N]$, we have $x_1(t, s_j) = x_0(s_j - t)$, and that $\langle \delta_{s_j}, f \rangle = f(s_j)$,

$$U_{\frac{M}{N}}(c_1) = \sum_{j=1}^N f(s_j) \int_0^{\frac{M}{N}} x_0(s_j - t) \, dt = \sum_{j=1}^N f(s_j) \int_{s_{j-1}}^{s_j} x_0(r) \, dr.$$

By means of the periodicity of x_1 for $t \geq M/N$, we then derive

$$U_{T_n}(c_1) = U_{\frac{M}{N}}(c_1) + n f(M) \left[\int_{\frac{M}{N}}^M x_1(t, M) \, dt + \int_M^{M + \frac{M}{N}} x_1(t, M) \, dt \right].$$

Note that, for $\frac{M}{N} < t < M$, we have $x_1(t, M) = 0$, while, for $M < t < M + \frac{M}{N}$, we have $x_1(t, M) = \sum_{j=1}^N x_0(s_{j-1} + M + \frac{M}{N} - t)$ so that

$$\int_M^{M + \frac{M}{N}} x_1(t, M) \, dt = \sum_{j=1}^N \int_M^{M + \frac{M}{N}} x_0(s_{j-1} + M + \frac{M}{N} - t) \, dt = \sum_{j=1}^N \int_{s_{j-1}}^{s_j} x_0(r) \, dr = 1$$

which implies

$$U_{T_n}(c_1) = \sum_{j=1}^N f(s_j) \int_{s_{j-1}}^{s_j} x_0(r) \, dr + n f(M).$$

The difference between such utility and that yielded by means of the Faustmann policy \hat{c} is then

$$(70) \quad U_{T_n}(\hat{c}) - U_{T_n}(c_1) = f(M) \int_{M - \frac{M}{N}}^M x_0(r) \, dr - \sum_{j=1}^N f(s_j) \int_{s_{j-1}}^{s_j} x_0(r) \, dr = - \sum_{j=1}^{N-1} f(s_j) \int_{s_{j-1}}^{s_j} x_0(r) \, dr =: -a$$

Note that $a > 0$, as the last term of the sum above is strictly positive in view of (69). As a consequence, \widehat{c} is not optimal. The proof for the case when (69) is not satisfied is easily obtained by applying a control c_2 in place of c_1 shaped as follows. More precisely, if

$$m := \max \left\{ j : 1 \leq j \leq N-1, \int_{s_{j-1}}^{s_j} x_0(r) dr > 0 \right\}$$

(a maximum exists as the forest has positive density and extension 1), we define $\tau := \frac{M}{N}(N-1-m)$, and

$$(71) \quad c_2(t) = \bar{c} \chi_{[0,\tau]}(t) + c_1(t-\tau) \chi_{[\tau,+\infty)}(t)$$

that is, the control which coincides with \bar{c} until the associated trajectory x_2 yields a positive integral (with respect to s) on $[s_{N-2}, s_{N-1}]$, and behaves like c_1 afterwards.

Claim 3: an optimal control does not exist. We assume by contradiction that $\tilde{c}(t) \in U_{x_0}$ is an optimal control. Then in particular, given any $\varepsilon > 0$, there exists T_ε such that

$$(72) \quad U_T(\tilde{c}) - U_T(\widehat{c}) \geq -\varepsilon \text{ and } U_T(\tilde{c}) - U_T(c_1) \geq -\varepsilon \quad \forall T \geq T_\varepsilon.$$

On the other hand (70) implies, for a sufficiently small $\nu \in [0, M]$ not depending on n , that

$$U_T(c_1) - U_T(\widehat{c}) \geq \frac{a}{2}, \quad \forall T \in [T_n, T_n + \nu].$$

from which, if $n_\varepsilon \in N$ is such that $T_n > T_\varepsilon$ for all $n \geq n_\varepsilon$, we derive also

$$(73) \quad U_T(\tilde{c}) - U_T(\widehat{c}) \geq \frac{a}{2} - \varepsilon, \quad \forall T \in [T_n, T_n + \nu], \quad \forall n \geq n_\varepsilon.$$

