Optimal switching over multiple regimes

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Abstract

This paper studies the optimal switching problem for a general one-dimensional diffusion with multiple (more than two) regimes. This is motivated in the real options literature by the investment problem of a firm managing several production modes while facing uncertainties. A viscosity solutions approach is employed to carry out a fine analysis on the associated system of variational inequalities, leading to sharp qualitative characterizations of the switching regions. These characterizations, in turn, reduce the switching problem into one of finding a finite number of threshold values in state that would trigger switchings. The results of our analysis take several qualitatively different forms depending on model parameters, and the issue of when and where it is optimal to switch is addressed. The general results are then demonstrated by the three-regime case, where a quasi-explicit solution is obtained, and a numerical procedure to find these critical values is devised in terms of the expectation functionals of hitting times for one-dimensional diffusions.

Key words : Optimal multiple switching, variational inequalities, switching regions, viscosity solutions, hitting times of Itô diffusion, real options.

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1 Introduction

Optimal multiple switching is the problem of determining an optimal sequence of stopping times for a stochastic process with several regimes (or modes). This is a classical and important problem, extensively studied since the late seventies. It has recently received renewed and increasing interest due to many applications in economics and finance, especially to real options. Actually, optimal switching provides a suitable model to capture the value

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of managerial flexibility in making decisions under uncertainty, and has been used in the pioneering works by Brennan and Schwarz [7] for resource extraction, and Dixit [11] for production facility problems.

The optimal two-regime switching problem has been the most largely studied in the literature, and is often referred to as the starting-and-stopping problem : in the case of a geometric Brownian motion and some special profit functions on infinite horizon, Brekke and Øksendal [6], Duckworth and Zervos [13], and Zervos [21] apply a verification approach for solving the variational inequality associated with this impulse control problem (see Bensoussan and Lions [4] or Tang and Yong [20]). Various extensions of this model are solved in Pham and Vath [19]. Hamadène and Jeanblanc [16] consider a finite horizon starting-and-stopping problem by using reflected backward stochastic differential equations (BSDEs). Bayraktar and Egami [3] employ optimal stopping theory for studying the optimal two-regime switching problem on infinite horizon for one-dimensional diffusions. In this latter framework, Guo and Tomecek [15] solve some special cases by connecting optimal switching to singular control problem.

The applications of the starting-and-stopping problem to real options, for example the management of a power plant, are limited to the case of two modes, e.g. operating and closed. In practice, however, the efficient management of a power plant requires more than two production modes to include intermediate operating modes corresponding to different subsets of turbine running. Such an example of multiple switching problems applied to energy tolling agreements was considered by Carmona and Ludkovski [8], and Deng and Xia [10], who focus mainly on a numerical resolution based on Monte-Carlo regressions. Yet, there is little work addressing a complete treatment and mathematical resolution of the optimal multiple switching problem, especially the determination of the switching regions. While the regularity analysis of the value function is similar to the two-regime case, and is addressed in [18], the difficulty with a multi-regime problem in the determination of the switching regions is evident: In sharp contrast with the two-regime problem, a multiple switching problem needs to decide not only when to switch, but also *where* to switch. Recently, Djehiche, Hamadène and Popier [12], and Hu and Tang [17] have studied optimal multiple switching problems for general adapted processes by means of reflected BSDEs, and they are mainly concerned with the existence and uniqueness of solution to these reflected BSDEs. However, the important issue as to which regime to optimally switch has been left completely open.

In this paper, we consider the optimal multiple switching problem on infinite horizon for a general one-dimensional diffusion. The multiple regimes are differentiated via their profit functions, which are of very general form. The numbering on the regimes is ordered by increasing level of profitability. The transition from one regime to another one is realized sequentially at random times (which are part of the decisions), and incurs a fixed cost. Our objective is to provide an explicit characterization of the switching regions showing when and where it is optimal to change the regime. We adopt a direct solution method via the viscosity solutions technique. By carrying out a detailed analysis of the system of variational inequalities associated with the optimal switching problem, we give a sharp qualitative description of the switching regions. Specifically, we give conditions under which one should switch to a regime with higher profit, and to a regime with lower profit, and we identify these destination regimes. This extends results of [19] obtained in the tworegime case, to the multiple regime case. The switching regions take various structures, depending on model parameters via explicit conditions, which have meaningful economic interpretations. It appears that in some situations, it is optimal to switch to a regime for a range of state values, and to switch to another regime for a different set of state values. Such a feature is new with respect to the two-regime case, where we can only switch to one regime. We showcase our general results by the three-regime case, where we present a complete picture of the situations as to when and where it is optimal to switch, and we reduce the problem into one of finding a finite number of threshold values of the switching regions. We also design an algorithm to compute these critical values based on the computations of expectation functionals of hitting times for one-dimensional diffusions.

The rest of the paper is organized as follows. In Section 2, we formulate precisely the optimal multiple switching problem. We recall in Section 3 the system of variational inequalities and the boundary data that characterize theoretically the value functions in the viscosity sense. The continuous differentiability of the value functions serves also as an important result in our subsequent analysis. Section 4 is devoted to the qualitative description of the switching regions. In the three-regime case considered in Section 5, we give a complete solution by reducing the original switching problem into one of finding a finite number of threshold values of the switching regions. Finally, in Section 6, we provide a numerical procedure for computing these critical values.

2 Model and problem formulation

2.1 General setup and assumptions

We present our general model and emphasize the key assumptions. The state process X is a one-dimensional diffusion on $(0, \infty)$ whose dynamics are given by :

$$dX_t = b(X_t)dt + \sigma(X_t)dW_t, \qquad (2.1)$$

where W is a standard Brownian motion on a filtered probability space $(\Omega, \mathcal{F}, \mathbb{F} = (\mathcal{F}_t)_{t \geq 0}, P)$ satisfying the usual conditions, and b, σ are measurable functions on $(0, \infty)$. We assume that the SDE (2.1) has a unique strong solution, denoted by X^x , given an initial condition $X_0 = x \in (0, \infty)$. The coefficients b and σ satisfy the growth condition

$$xb(x), \quad \frac{1}{2}\sigma^2(x) \leq C(1+x^2), \quad \forall x > 0,$$
 (2.2)

for some positive constant C, and the nondegeneracy condition :

$$\sigma(x) > 0, \quad \forall x > 0. \tag{2.3}$$

We also make the standing assumption that 0 and ∞ are *natural* boundaries for X (see e.g. [5] for boundary classification of one-dimensional diffusions), and for all $t \ge 0$,

$$E[X_t^x] \longrightarrow 0$$
 when x goes to 0 (2.4)

$$X_t^x \longrightarrow \infty, \ a.s.$$
 when x goes to ∞ (2.5)

Remark 2.1 A sufficient condition ensuring (2.4) is to assume that the functions b and σ on $(0, \infty)$ can be extended to 0, with $b(0) = \sigma(0) = 0$, and are Lipschitz on $[0, \infty)$ (this is the case for example of the geometric brownian motion). Then, we have $X_t^0 = 0$, and a standard estimate yields for all $t \ge 0$, the existence of a constant C(t) s.t. $E[X_t^x] \le C(t)x$, x > 0, and so (2.4) is satisfied.

It would be of interest to state conditions under which (2.5) holds. We have not been able to address this issue in general. However, we can provide a class of sde (2.1) for which it is valid. Suppose that $b(x) = x\beta(x)$, $\sigma(x) = x\gamma$ for some measurable function β on $(0, \infty)$ and positive constant $\gamma > 0$. Fix some constant $\mu \in \mathbb{R}$, and set $Y_t = e^{\mu t} \ln X_t$. By Itô's formula, we have

$$dY_t = e^{\mu t} [\mu \ln X_t + \beta(X_t) - \frac{1}{2}\gamma^2] dt + e^{\mu t}\gamma dW_t.$$
 (2.6)

By choosing the coefficient β such that

$$|\beta(x) + \mu \ln x| \le \beta_0, \ x > 0,$$
 (2.7)

for some positive constant β_0 , we deduce from comparison theorem for side that $Y_0 + \tilde{Y}_t^0 \leq Y_t \leq Y_0 + \bar{Y}_t^0$, $t \geq 0$ a.s. where

$$\begin{split} \tilde{Y}_{t}^{0} &= \int_{0}^{t} \left[-\beta_{0} - \frac{1}{2} \gamma^{2} \right] e^{\mu(s-t)} ds + \int_{0}^{t} \gamma e^{\mu(s-t)} dW_{s} \\ \bar{Y}_{t}^{0} &= \int_{0}^{t} \left[\beta_{0} - \frac{1}{2} \gamma^{2} \right] e^{\mu(s-t)} ds + \int_{0}^{t} \gamma e^{\mu(s-t)} dW_{s}. \end{split}$$

Thus, the condition (2.7) yields the following bounds for X^x :

$$x^{e^{-\mu t}} \exp(\tilde{Y}^0_t) \le X^x_t \le x^{e^{-\mu t}} \exp(\bar{Y}^0_t), \quad t \ge 0, \ a.s.$$

It is clear that (2.5) holds true. Moreover, we have for all $t \ge 0$, $X_t^0 = 0$ (which means that 0 is an absorbing state), and $X_t^x \to 0$ as x goes to zero, which implies (2.4). In the particular cases when $\mu = 0$ and β is a constant, X is a geometric brownian motion (GBM), and when $\mu > 0$ and $\beta(x) = \beta_0 - \mu \ln x$, X is a geometric mean-reverting (GMR) process.

Throughout the paper, we denote by \mathcal{L} the infinitesimal generator of the diffusion X, i.e.

$$\mathcal{L}\varphi(x) = b(x)\varphi'(x) + \frac{1}{2}\sigma^2(x)\varphi''(x).$$

For any positive constant r, the ordinary differential equation of second order :

$$r\varphi - \mathcal{L}\varphi = 0 \tag{2.8}$$

has two linearly independent solutions. These solutions are uniquely determined (up to a multiplication), if we require one of them to be increasing and the other decreasing. We denote by ψ_r^+ the increasing solution and ψ_r^- the decreasing solution. They are called

fundamental solutions of (2.8) and all other solutions can be expressed as their linear combinations. Moreover, since 0 and ∞ are natural boundaries for X, we have (see [5]) :

$$\psi_r^+(0) = \psi_r^-(\infty) = 0, \quad \psi_r^+(\infty) = \psi_r^-(0) = \infty.$$
 (2.9)

In the sequel, r will be fixed, and hence we omit the dependence in r by setting $\psi^+ = \psi_r^+$ and $\psi^- = \psi_r^-$.

Canonical examples

Our two basic examples in finance for X satisfying the above assumptions are

• a geometric Brownian motion (GBM)

$$dX_t = \mu X_t dt + \vartheta X_t dW_t,$$

where μ and ϑ are two constants with $\vartheta > 0$. The two fundamental solutions of (2.8) are given by $\psi^{-}(x) = x^{m^{-}}, \ \psi^{+}(x) = x^{m^{+}}, \ \text{where } m^{-} < 0 < m_{+} \ \text{are the roots of } \frac{1}{2}\vartheta^{2}m^{2} + (\mu - \frac{1}{2}\vartheta^{2})m - r = 0.$

• a geometric mean-reverting (GMR) process

$$dX_t = \mu X_t (\bar{y} - \ln X_t) dt + \vartheta X_t dW_t, \qquad (2.10)$$

where μ , \bar{y} , ϑ are constants with μ , $\vartheta > 0$. This last example for X is well suited for modelling commodity and electricity prices since they are expected to be long-term stationary.

The operational regimes are characterized by their running reward functions $f_i : \mathbb{R}_+ \to \mathbb{R}$, $i \in \mathbb{I}_d = \{1, \ldots, d\}$. We assume that for each $i \in \mathbb{I}_d$, the function f_i is nonnegative, without loss of generality (w.l.o.g.) $f_i(0) = 0$, continuous, and satisfies the linear growth condition:

$$f_i(x) \leq C(1+|x|), \quad \forall x \in \mathbb{R}_+, \tag{2.11}$$

for some positive constant C. The numbering $i = 1, \ldots, d$, on the regimes is ordered by increasing level of profitability, which roughly means that the sequence of functions f_i is increasing in i. The ordering condition on the profit functions will be detailed later (see condition (**Hf**) in Section 4).

Switching from regime *i* to *j* incurs an instantaneous cost, denoted by g_{ij} , with the convention $g_{ii} = 0$. The following triangular condition is reasonable:

$$g_{ik} < g_{ij} + g_{jk}, \quad j \neq i, k,$$
 (2.12)

which means that it is less expensive to switch directly in one step from regime i to k than in two steps via an intermediate regime j. Notice that a switching cost g_{ij} may be negative, and condition (2.12) for i = k prevents an arbitrage by simply switching back and forth, i.e.

$$g_{ij} + g_{ji} > 0, \quad i \neq j \in \mathbb{I}_d.$$

$$(2.13)$$

2.2 The optimal switching problem

A decision (strategy) for the operator is an impulse control α consisting of a double sequence $\tau_1, \ldots, \tau_n, \ldots, \iota_1, \ldots, \iota_n, \ldots, n \in \mathbb{N}^* = \mathbb{N} \setminus \{0\}$, where $\tau_n \in \mathbb{N}^*$, are \mathbb{F} -stopping times in $[0, \infty]$, denoted by $\tau_n \in \mathcal{T}, \tau_n < \tau_{n+1}$ and $\tau_n \to \infty$ a.s., representing the decision on "when to switch", and ι_n are \mathcal{F}_{τ_n} -measurable valued in \mathbb{I}_d , representing the new value of the regime at time τ_n until time τ_{n+1} or the decision on "where to switch". We denote by \mathcal{A} the set of all such impulse controls. Given an initial regime value $i \in \mathbb{I}_d$, and a control $\alpha = (\tau_n, \iota_n)_{n \geq 1} \in \mathcal{A}$, we denote

$$I_t^i = \sum_{n \ge 0} \iota_n \mathbf{1}_{[\tau_n, \tau_{n+1})}(t), \quad t \ge 0, \quad I_{0^-}^i = i,$$

which is the piecewise constant process indicating the regime value at any time t. Here, we set $\tau_0 = 0$ and $\iota_0 = i$. We notice that I^i is a cadlag process, possibly with a jump at time 0 if $\tau_1 = 0$ and so $I_0^i = \iota_1$.