We show first that

$$(74) \quad \liminf_{n \rightarrow +\infty} \frac{1}{T_n + \nu} \int_0^{T_n + \nu} (U_T(\tilde{c}) - U_T(\widehat{c})) dT \geq \frac{\nu a}{4}.$$

Indeed $\int_0^{T_n + \nu} (U_T(\tilde{c}) - U_T(\widehat{c})) dT = A + B_n + C_n$, where $A \equiv \int_0^{T_{n_\varepsilon}} (U_T(\tilde{c}) - U_T(\widehat{c})) dT$ and in view of (72)

$$B_n \equiv \sum_{i=n_\varepsilon}^{n-1} \int_{T_i + \nu}^{T_{i+1}} (U_T(\tilde{c}) - U_T(\widehat{c})) dT \geq -\varepsilon \nu (n - n_\varepsilon)$$

and in view of (73)

$$C_n \equiv \sum_{i=n_\varepsilon}^n \int_{T_i}^{T_i + \nu} (U_T(\tilde{c}) - U_T(\widehat{c})) dT \geq \left(\frac{a}{2} - \varepsilon\right) \nu (n - n_\varepsilon + 1)$$

so that (recall that $T_n = nM + M/N$), if $\omega(1/n)$ is infinitesimal as $n \rightarrow \infty$, we have

$$\frac{1}{T_n + \nu} \int_0^{T_n + \nu} (U_T(\tilde{c}) - U_T(\widehat{c})) dT \geq \frac{\nu}{M} \left(\frac{a}{2} - 2\varepsilon\right) + \omega(1/n).$$

Choosing $\varepsilon \leq a(2 - M/2)$, and passing to limits we obtain (74).

On the other hand, if we choose $\gamma = \widehat{x}(t, M)$ in Lemma A.8, we make use of Proposition A.2 (note that that $p_0(s) = \beta_0 s \psi(s)$, $\langle \delta_0, p_0 \rangle = 0$, $A^* p_0(s) = \beta_0 \chi_{[0, s]}(s)$) we obtain

$$\begin{aligned} \langle \tilde{c}(t) - \widehat{c}(t), f \rangle &\leq \langle \tilde{c}(t) - \widehat{c}(t), p_0 \rangle - \langle \widehat{x}(t) - x_0, A^* p_0 \rangle \\ &= -\frac{d}{dt} \langle \widehat{x}(t) - \widehat{x}(t), p_0 \rangle - \langle \widehat{x}(t) - x_0, A^* p_0 \rangle = -\frac{d}{dt} \langle \widehat{x}(t) - \widehat{x}(t), p_0 \rangle \end{aligned}$$

where $\langle \widehat{x}(t) - x_0, A^* p_0 \rangle = 0$ follows from Remark A.7. Then, for all T , integrating on $[0, T]$ one derives

$$(75) \quad U_T(\tilde{c}) - U_T(\widehat{c}) \leq \int_0^T \frac{d}{dt} \langle \widehat{x}(t) - \widehat{x}(t), p_0 \rangle dt = \langle \widehat{x}(T) - \widehat{x}(T), p_0 \rangle$$

Now, we denote by $x^A(t)$ the average of the trajectory x , as defined in (63). Integrating on $[0, S]$ both sides of (75) and dividing by S one has

$$(76) \quad \frac{1}{S} \int_0^S (U_T(\tilde{c}) - U_T(\bar{c})) dT \leq \langle \tilde{x}^A(S) - \bar{x}^A(S), p_0 \rangle.$$

Since \hat{c} and \tilde{c} are good, by Lemma A.17 one has

$$\langle \tilde{x}^A(S) - \bar{x}^A(S), p_0 \rangle \xrightarrow{S \rightarrow \infty} 0, \quad S \rightarrow \infty$$

which together with (76) contradicts (74). Q.E.D.