The expected total profit of running the system when initial state is (x, i) and using the impulse control $\alpha = (\tau_n, \iota_n)_{n \ge 1} \in \mathcal{A}$ is

$$J_i(x,\alpha) = E\left[\int_0^\infty e^{-rt} f_{I_t^i}(X_t^x) dt - \sum_{n=1}^\infty e^{-r\tau_n} g_{\iota_{n-1},\iota_n}\right].$$

Here r > 0 is a positive discount factor, and we use the convention that $e^{-r\tau_n(\omega)} = 0$ when $\tau_n(\omega) = \infty$. The objective is to maximize this expected total profit over \mathcal{A} . Accordingly, we define the value functions

$$v_i(x) = \sup_{\alpha \in \mathcal{A}} J_i(x, \alpha), \quad x > 0, \ i \in \mathbb{I}_d.$$
(2.14)

We shall see later that for r large enough, the expectation defining $J_i(x, \alpha)$ is well-defined and the value function v_i is finite.

3 Dynamic programming PDE characterization

In this section, we state some general PDE characterizations of the value functions by using the dynamic programming approach.

We first state the linear growth property and the boundary condition on the value functions.

Lemma 3.1 There exists some positive constant ρ such that for $r > \rho$, the value functions $v_i, i \in \mathbb{I}_d$, are finite on $(0, \infty)$. In this case, the value functions $v_i, i \in \mathbb{I}_d$, satisfy a linear growth condition

$$0 \leq v_i(x) \leq K(1+x), \quad \forall x > 0,$$

for some positive constant K. Moreover, we have for all $i \in \mathbb{I}_d$,

$$v_i(0^+) := \lim_{x \downarrow 0} v_i(x) = \max_{j \in \mathbb{I}_d} (-g_{ij})$$

Proof. We first show the finiteness of the value functions and their linear growth. By induction, we obtain that for all $N \ge 1$, $\tau_1 \le \ldots \le \tau_N$, $\kappa_0 = i$, $\kappa_n \in \mathbb{I}_d$, $n = 1, \ldots, N$ (see [19]):

$$-\sum_{n=1}^N e^{-r\tau_n}g_{\kappa_{n-1},\kappa_n} \leq \max_{j\in\mathbb{I}_d}(-g_{ij}), \quad a.s.$$

By definition, using the latter inequality and the growth condition (2.11), we have for all $i \in \mathbb{I}_d$, x > 0, and $\alpha \in \mathcal{A}$:

$$J_{i}(x,\alpha) \leq E\Big[\int_{0}^{\infty} e^{-rt} f_{I_{t}^{i}}(X_{t}^{x}) dt + \max_{j \in \mathbb{I}_{d}}(-g_{ij})\Big]$$

$$\leq E\Big[\int_{0}^{\infty} e^{-rt} C(1+|X_{t}^{x}|) dt + \max_{j \in \mathbb{I}_{d}}(-g_{ij})\Big].$$
(3.1)

Now, a standard estimate on the process $(X_t)_{t\geq 0}$, based on Itô's formula and Gromwall's lemma, yields

$$E[|X_t^x|] \le E[(X_t^x)^2]^{\frac{1}{2}} \le e^{2Ct}(1+x),$$

for some positive constant C (independent of t and x). Plugging the above inequality into (3.1), and from the arbitrariness of $\alpha \in \mathcal{A}$, we get

$$v_i(x) \leq \frac{C}{r} + C(1+x) \int_0^\infty e^{(2C-r)t} dt + \max_{j \in \mathbb{I}_d} (-g_{ij}).$$

We therefore have the finiteness of the value functions if r > 2C, in which case the value functions satisfy the linear growth condition.

We now turn to the boundary data at 0 of the value functions. By considering the particular strategy $\tilde{\alpha} = (\tilde{\tau}_n, \tilde{\kappa}_n)$ of immediately switching from the initial state (x, i) to state $(x, j), j \in \mathbb{I}_d$, at cost g_{ij} and then doing nothing, i.e. $\tilde{\tau}_1 = 0, \tilde{\kappa}_1 = j, \tilde{\tau}_n = \infty, \tilde{\kappa}_n = j$ for all $n \geq 2$, we have

$$J_i(x,\tilde{\alpha}) = E\bigg[\int_0^\infty e^{-rt} f_j(X_t^x) dt - g_{ij}\bigg].$$

Since f_j is nonnegative, and by the arbitrariness of j, we obtain:

$$0 \leq \max_{j \in \mathbb{I}_d} (-g_{ij}) \leq v_i(x).$$

To obtain the reverse inequality, we introduce the conjugate of f_i :

$$\widetilde{f}_i(y)$$
 := $\sup_{x \ge 0} [f_i(x) - xy], \quad y > 0,$

and we notice by the growth linear condition (2.11) on f_i that for y large enough (y > C), $\tilde{f}_i(y)$ is finite. Following the proof of Lemma 3.1 in [19], we thus obtain for all x > 0, y > C, and $\alpha \in \mathcal{A}$,

$$\begin{aligned} J_i(x,\alpha) &\leq E\Big[\int_0^\infty e^{-rt} \left(yX_t^x + \tilde{f}_i(y)\right) dt + \max_{j\in\mathbb{I}_d}(-g_{ij})\Big] \\ &= \int_0^\infty e^{-rt} yE[X_t^x] dt + \int_0^\infty e^{-rt} \tilde{f}_i(y) dt + \max_{j\in\mathbb{I}_d}(-g_{ij}). \end{aligned}$$

By sending x to zero and then y to infinity, using (2.4) and recalling that $\tilde{f}_i(\infty) = f_i(0) = 0$ for $i \in \mathbb{I}_d$, we have $J_i(0^+, \alpha) \leq \max_{j \in \mathbb{I}_d} (-g_{ij})$. From the arbitrariness of α , we obtain

$$v_i(0^+) \le \max_{j \in \mathbb{I}_d} (-g_{ij}),$$

which concludes our proof.

Remark 3.1 The original problem is called well-posed if the value functions are finite (otherwise, ill-posed). An ill-posed problem is a wrongly modelled problem where the tradeoff is not set right so it is possible to push the profit arbitrarily high. The preceding result indicates that the discount rate must be sufficiently high to avoid an ill-posed problem.

However, for the GMR case (see (2.10)), the problem is well-posed so long as r > 0, which is due to the mean-reverting properties of the GMR state process itself. Indeed, by Itô's formula, we have :

$$E[|X_t^x|^2] = x^2 + \int_0^t \left(2\mu \overline{y} + \nu - 2\mu \ln X_t^x\right) |X_t^x|^2 dt.$$

A basic study of the function $x \to (2\mu \overline{y} + \nu - 2\mu \ln x)x^2$ shows that it admits a global maximum point. As such, there exists some constant K such that

$$E[|X_t^x|^2] \le x^2 + Kt.$$

Using the same arguments as in the proof of Lemma 3.1, we thus obtain the finiteness of the value functions when r > 0.

The dynamic programming principle combined with the notion of viscosity solutions are known to be a general and powerful tool for characterizing the value function of a stochastic control problem via a PDE representation. In our context, we have the following PDE characterization on the value functions.

Theorem 3.1 The value functions v_i , $i \in \mathbb{I}_d$, are the unique viscosity solutions to the system of variational inequalities :

$$\min\left\{rv_i - \mathcal{L}_i v_i - f_i, \ v_i - \max_{j \neq i} (v_j - g_{ij})\right\} = 0, \quad x \in (0, \infty), \quad i \in \mathbb{I}_d,$$
(3.2)

in the following sense:

(1) Viscosity property. For each $i \in \mathbb{I}_d$, v_i is a viscosity solution to

$$\min\left\{rv_{i} - \mathcal{L}_{i}v_{i} - f_{i}, v_{i} - \max_{j \neq i}(v_{j} - g_{ij})\right\} = 0, \quad x \in (0, \infty).$$
(3.3)

(2) <u>Uniqueness property</u>. If w_i , $i \in \mathbb{I}_d$, are viscosity solutions with linear growth conditions on $(0,\infty)$ and boundary conditions $w_i(0^+) = \max_{j \in \mathbb{I}_d} [-g_{ij}]$ to the system of variational inequalities (3.2), then $v_i = w_i$ on $(0,\infty)$.

We also quote the useful smooth fit property on the value functions, proved in [18].

Theorem 3.2 For all $i \in \mathbb{I}_d$, the value function v_i is continuously differentiable on $(0, \infty)$.

Remark 3.2 (1) The viscosity and smooth-fit properties are proved in [18]. Actually, in this paper, the statement was proved for Lipschitz f_i , and this was used for deriving a priori the Lipschitz property on the value functions v_i . The Lipschitz condition on f_i can be relaxed by assuming only continuity and linear growth condition on f_i . In this case, we proved similarly the viscosity property by means of discontinuous viscosity solutions. The uniqueness result for continuous viscosity solutions is proved in [19] when X is a geometric brownian motion. A straightforward modification of their proof, see their Step 1 in the construction of a strict supersolution, provides the result for discontinuous viscosity solutions and for general one-dimensional diffusion satisfying condition (2.2). This implies as usual the continuity of the value functions. The arguments in [18] for proving the smooth fit property rely on the viscosity property of the value functions and the nondegeneracy (2.3) of the diffusion coefficient σ , and so do not require the Lipschitz condition on f_i .

(2) For fixed $i \in \mathbb{I}_d$, we also have uniqueness of viscosity solution to equation (3.3) in the class of continuous functions with linear growth condition on $(0, \infty)$ and given boundary condition on 0. In the next sections, we shall use either uniqueness of viscosity solutions to the system (3.2) or for fixed *i* to equation (3.3).

(3) For fixed $i \in \mathbb{I}_d$, and by setting $h_i = \max_j (v_j - g_{ij})$, we notice from the free-boundary characterization (3.2) that v_i may be represented as the value function of the optimal stopping problem :

$$v_i(x) = \sup_{\tau \in \mathcal{T}} E\Big[\int_0^\tau e^{-rt} f_i(X_t^x) dt + e^{-r\tau} h_i(X_\tau^x)\Big], \quad x > 0.$$
(3.4)

For any regime $i \in \mathbb{I}_d$, we introduce the switching region :

$$S_i = \left\{ x \in (0,\infty) : v_i(x) = \max_{j \neq i} (v_j - g_{ij})(x) \right\}.$$
 (3.5)

 S_i is a closed subset of $(0, \infty)$ and corresponds to the region where it is optimal for the controller to change regime. The complement set C_i of S_i in $(0, \infty)$ is the so-called continuation region :

$$C_i = \left\{ x \in (0,\infty) : v_i(x) > \max_{j \neq i} (v_j - g_{ij})(x) \right\},$$

where it is optimal to stay in regime *i*. In this open domain, the value function v_i is smooth $(C^2 \text{ on } C_i)$ and satisfies in a classical sense :

$$rv_i(x) - \mathcal{L}_i v_i(x) - f_i(x) = 0, \quad x \in \mathcal{C}_i.$$

Remark 3.3 There are no isolated points in a switching region S_i : for any $x_0 \in S_i$, there exists some $\varepsilon > 0$ s.t. either $(x_0 - \varepsilon, x_0)$ or $(x_0, x_0 + \varepsilon)$ is included in S_i . Indeed, otherwise, recalling that $C_i = (0, \infty) \setminus S_i$ is open, one could find some $\varepsilon > 0$ s.t. $(x_0 - \varepsilon, x_0) \cup (x_0, x_0 + \varepsilon) \subset C_i$. Hence, on $(x_0 - \varepsilon, x_0) \cup (x_0, x_0 + \varepsilon)$, v_i satisfies : $rv_i - \mathcal{L}v_i - f_i = 0$. By the smooth-fit property of v_i at x_0 , this implies that v_i is actually C^2 at x_0 and satisfies $rv_i - \mathcal{L}v_i - f_i = 0$ on $(x_0 - \varepsilon, x_0 + \varepsilon)$. Hence, x_0 lies in C_i , a contradiction.

Following the general theory of optimal stopping and dynamic programming principle, see e.g. [14], we introduce the sequence of stopping times and regime decisions for $v_i(x)$:

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$$\tau_1^* = \inf \{ t \ge 0 : X_t^x \in \mathcal{S}_i \}, \qquad \iota_1^* \in \arg \max_{j \ne i} (v_j - g_{ij})(X_{\tau_1^*}^x)$$
(3.6)

$$\tau_n^* = \inf \{ t \ge \tau_{n-1}^* : X_t^x \in \mathcal{S}_{\iota_{n-1}^*} \}, \qquad \iota_n^* \in \arg \max_{j \ne \iota_{n-1}^*} (v_j - g_{\iota_{n-1}^*, j})(X_{\tau_{n-1}^*}^x).$$
(3.7)

The condition

$$\limsup_{x \to \infty} \frac{x}{\psi^+(x)} = 0 \tag{3.8}$$

ensures that the sequence given in (3.6)-(3.7) is optimal for $v_i(x)$, see Proposition 2.2 in [3]. In the remainder of the paper, we assume that (3.8) is satisfied.