A.3.3. Strictly concave utility, null discount

Proof of Theorem 5.9. We set $H := L^2(0, S)$, recalling that $D \hookrightarrow L^2(0, S)$ (with continuous inclusion) and use this property when necessary without further notice. Since $c(t) \in L^\infty(0, +\infty; \mathcal{R})$ we have that $K := \sup |x(t)|_{\mathcal{R}} < +\infty$. Let $\varepsilon > 0$ be fixed. We have to prove that there exists $t(\varepsilon) > 0$, such that

$$(77) \quad i(t) := |x(t) - \bar{x}|_H \leq \varepsilon, \quad \text{for all } t \geq t(\varepsilon).$$

Any $c \in \mathcal{R}$ may be decomposed as follows. We recall that by assumption M is the unique positive maximum point of $f(s)/s$ (which implies $p_0(s) - f(s) \geq 0$ for all $s \in [0, S]$, and with the equality holding only at $s = 0$ and $s = M$). Then, from the continuity of f and for a sufficiently small $\xi > 0$, there exists $\zeta(\xi) > 0$, with $\lambda < M - \zeta(\xi)$, such that $|s - M| \geq \zeta(\xi)$ implies $p(s) - f(s) \geq \xi$. Note that $\zeta(\xi) \xrightarrow{\xi \rightarrow 0} 0$. Furthermore, since c is a positive measure with $\text{supp}(c) \subseteq [\lambda, S]$, then

$$c = c_n + c_f, \quad \text{with } c_n = c\nu_\xi \text{ and } c_f = c(1 - \nu_\xi),$$

where ν_ξ is a $[0, 1]$ -valued smooth cut-off function with $\nu_\xi(s) \equiv 1$ for $|s - M| \leq \zeta(\xi)/2$ and $\nu_\xi(s) \equiv 0$ when $|s - M| \geq \zeta(\xi)$. Now, we may assume $t > S$. As a consequence, in (9) we have $T(t)x_0(s) = 0$ for all $s \in [0, S]$, and $T(t - \tau)Bc(\tau) = 0$ for all $\tau \leq t - S$, so that in (77)

$$(78) \quad x(t) - \bar{x} = \int_{t-S}^t T(t - \tau)Bc(\tau) d\tau - \bar{x} = \int_{t-S}^t T(t - \tau)B(c(\tau) - \bar{c}) d\tau = I_1(t, \xi) + I_2(t, \xi) + I_3(t, \xi),$$

where

$$\begin{aligned} I_1(t, \xi) &:= \int_{t-S}^t T(t - \tau)Bc_f(\tau) d\tau, & I_2(t, \xi) &:= \int_{t-S}^t T(t - \tau)B(c_n(\tau) - |c_n(\tau)|_{\mathcal{R}}\delta_M) d\tau \\ & & \text{and } I_3(t, \xi) &:= \int_{t-S}^t T(t - \tau)B(|c_n(\tau)|_{\mathcal{R}}\delta_M - \bar{c}) d\tau. \end{aligned}$$

In next steps we will estimate the H -norm of $I_1(t, \xi)$, $I_2(t, \xi)$ and $I_3(t, \xi)$.

Step 1: A preliminary estimate. We start by estimating the quantity $|c_f|_{D'}$. Given $x \in \Pi$, and $c \in \mathcal{R}$, from Remark A.13 and from the concavity of u follows

$$\begin{aligned} \theta(c) &= \theta(c) - \theta(\bar{c}) = u(\langle \bar{c}, f \rangle) - u(\langle c, f \rangle) + \alpha_0 \langle p, c - \bar{c} \rangle \\ &\geq -\alpha_0 \langle c - \bar{c}, f \rangle + \alpha_0 \langle c - \bar{c}, p_0 \rangle = \alpha_0 \langle c - \bar{c}, p_0 - f \rangle = \alpha_0 \langle c, p_0 - f \rangle \geq \alpha_0 \xi |c_f|_{\mathcal{R}}, \end{aligned}$$

as c is positive and $\langle \bar{c}, p_0 - f \rangle = 0$. Observe that $D \hookrightarrow C^0([0, S])$ with continuous inclusion so that $\mathcal{R} \hookrightarrow D'$, and in particular $|c_f|_{D'} \leq C|c_f|_{\mathcal{R}}$ for some fixed positive constant $C > 0$. Then from the previous inequalities follows that

$$(79) \quad |c_f|_{D'} \leq C|c_f|_{\mathcal{R}} \leq \frac{C}{\alpha_0 \xi} \theta(c), \quad \forall x \in \Pi, \quad \forall c \in \mathcal{R}.$$

As observed in Remark A.13, the estimate does not depend on x .

Step 2: Estimate on $I_1(t, \xi)$. Note that $\|T(t)\|_{L(D')} \leq 1$, so that (79) implies

$$|I_1(t, \xi)|_H \leq \|B\|_{L(D')} \int_{t-S}^t |c_f(\tau)|_{D'} d\tau \leq \frac{C\|B\|_{L(D')}}{\alpha_0 \xi} \int_{t-S}^t \theta(c(\tau)) d\tau$$

which implies that for some modulus ω_1 one has

$$|I_1(t, \xi)|_H \leq \omega_1(1/t; \xi).$$

Step 3: Estimate on I_2 . Given $\phi \in D$ and any $c \in \mathcal{R}$, one has

$$\begin{aligned} |\langle c - |c|_{\mathcal{R}} \delta_M, \phi \rangle| &\leq |c|_{\mathcal{R}} \max_{|s-M| \leq \zeta(\xi)} |\phi(s) - \phi(M)| \leq |c|_{\mathcal{R}} \int_{M-\zeta(\xi)}^{M+\zeta(\xi)} |\phi'(s)| ds \\ &\leq \sqrt{2\zeta(\xi)} |c|_{\mathcal{R}} \left(\int_{M-\zeta(\xi)}^{M+\zeta(\xi)} |\phi'(s)|^2 ds \right)^{1/2} \leq \sqrt{2\zeta(\xi)} |c|_{\mathcal{R}} |\phi|_D \end{aligned}$$

which in particular gives, for any $\tau \in [t-S, t]$,

$$(80) \quad |c_n(\tau) - |c_n(\tau)|_{\mathcal{R}} \delta_M|_{D'} \leq \sqrt{2\zeta(\xi)} |c_n(\tau)|_{\mathcal{R}} \leq \sqrt{2\zeta(\xi)} K$$

and then $|c_n(\cdot) - |c_n(\cdot)|_{\mathcal{R}} \delta_M|_{L^2(t-S, t; D')} \leq \sqrt{S} \sqrt{2\zeta(\xi)} K$. By means of Proposition 3.1 page 212 of Bensoussan et al. (2007) (whose assumptions were checked in the proof of Proposition 3.3 (ii)), the $L^2(t-S, t; D')$ -norm can be estimated by means of the H -norm of the convolution defining I_2 so to obtain

$$|I_2(t, \xi)|_H \leq C \sqrt{2\zeta(\xi)} \sqrt{S} K \leq \omega_2(\xi)$$

where C is a constant independent on the control $c(\cdot)$ and ω_2 is a modulus (with $\omega_2(\xi) \xrightarrow{\xi \rightarrow 0} 0$, uniformly with respect to t).

Step 4: Estimate on $I_3(t, \xi)$. In order to estimate $I_3(t, \xi)$ we need to define, besides ξ , some other parameters. Since u is strictly convex and differentiable, recalling that $\alpha_0 = u'(\bar{c}, p_0)$, $\beta_0 = \langle \bar{c}, p_0 \rangle$ and defining $\beta_\eta = (1 + \eta) \langle \bar{c}, p_0 \rangle = (1 + \eta) \beta_0$, one may consider $\gamma > 0$ and $0 < \eta < 1$ such that

$$\gamma = u(\beta_0) - u(\beta_\eta) + \alpha_0 \eta \beta_0 > 0.$$

and moreover

$$\Delta = -[u'(\beta_\eta) - \alpha_0] < 0$$

since u' is strictly decreasing. Note that γ as a function of η is strictly increasing and attains the value zero at $\eta = 0$, so that its inverse $\eta(\gamma)$ is well defined and enjoys the same property, in particular $\eta(\gamma) \xrightarrow{\gamma \rightarrow 0} 0$. As a consequence, Δ may itself be regarded as a function of γ , with $\Delta(\gamma) \xrightarrow{\gamma \rightarrow 0} 0$.