Remark 3.4 Notice that condition (3.8) is satisfied for $r > \mu$ in the example of a GBM, since in this case $\psi^+(x) = x^{m^+}$ with $m^+ > 1$. We now check for the case of a GMR process. In this case as defined in (2.10), the ODE (2.8) becomes

$$r\varphi - \mu x(\overline{y} - \ln x)\varphi' - \frac{1}{2}\nu^2 x^2 \varphi'' = 0.$$
(3.9)

By a change of variable $z = \ln x + \frac{\beta}{\delta}$, the preceding equation becomes

$$\varphi'' - \delta z \varphi' - \gamma \varphi = 0, \qquad (3.10)$$

where $\beta = 1 - \frac{2\mu\bar{y}}{\nu^2}$, $\delta = \frac{2\mu}{\nu^2} > 0$, and $\gamma = \frac{2r}{\nu^2} > 0$. Using power series as likely candidate solutions to our ODE, we obtain two linearly independent solutions ϕ_1 and ϕ_2 described as follows :

$$\phi_1(z) = \sum_{n=0}^{\infty} a_{2n} z^{2n} \text{ and } \phi_2(z) = \sum_{n=0}^{\infty} a_{2n+1} z^{2n+1},$$
 (3.11)

where $a_0 = a_1 = 1$, $a_{2n+2} = \frac{\prod_{k=0}^n (2k\alpha+\gamma)}{(2n+2)!} > 0$, and $a_{2n+3} = \frac{\prod_{k=0}^n [(2k+1)\alpha+\gamma]}{(2n+3)!} > 0$, for all $n \ge 2$. The radius of convergence of both functions ϕ_1 and ϕ_2 are infinite, as such they are of class C^{∞} . Let us now denote by χ^+ , one of the two fundamental solutions to (3.10), which is non-decreasing and satisfies

$$\lim_{z \to +\infty} \chi^+(z) = +\infty, \quad \lim_{z \to -\infty} \chi^+(z) = 0.$$
(3.12)

 χ^+ could therefore be written as a linear combination of ϕ_1 and ϕ_2 :

$$\chi^+ = A\phi_1 + B\phi_2$$

where we see that A and B should be strictly non-positive due to condition (3.12) and the fact that ϕ_1 and ϕ_2 are respectively even and odd functions. Notice that proving that ψ^+ ,

one of the two fundamental solutions of (3.9) as described by (2.9), satisfies condition (3.8) is equivalent to proving that

$$\lim_{z \to +\infty} \frac{e^z}{\chi^+(z)} = 0.$$
 (3.13)

Since $\chi^+(z) \ge \min(A, B)(\phi_1 + \phi_2)(z)$, it suffices to show there exists some $\varepsilon > 0$ such that for all z large enough,

$$\phi_1(z) + \phi_2(z) \ge e^{(1+\varepsilon)z}.$$
 (3.14)

Actually, fix some $\varepsilon > 0$, and develop the function $e^{(1+\varepsilon)z}$ into power series. Then, by using the Stirling formula, we may compare the coefficients $b_n = \frac{(1+\varepsilon)^n}{n!}$ of the latter power series with the coefficients a_n of $(\phi_1 + \phi_2)$. We obtain $b_n < a_n$ for n sufficiently large enough, which shows (3.14), and hence (3.13) or (2.9).

We end this section with some notation. We introduce the functions

$$\hat{V}_i(x) = E\left[\int_0^\infty e^{-rt} f_i(X_t^x) dt\right], \quad x > 0, \ i \in \mathbb{I}_d,$$
(3.15)

which are particular solutions to

$$rw - \mathcal{L}w - f_i = 0. aga{3.16}$$

All other solutions to (3.16) are in the form

$$w = \hat{V}_i + A\psi^- + B\psi^+$$

for some constants A and B.

Remark 3.5 Notice that \hat{V}_i corresponds to the expected profit $J_i(x, \alpha)$ where α is the strategy of never switching. In particular, we obviously have $v_i \geq \hat{V}_i$. Moreover, from definition (3.15) and condition (2.5) together with **(Hf)** to be introduced below, we may apply monotone convergence theorem to get :

$$\lim_{x \to \infty} (\hat{V}_j - \hat{V}_i)(x) = \frac{(f_j - f_i)(\infty)}{r}, \quad \forall i, j \in \mathbb{I}_d.$$
(3.17)

4 Qualitative properties of the switching regions

In this section, we focus on the qualitative aspects of deriving the solution to the switching problem. Basically, we raise the following questions : When and where does one switch? In view of the general dynamic programming results stated in the previous section, the answer to these questions is provided by the description of the switching regions S_i and the determination of the argument maximum in $v_i(x) = \max_{j \neq i} (v_j - g_{ij})(x), i \in \mathbb{I}_d$.

From the definition (3.5) of the switching regions, we have the elementary decomposition property :

$$\mathcal{S}_i = \bigcup_{j \neq i} \mathcal{S}_{ij}, \quad i \in \mathbb{I}_d,$$

where

$$S_{ij} = \{x \in (0,\infty) : v_i(x) = (v_j - g_{ij})(x)\}$$

is the switching region from regime i to regime j. Moreover, from condition (2.12), when one switchs from regime i to regime j, one does not switch immediately to another regime, i.e. one stays for a while in the continuation region of regime j. In other words,

$$\mathcal{S}_{ij} \subset \mathcal{C}_j, \quad j \neq i \in \mathbb{I}_d.$$

The following useful lemma gives some partial information about the structure of the switching regions.

Lemma 4.1 For all $i \neq j$ in \mathbb{I}_d , we have

$$S_{ij} \subset Q_{ij} := \{x > 0 : (f_j - f_i)(x) - rg_{ij} \ge 0\}.$$

Proof. Let $x \in S_{ij}$. By setting $\varphi_j = v_j - g_{ij}$, it follows that x is a minimum of $v_i - \varphi_j$ with $v_i(x) = \varphi_j(x)$. Moreover, since x lies in the open set C_j where v_j is smooth, we have that φ_j is C^2 in a neighborhood of x. By the supersolution viscosity property of v_i to the PDE (3.2), this yields :

$$r\varphi_j(x) - \mathcal{L}\varphi_j(x) - f_i(x) \ge 0.$$
(4.1)

Now recall that for $x \in \mathcal{C}_j$, we have

$$rv_j(x) - \mathcal{L}v_j(x) - f_j(x) = 0;$$

so by substituting into (4.1), we obtain :

$$(f_j - f_i)(x) - rg_{ij} \geq 0,$$

which is the required result.

The economic interpretation of the set Q_{ij} is the following: if we are in Q_{ij} , then staying at regime *i* forever is worse than switching to regime *j* and staying there forever.