Now we rewrite $I_3(t, \xi)$ as the sum of four terms and estimate them separately:

$$\begin{aligned} (81) \quad I_3(t, \xi) &= \int_{t-S}^t T(t-\tau) B(|c_n(\tau)|_{\mathcal{R}} \delta_M - \bar{c}) d\tau \equiv I_{31}(t, \xi) + I_{32}(t, \xi) + I_{33}(t, \xi) + I_{34}(t, \xi) \\ &= \int_{t-S}^t \left(|c_n(\tau)|_{\mathcal{R}} - \frac{1}{M} - \eta - \frac{\theta(|c_n(\tau)|_{\mathcal{R}} \delta_M)}{\Delta} \right) T(t-\tau) B \delta_M d\tau \\ &\quad + \int_{t-S}^t \left(\frac{\theta(|c_n(\tau)|_{\mathcal{R}} \delta_M) - \theta(c(\tau))}{\Delta} \right) T(t-\tau) B \delta_M d\tau \\ &\quad \quad \quad + \int_{t-S}^t \left(\frac{\theta(c(\tau))}{\Delta} \right) T(t-\tau) B \delta_M d\tau + \eta \int_{t-S}^t T(t-\tau) B \delta_M d\tau. \end{aligned}$$

To estimate $I_{34}(t, \xi)$ it suffices to observe that for every we fixed γ ,

$$(82) \quad |I_{34}(t, \xi)|_H \leq \|B\| |\delta_M|_{D'} S \eta =: \omega_{34}(\gamma; \xi)$$

where ω_{34} is a local modulus.

Next, to estimate $I_{33}(t, \xi)$ Remark A.15 implies

$$(83) \quad |I_{33}(t, \xi)|_H \leq \|B\| |\delta_M|_{D'} \int_{t-S}^t \left| \frac{\theta(c(\tau))}{\Delta} \right| d\tau \leq \omega_{33}(1/t; \gamma, \xi).$$

for some local modulus ω_{33} . Then, to estimate $I_{32}(t, \xi)$, we note that

$$|\theta(c(\tau)) - \theta(c_n(\tau))| \leq \omega_\theta(|c_f|_{D'}) \leq \omega_\theta((C/\alpha_0\xi)\theta(c(\tau)))$$

so that, as a consequence of Remark A.15 for a sufficiently large t , the quantity $\theta(c(\tau))$ is infinitesimal, at least outside a subset of $[t - S, t]$ of arbitrarily small Lebesgue measure, so that one has

$$\int_{t-S}^t \frac{|\theta(c(\tau)) - \theta(c_n(\tau))|}{\Delta} d\tau \leq \hat{\omega}(1/t; \gamma, \xi)$$

for some local modulus $\hat{\omega}$. Moreover, in view of (80) one has

$$|\theta(|c_n(\tau)|_{\mathcal{R}\delta_M}) - \theta(c_n(\tau))| \leq \omega_\theta(|c_n(\tau)|_{\mathcal{R}\delta_M} - c_n(\tau)|_{D'}) \leq \omega_\theta(\sqrt{2\zeta(\xi)}K)$$

so that

$$\int_{t-S}^t \frac{|\theta(|c_n(\tau)|_{\mathcal{R}\delta_M}) - \theta(c_n(\tau))|}{\Delta} d\tau \leq \tilde{\omega}(\gamma; \xi),$$

for some modulus $\tilde{\omega}$. Hence, once set $\omega_{32} = \|B\|\delta_M|_{D'}(\hat{\omega} + \tilde{\omega})$, one derives

$$(84) \quad I_{32}(t, \xi) \leq \omega_{32}(1/t; \gamma, \xi).$$

We are left with the estimate on I_{31} . By definition of $\theta(c)$ and concavity of u

$$-\theta(c) + \alpha_0 \langle c - \bar{c}, p_0 \rangle + \gamma - \alpha_0 \eta \langle \bar{c}, p_0 \rangle = u(\beta_0) - u(\beta_\eta) \leq u'(\beta_\eta) \langle c - (1 + \eta)\bar{c}, f \rangle$$

where we used first the definition of $\theta(c)$ and the strict concavity of u . Recalling that $\langle \bar{c}, p \rangle = \langle \bar{c}, f \rangle = \beta_0$, we obtain

$$\theta(c) \geq \gamma + \alpha_0 \langle c, p_0 \rangle - \alpha_0 \beta_0 - \alpha_0 \eta \beta_0 - u'(\beta_\eta) \langle f, c \rangle + u'(\beta_\eta)(1 + \eta)\beta_0$$

so that, once set the expression above becomes

$$\theta(c) \geq \gamma - \eta \alpha_0 \Delta + \langle c - \bar{c}, \alpha_0 p - u'(\beta_\eta) f \rangle.$$

For $c \equiv |c_n(t)|_{\mathcal{R}\delta_M}$ the previous inequality reads

$$\varphi(\tau) := |c_n(t)|_{\mathcal{R}} - \frac{1}{M} - \eta - \frac{\theta(|c_n(t)|_{\mathcal{R}\delta_M})}{\Delta} \leq 0.$$

Now note that, as a consequence of (78), step 2 and step 3, (81) (82) (83) (84), Hölder inequality and the fact that $\langle x(t) - \bar{x}, \psi \rangle = 0$, one has

$$|\langle x(t) - \bar{x} - I_{31}(t, \xi), \psi \rangle| = \left| \int_{t-S}^t \varphi(\tau) \langle T(t - \tau)B\delta_M, \psi \rangle ds \right| + |\langle I_1 + I_2 + I_{32} + I_{33} + I_{34}, \psi \rangle| \leq \omega_{31}(1/t; \gamma, \xi),$$

with $\omega_{31} = \sqrt{S}(\omega_1 + \omega_2 + \omega_{32} + \omega_{33} + \omega_{34})$. Now since

$$\langle T(t - \tau)B\delta_M, \psi \rangle = \langle \delta_0 - \delta_M, T^*(t - \tau)\psi \rangle = \psi(t - \tau) - \psi(t - \tau + M)$$

the previous estimate may be rewritten as

$$\left| \int_{t-S}^t \varphi(\tau) (\psi(t - \tau) - \psi(t - \tau + M)) d\tau \right| \leq \omega_5(1/t; \gamma, \xi)$$

By definition of ψ we have $0 \leq \psi(t - \tau) - \psi(t - \tau + M) \leq 1$, moreover $\varphi(\tau) \leq 0$, so that the integrand of the last equation is always negative. Moreover, from the definition of $\psi(s)$ it is easily shown that on some interval $[t_1, t_2] \subseteq [0, S]$ one has $\psi(t - \tau) - \psi(t - \tau + M) \geq c$ for a suitable $c > 0$. As a consequence

$$c \int_{t_1}^{t_2} |\varphi(t - \sigma)| d\sigma \leq \int_{t_1}^{t_2} |\varphi(t - \sigma)| |\psi(\sigma) - \psi(\sigma + M)| d\sigma \leq \omega_5(1/t; \gamma, \xi).$$

Now, since the previous equation holds for all t iterating the argument $[S/(t_2 - t_1)] + 1$ times, we obtain

$$\int_0^S |\varphi(t - \sigma)| d\sigma \leq \frac{1}{c} \frac{S}{(t_2 - t_1)} \omega_5(1/t; \gamma, \xi) = \omega_4(1/t; \gamma, \xi).$$

We are now ready to draw the conclusion. It is sufficient to choose, in order, ξ, γ sufficiently small and $t(\varepsilon)$ sufficiently large to derive from all of the previous steps that $t \geq t(\varepsilon)$ implies $i(t) \leq \varepsilon$, as we intended to show. *Q.E.D.*

Proof of Theorem 5.10. In view of Proposition 4.14 it is sufficient that we compare \bar{c} to good controls. Indeed for any good control c , Theorem 5.9 implies

$$\lim_{T \rightarrow +\infty} \int_0^T \frac{d}{dt} [\langle \bar{x} - x_{\bar{x},c}(t), \alpha_0 p_0 \rangle] dt = \lim_{T \rightarrow +\infty} \langle \bar{x} - x_{\bar{x},c}(T), \alpha_0 p_0 \rangle = 0.$$

so that

$$\begin{aligned} \liminf_{T \rightarrow +\infty} (U_T(\bar{c}) - U_T(c)) &= \liminf_{T \rightarrow +\infty} \int_0^T (U_T(\bar{c}) - U_T(c) + \frac{d}{dt} [\langle \bar{x} - x_{\bar{x},c}(t), \alpha_0 p_0 \rangle]) dt \\ &= \liminf_{T \rightarrow +\infty} \int_0^T [u(\langle f, \bar{c} \rangle) - u(\langle f, c(t) \rangle) - \alpha_0 \langle x(t), A^* p_0 \rangle + \langle c(t), p_0 \rangle] u dt \geq 0 \end{aligned}$$

where the last inequality follows from Corollary A.6. *Q.E.D.*

Proof of Theorem 5.12 To prove (i) we first build the candidate optimal control \bar{c} as limit of a suitable sequence. We consider the quantity

$$(85) \quad S \equiv \sup_{c \in \mathcal{U}_{x_0}^{K,\lambda}} \left(\limsup_{T \rightarrow +\infty} [U_T(c) - U_T(\bar{c})] \right),$$