Economic assumptions on profit functions and switching costs

We now formalize the difference between the operational regimes. We consider the following ordering conditions on the regimes through their reward functions :

$$(\mathbf{Hf}) \qquad \qquad f_1 \prec f_2 \prec \ldots \prec f_d$$

 \iff

for all $i < j \in \mathbb{I}_d$, $f_j - f_i$ is strictly decreasing on $(0, \hat{x}_{ij})$ and strictly increasing on (\hat{x}_{ij}, ∞) for some $\hat{x}_{ij} \in \mathbb{R}_+$.

Economically speaking, the ordering condition $f_i \prec f_j$ means that the profit in regime j > i is "better" than profit in regime *i* from a certain level, and the improvement becomes

better and better, possibly with a saturation when $(f_j - f_i)(\infty) < \infty$. A typical example class of profit functions satisfying **(Hf)** is given by : $f_i(x) = k_i x^{\gamma_i}, k_i \ge 0, i \in \mathbb{I}_d, 0 < \gamma_1 < \ldots \gamma_d \le 1$.

In view of the ordering condition (**Hf**) on the regimes, it is natural to assume that the switching cost for an access to a higher regime is positive, i.e.

$$(\mathbf{Hg}+) \qquad \qquad g_{ij} > 0 \quad \text{for} \quad i < j \in \mathbb{I}_d$$

We shall also assume that one receives some compensation when one switches to a lower regime, i.e.

$$(\mathbf{Hg}-) \qquad \qquad g_{ij} < 0 \quad \text{for} \quad j < i \in \mathbb{I}_d$$

Notice that these conditions together with (2.12) imply the following ordering condition on the switching costs

$$0 < g_{ij} < g_{ik} \quad \text{for} \quad i < j < k \in \mathbb{I}_d \tag{4.2}$$

$$0 < -g_{ij} < -g_{ik} \quad \text{for} \quad k < j < i \in \mathbb{I}_d.$$

$$(4.3)$$

In other words, the higher regime one wants to reach, the more one has to pay; and the lower regime one reaches, the more one receives.

Given a regime $i \in \mathbb{I}_d$, the first question is to determine under which conditions the switching region S_i is nonempty. Specifically, for $x \in S_i$, one would like to know if one jumps to a higher regime, i.e. $x \in S_{ij}$ for some j > i, or if one switches to a lower regime, i.e. $x \in S_{ij}$ for j < i. So we introduce for any $i \in \mathbb{I}_d$ the upward and downward switching regions

$$\mathcal{S}_i^+ = \cup_{j>i} \mathcal{S}_{ij}, \quad \mathcal{S}_i^- = \cup_{j$$

with the convention that $S_i^+ = \emptyset$ for i = d and $S_i^- = \emptyset$ for i = 1. By definition, we have $S_i = S_i^+ \cup S_i^-$.

Remark 4.1 From the ordering condition **(Hf)**, we see that for all j > i, $Q_{ij} = [x_{ij}, \infty)$ for certain $x_{ij} > 0$. Symmetrically, we have that for all j < i, $Q_{ij} = (0, y_{ij}]$ for some $y_{ij} > 0$. Together with Lemma 4.1, we get :

$$(f_i - f_j)(x) + rg_{ij} \leq 0, \quad \forall j > i, \ \forall x_0 \in \mathcal{S}_{ij}, \ \forall x \ge x_0,$$

$$(4.4)$$

$$(f_i - f_j)(x) + rg_{ij} \leq 0, \quad \forall j < i, \ \forall y_0 \in \mathcal{S}_{ij}, \ \forall x \leq y_0.$$

$$(4.5)$$

We also have

$$\mathcal{S}_i^+ \subset \cup_{j>i} Q_{ij} = [x_i, \infty), \quad \mathcal{S}_i^- \subset \cup_{j$$

for some $x_i \in (0, \infty], y_i \in (0, \infty)$. In particular,

$$0 < \inf \mathcal{S}_i^+ \le \inf \mathcal{S}_{ij}, \quad \forall 1 \le i < j \le d, \tag{4.6}$$

$$\sup \mathcal{S}_{ij} \leq \sup \mathcal{S}_i^- < \infty, \qquad \forall 1 \leq j < i \leq d.$$

$$(4.7)$$

We give some properties on switching regions that do intersect. We introduce the following definitions.

Definition 4.1 Let $i, j \in \mathbb{I}_d$, $j \neq i$, and $x_0 \in S_{ij}$. We say that x_0 is a left -boundary (resp. right-boundary) of S_{ij} if there exists some $\varepsilon > 0$ s.t. $[x_0, x_0 + \varepsilon)$ (resp. $(x_0 - \varepsilon, x_0]) \subset S_{ij}$ and $(x_0 - \varepsilon, x_0) \cap S_{ij}$ (resp. $(x_0, x_0 + \varepsilon) \cap S_{ij}) = \emptyset$.

Definition 4.2 Let $i, j, k \in \mathbb{I}_d$, $j \neq i, k \neq i, j \neq k$, and $x_0 \in S_{ij} \cap S_{ik}$.

- x₀ is a crossing boundary point if it is a left-boundary of S_{ij} (or S_{ik}) and a rightboundary of S_{ik} (or S_{ij}).
- x_0 is a *j*-isolated point if there exists some $\varepsilon > 0$ s.t. $(x_0 \varepsilon, x_0]$ (resp. $[x_0, x_0 + \varepsilon)$) $\subset S_{ik}$, and $(x_0 - \varepsilon, x_0 + \varepsilon) \cap S_{ij} = \{x_0\}$.

Remark 4.2 Let $i, j \in \mathbb{I}_d, j \neq i$, with S_{ij} a singleton. Then, by Remark 3.3, S_{ij} is reduced to a single *j*-isolated point.

Lemma 4.2 For any $i, j, k \in \mathbb{I}_d$, $j \neq i, k \neq i, j \neq k$, we have $\operatorname{int}(S_{ij} \cap S_{ik}) = \emptyset$. Therefore, $S_{ij} \cap S_{ik}$ consists of only isolated or crossing boundaries points.

Proof. We argue by contradiction, and assume on the contrary that there exist some $i, j, k \in \mathbb{I}_d$, $j \neq i, k \neq i, j \neq k, x_0 \in S_{ij} \cap S_{ik}$, and $\varepsilon > 0$ such that $(x_0 - \varepsilon, x_0 + \varepsilon) \subset S_{ij} \cap S_{ik}$. Since S_{ij} is included in the open continuation region C_j on which v_j is C^2 , we deduce that v_i is C^2 on $(x_0 - \varepsilon, x_0 + \varepsilon)$. Recalling that $rv_j - \mathcal{L}v_j - f_j = 0$ on C_j , and since $v_i = v_j - g_{ij}$ on $(x_0 - \varepsilon, x_0 + \varepsilon) \subset C_j$, we deduce that

$$rv_i(x) - \mathcal{L}v_i(x) = r(v_j(x) - g_{ij}) - \mathcal{L}v_j(x)$$

= $f_j(x) - rg_{ij}, \quad \forall x \in (x_0 - \varepsilon, x_0 + \varepsilon).$

Similarly, we have

$$rv_i(x) - \mathcal{L}v_i(x) = f_k(x) - rg_{ik}, \quad \forall x \in (x_0 - \varepsilon, x_0 + \varepsilon).$$

By comparing the two previous equalities, we obtain

$$(f_j - f_k)(x) = r(g_{ij} - g_{ik}), \quad \forall x \in (x_0 - \varepsilon, x_0 + \varepsilon),$$

which is an obvious contradiction with (Hf).