(S possibly equal to $+\infty$). Let $\{c_n\}$ be a maximizing sequence in $\mathcal{U}_{x_0}^{K,\lambda}$, and let θ be the function defined in (56). Then for $T > 0$, we have

$$(86) \quad U_T(c_n) - U_T(\bar{c}) = - \int_0^T \left(\theta(c_n(t)) + \frac{d}{dt} \alpha_0 \langle x_n(t), p_0 \rangle \right) dt = - \int_0^T \theta(c_n(t)) dt - \alpha_0 \langle x_n(T) - x_0, p_0 \rangle.$$

Since $|p_0|_\infty < +\infty$ and $|x_n(t)|_{L^1} = |x_0|_{L^1} = 1$ then $|\alpha_0 \langle x_n(t) - x_0, p_0 \rangle| \leq 2\alpha_0 |p_0|_\infty$, so that, being $\theta(c_n(t))$ positive for all t , it may happen either (a) $\lim_{T \rightarrow +\infty} (U_T(c_n) - U_T(\bar{c})) = -\infty$, which we may exclude as $\{c_n\}$ is a maximizing sequence, or (b) $\liminf_{T \rightarrow +\infty} (U_T(c_n) - U_T(\bar{c})) \geq -\infty$, the latter implying c_n is a good control. Note also that from (86) and the positivity of θ follows also

$$U_T(c_n) - U_T(\bar{c}) \geq -2\alpha_0 |p_0|_\infty$$

implying that $S < +\infty$. Hence with no loss of generality, we may assume that c_n are good controls. Note also that Lemma A.14 and Theorem 5.9 imply that for any good control in $\mathcal{U}_{x_0}^{K,\lambda}$, c the following limit exists and is finite:

$$\lim_{T \rightarrow +\infty} (U_T(c) - U_T(\bar{c})) = - \lim_{T \rightarrow +\infty} \int_0^T \left(\theta(c(t)) + \frac{d}{dt} \alpha_0 \langle x(t), p_0 \rangle \right) dt = -L_c - \alpha_0 \langle \bar{x} - x_0, p_0 \rangle,$$

so that (85) implies

$$(87) \quad S = \lim_{n \rightarrow \infty} \lim_{T \rightarrow +\infty} [U_T(c_n) - U_T(\bar{c})].$$

Now, set $h > 0$ and $L_h^2([0, +\infty); D')$ the Hilbert space of all functions $\phi: [0, +\infty) \rightarrow D'$ such that the norm $\int_0^{+\infty} e^{-ht} |\phi(t)|_{D'}^2 dt < +\infty$. It is a tedious but standard proof that $\mathcal{U}_{x_0}^{K,\lambda}$ is a sequentially weakly compact subset of $L_h^2([0, +\infty); D')$. Hence from $\{c_n(\cdot)\}$ one may extract a subsequence weakly converging to some $\tilde{c}(\cdot) \in L_h^2([0, +\infty); D')$ and $\tilde{c}(\cdot) \in \mathcal{U}_{x_0}^{K,\lambda}$.

Next we show that \tilde{c} is optimal. Note that

$$\liminf_{T \rightarrow +\infty} (U_T(\tilde{c}) - U_T(c)) \geq \liminf_{T \rightarrow +\infty} [U_T(\tilde{c}) - U_T(\bar{c})] - \limsup_{T \rightarrow +\infty} [U_T(c) - U_T(\bar{c})] \geq \liminf_{T \rightarrow +\infty} [U_T(\tilde{c}) - U_T(\bar{c})] - S$$

so that it is enough to prove that $\liminf_{T \rightarrow +\infty} [U_T(\tilde{c}) - U_T(\bar{c})] \geq S$ to derive the positivity of the right hand side, and the conclusion. We start by proving that

$$\limsup_{T \rightarrow +\infty} [U_T(\tilde{c}) - U_T(\bar{c})] = S.$$

By definition of S , the left hand side is smaller than the right hand side, while the reverse inequality is obtained by observing that $c \mapsto \limsup_{T \rightarrow +\infty} [U_T(c) - U_T(\bar{c})]$ is a concave functional on $\mathcal{U}_{x_0}^{K,\lambda}$ (note that $\mathcal{U}_{x_0}^{K,\lambda}$ is a convex subset of $L_h^2([0, +\infty); D')$) so that passing to limits one obtains

$$\lim_n \lim_{T \rightarrow +\infty} [U_T(c_n) - U_T(\bar{c})] \leq S.$$

Note that, since $\{c_n(\cdot)\}$ is a maximizing sequence for the limsup, it is also maximizing for the liminf, more precisely

$$\sup_{c \in \mathcal{U}_{x_0}^{K,\lambda}} \left[\liminf_{T \rightarrow +\infty} (U_T(c) - U_T(\bar{c})) \right] = \sup_{c \in \mathcal{U}_{x_0}^{K,\lambda}} \left[\limsup_{T \rightarrow +\infty} (U_T(c) - U_T(\bar{c})) \right] = S.$$

Then, arguing as before about concavity of $c \mapsto \liminf_{T \rightarrow +\infty} [U_T(c) - U_T(\bar{c})]$, one derives

$$\liminf_{T \rightarrow +\infty} [U_T(\tilde{c}) - U_T(\bar{c})] = S$$

and the conclusion. To prove (ii), let c be a good admissible control, x the associated trajectory, and $R > 0$:

$$\begin{aligned} \frac{1}{R} \int_0^R (U_T(c) - U_T(\bar{c})) \, dT &= \frac{1}{R} \int_0^R \int_0^T \theta(c(t)) + \frac{d}{dt} \alpha_0 \langle x(t), p_0 \rangle \, dt \, dT \\ &= \frac{1}{R} \int_0^R \int_0^T \theta(c(t)) \, dt \, dT + \frac{\alpha_0}{R} \int_0^R [\langle x(T) - \bar{x}, p_0 \rangle - \langle x_0 - \bar{x}, p_0 \rangle] \, dT. \end{aligned}$$

On one hand, Lemma A.14 implies the first addendum converges to L_c when $R \rightarrow \infty$, on the other hand, as a consequence of Theorem 9.1.3 p. 260 in Zaslavski (2006) (cf. proof of Proposition 5.5 for details)

$$\frac{1}{R} \int_0^R \langle x(T) - \bar{x}, p_0 \rangle \, dT = \left\langle \frac{1}{R} \int_0^R x(T) \, dT - \bar{x}, p_0 \right\rangle \rightarrow 0, \quad R \rightarrow \infty.$$

Hence

$$\lim_{R \rightarrow +\infty} \frac{1}{R} \int_0^R (U_T(c) - U_T(\bar{c})) \, dT = L_c + \alpha_0 \langle \bar{x} - x_0, p_0 \rangle,$$

so that the limit exists and is finite. We can now argue as in the proof of part (i) and prove that, given c in $\mathcal{U}_{x_0}^{K,\lambda}$, there exists a control \tilde{c} that maximizes $\limsup_{R \rightarrow +\infty} \frac{1}{R} \int_0^R (U_T(c) - U_T(\bar{c})) \, dT$ in $\mathcal{U}_{x_0}^{K,\lambda}$. Indeed

$$\begin{aligned} \lim_{R \rightarrow +\infty} \sup \frac{1}{R} \int_0^R (U_T(\tilde{c}) - U_T(c)) \, dT \\ \geq \lim_{R \rightarrow +\infty} \sup \frac{1}{R} \int_0^R (U_T(\tilde{c}) - U_T(\bar{c})) \, dT - \lim_{R \rightarrow +\infty} \sup \frac{1}{R} \int_0^R (U_T(c) - U_T(\bar{c})) \, dT \geq 0 \end{aligned}$$

and this implies $\limsup_{T \rightarrow \infty} (U_T(\tilde{c}) - U_T(c)) \geq 0$, for all c in $\mathcal{U}_{x_0}^{K,\lambda}$, hence \tilde{c} is maximal in $\mathcal{U}_{x_0}^{K,\lambda}$. *Q.E.D.*

Remark A.18 The control that we have proved to be maximal is exactly the one that minimizes

$$\lim_{T \rightarrow \infty} \int_0^T \theta(c(t)) \, dt.$$