Lemma 4.3 Let $x_0 \in S_{ij} \cap S_{ik}$, $i, j, k \in \mathbb{I}_d$, $j \neq i$, $k \neq i$, $j \neq k$.

- If x_0 is a *j*-isolated point, then $(f_k f_j)(x_0) \leq r(g_{ik} g_{ij})$.
- If x_0 is a crossing boundary point, then $(f_k f_j)(x_0) = r(g_{ik} g_{ij})$.

Proof. Suppose that x_0 is a *j*-isolated point, i.e. w.l.o.g. there exists some $\varepsilon > 0$ s.t. $(x_0 - \varepsilon, x_0]$ (resp. $[x_0, x_0 + \varepsilon)$) $\subset S_{ik}$, and $(x_0 - \varepsilon, x_0 + \varepsilon) \cap S_{ij} = \{x_0\}$. We set $G = (v_j - g_{ij}) - (v_k - g_{ik})$. Since S_{ij} (resp. S_{ik}) is included in the open continuation set C_j (resp. C_k) where v_j (resp. v_k) satisfies $rv_j - \mathcal{L}v_j - f_j = 0$ (resp. $rv_k - \mathcal{L}v_k - f_k = 0$), we deduce that G is C^2 in a neighborhood of x_0 and satisfies :

$$rG - \mathcal{L}G = f_j - f_k - r(g_{ij} - g_{ik})$$
 on $(x_0 - \delta, x_0 + \delta)$ (4.8)

for some $0 < \delta < \varepsilon$. By definition of the switching regions, we have $v_i = v_k - g_{ik} > v_j - g_{ij}$ on $(x_0, x_0 + \varepsilon)$ with equality at x_0 . By the smooth-fit property of v_i at x_0 , this implies : $G(x_0) = G'(x_0) = 0$, and $G''(x_0) \le 0$. By sending x to x_0 into (4.8), it follows that $(f_j - f_k)(x_0) - r(g_{ij} - g_{ik}) \ge 0$.

Suppose now that x_0 is a crossing boundary point, and so w.l.o.g. there exists $\varepsilon > 0$ s.t. $(x_0 - \varepsilon, x_0] \subset S_{ij}, [x_0, x_0 + \varepsilon) \subset S_{ik}$, with $(x_0 - \varepsilon, x_0) \cap S_{ik} = (x_0, x_0 + \varepsilon) \cap S_{ij} = \emptyset$. By setting $G = (v_j - g_{ij}) - (v_k - g_{ik})$, and by the same arguments as above, we have (4.8). Moreover, we have $v_i = v_j - g_{ij} > v_k - g_{ik}$ on $(x_0 - \varepsilon, x_0)$, and $v_i = v_k - g_{ik} > v_j - g_{ij}$ on $(x_0, x_0 + \varepsilon)$, with equality at x_0 . By the smooth-fit property of v_i at x_0 , this implies : $G(x_0) = G'(x_0) = 0, G''(x_0) \ge 0, G''(x_0) \le 0$, and so $G''(x_0) = 0$. Similarly, by observing that $v_i = v_k - g_{ik} > v_i - g_{ij}$ on $(x_0, x_0 + \varepsilon)$, with equality at x_0 , we have $G''(x_0^+) \le 0$. By sending x to x_0 in (4.8), we obtain the equality $(f_j - f_k)(x_0) - r(g_{ij} - g_{ik}) = 0$.

We now give some properties on switching regions that are separated by a continuation region.

Lemma 4.4 Let $i \in \mathbb{I}_d$, and suppose (x_0, y_0) is a nonempty open bounded interval included in the continuation region C_i with $x_0 \in S_{ij}$ and $y_0 \in S_{ik}$ for some $j \neq i$ and $k \neq i$. 1) If j > i, then k > i and $(f_k - f_i)(x_0) < rg_{ik}$. In particular, $x_0 < \inf S_{ik}$ and $k \neq j$. 2) If j < i and k < i, then $(f_i - f_j)(y_0) > -rg_{ij}$. In particular, $y_0 > \sup S_{ij}$ and $k \neq j$.

Proof. We consider the (nonnegative) continuous function : $G = v_i - \max(v_j - g_{ij}, v_k - g_{ik})$. Since $rv_i - \mathcal{L}v_i - f_i = 0$ on \mathcal{C}_i and hence on $(x_0, y_0) \subset \mathcal{C}_i$, and $rv_j - \mathcal{L}v_j - f_j \ge 0$, $rv_k - \mathcal{L}v_k - f_k \ge 0$ on $(0, \infty)$ (in the viscosity sense), we easily check that G satisfies in the viscosity sense:

$$rG - \mathcal{L}G \leq \max(f_i - f_j + rg_{ij}, f_i - f_k + rg_{ik}) \quad \text{on} \quad (x_0, y_0).$$
 (4.9)

1) Suppose that j > i. Since $x_0 \in S_{ij}$, we have by (4.4) :

$$(f_i - f_j)(x) + rg_{ij} \leq 0, \quad \forall x \geq x_0.$$

We first show that k > i. If not, i.e. k < i, and since $y_0 \in S_{ik}$, we have by (4.5): $f_i - f_k + rg_{ik} \le 0$ on $(0, y_{ik})$ and in particular on (x_0, y_0) . Therefore, we get from (4.9):

$$rG - \mathcal{L}G \leq 0 \quad \text{on} \quad (x_0, y_0). \tag{4.10}$$

Now, since $x_0 \in S_{ij}$, $y_0 \in S_{ik}$, we have $G(x_0) = G(y_0) = 0$. By the comparison principle for viscosity solutions, G should non-positive on (x_0, y_0) . This is in contradiction with the fact that $(x_0, y_0) \in C_i$ on which G is strictly positive. Hence, k > i. We now prove that $(f_k - f_i)(x_0) < rg_{ik}$, i.e. $x_0 \notin Q_{ik}$. If it is not true, and since Q_{ik} is an interval in the form $[x_{ik}, \infty)$, we would have $(f_k - f_i)(x) - rg_{ik} \ge 0$ for all $x \ge x_0$, and hence (4.10). As above, this provides the required contradiction. Finally, since $S_{ik} \subset Q_{ik}$, this proves that $x_0 < \inf S_{ik}$, and in particular $k \ne j$.

2) Suppose that j < i and k < i. Recalling that $y_0 \in S_{ik}$, we have by (4.5) :

$$(f_i - f_k)(x) + rg_{ik} \leq 0, \quad \forall x \leq y_0$$

We now prove that $(f_i - f_j)(y_0) > -rg_{ij}$, i.e. $y_0 \notin Q_{ij}$. If not, and since Q_{ij} is an interval in the form $(0, y_{ij}]$, we would have $(f_i - f_j)(y) \leq -rg_{ij}$, for all $x \leq y_0$, and so holds (4.10). By the same arguments as in 1), we get the required contradiction.

4.1 Analysis of upward switching region

The main results of this section provide a qualitative description of the upward switching regions.

Proposition 4.1 Let $i \in \mathbb{I}_d$.

1) The switching region S_i^+ is nonempty if and only if :

$$\bigcup_{j>i} Q_{ij} \neq \emptyset \iff \exists j>i, \ (f_j - f_i)(\infty) > rg_{ij}.$$

$$(4.11)$$

2) Suppose $S_i^+ \neq \emptyset$. Then there exists a unique $j = j^+(i) > i$ such that $\sup S_i^+ = \sup S_{ij}$ = ∞ , and we have $\sup S_{ik} < \infty$ for all k > i, $k \neq j^+(i)$. Moreover, $S_{ij^+(i)}$ contains an interval in the form $[x_{ij^+(i)}, \infty)$ for some $x_{ij^+(i)} \in (0, \infty)$, and $j^+(i) = \min J(i)$ where

$$J(i) = \left\{ j \in \mathbb{I}_d, \ j > i \ : (f_k - f_j)(\infty) \le r(g_{ik} - g_{ij}), \ \forall k \in \mathbb{I}_d, \ k > i \right\}.$$
(4.12)

Economic interpretation. The first assertion gives explicit necessary and sufficient conditions under which, in a given regime, it is optimal to switch up. It means that one has interest to switch up if and only if one may find some higher regime so that the maximal net difference between the profit functions covers strictly the cost for changing regime. The interpretation of the second assertion is the following. In a given regime, say i, where one has interest to switch up (under the conditions of assertion 1), there is a unique regime where one should switch up to when the state is sufficiently large. Moreover, this uniquely chosen regime is explicitly determined as the minimum of the explicitly given set J(i). In the two-regime case, i.e. d = 2, we obviously have $j^+(1) = 2$. In the multi-regime case, it can be practically calculated as follows :

• one first tests if $j_1 = i + 1$ lies in J(i) i.e. if

 (P_{j_1}) $(f_k - f_{j_1})(\infty) \le r(g_{ik} - g_{ij_1}), \quad \forall k > j_1$ is satisfied. If yes, then $j^+(i) = j_1$.

• Otherwise, we denote

$$j_2 = \min \left\{ k > j_1 : (f_k - f_{j_1})(\infty) > r(g_{ik} - g_{ij_1}) \right\},$$

and we notice that $j \notin J(i)$ for $j_1 < j < j_2$. Indeed, by definition of j_2 , if $j < j_2$, then $(f_j - f_{j_1})(\infty) \le r(g_{ij} - g_{ij_1})$ and so $(f_{j_2} - f_j)(\infty) = (f_{j_2} - f_{j_1})(\infty) + (f_{j_1} - f_j)(\infty) > r(g_{ij_2} - g_{ij_1}) + r(g_{ij_1} - g_{ij_2}) = r(g_{ij_2} - g_{ij_1})$, or $j \notin J(i)$. Then, one tests if j_2 lies in J(i). By observing that for $k < j_2$, we have $(f_k - f_{j_2})(\infty) = (f_k - f_{j_1})(\infty) + (f_{j_1} - f_{j_2})(\infty) < r(g_{ik} - g_{ij_1}) + r(g_{ij_1} - g_{ij_2}) = r(g_{ik} - g_{ij_2})$, the test $j_2 \in J(i)$ is written as :

 (P_{j_2}) $(f_k - f_{j_1})(\infty) \le r(g_{ik} - g_{ij_1}), \quad \forall k > j_2.$

If (P_{j_2}) is satisfied, then $j^+(i) = j_2$.

• We continue by forward induction : if (P_{j_l}) is satisfied, then $j^+(i) = j_l$. Otherwise, we denote

$$j_{l+1} = \min \left\{ k > j_l : (f_k - f_{j_l})(\infty) > r(g_{ik} - g_{ij_l}) \right\},$$

and one tests if $j_{l+1} \in J(i)$, i.e., if

 $(P_{j_{l+1}}) \quad (f_k - f_{j_l})(\infty) \le r(g_{ik} - g_{ij_l}), \quad \forall k > j_l$

is satisfied. The property (P_{j_k}) may be never satisfied in which case $j^+(i) = d$. For example, for the reward profit functions $f_i(x) = k_i x^{\gamma_i}$ with $k_i > 0$ and $0 < \gamma_1 < \ldots \gamma_d \leq 1$, we have $(f_k - f_j)(\infty) = \infty$ for all k > j; and so $j^+(i) = d$.

Proof of Proposition 4.1

1) Notice first that the equivalence in (4.11) follows immediately from the definition of Q_{ij} and the ordering condition (**Hf**). The necessary condition $S_i^+ \neq \emptyset \implies \bigcup_{j>i} Q_{ij} \neq \emptyset$ is a direct consequence of the inclusion in Lemma 4.1. We now prove the following implication :

$$\sup \mathcal{S}_i^+ < \infty \implies (f_j - f_i)(\infty) \le rg_{ij}, \quad \forall j \in \mathbb{I}_d.$$
(4.13)

Indeed, assume that $\sup \mathcal{S}_i^+ < \infty$. Then from (4.7), the switching region \mathcal{S}_i is bounded. This means that the continuation region \mathcal{C}_i contains (y_i, ∞) for some $y_i > 0$, and so $rv_i(x) - \mathcal{L}v_i(x) - f_i(x) = 0$ for $x > y_i$. Then, on (y_i, ∞) , v_i is in the form : $v_i(x) = \hat{V}_i(x) + A\psi^-(x) + B\psi^+(x)$, for some constants A and B. Moreover, from the linear growth condition on v_i and condition (3.8), we must have B = 0. Hence, recalling that $v_i \ge v_j - g_{ij}$ and $v_j \ge \hat{V}_j$ for all j, we have

$$\hat{V}_i(x) + A\psi^-(x) \ge \hat{V}_j(x) - g_{ij}, \quad \forall x \ge y_i.$$

By sending x to infinity and from (2.9), (3.17), we obtain (4.13). This shows in particular that if $S_i^+ = \emptyset$, and so by convention $\sup S_i^+ = 0 < \infty$, then $\bigcup_{j>i} Q_{ij} = \emptyset$. The proof of the equivalence in the first assertion is thus completed.

2) Now, suppose that $S_i^+ \neq \emptyset$, which means equivalently that (4.11) is satisfied. Hence, from (4.13), we must have $\sup S_i^+ = \infty$. Let us then consider the nonempty set $J_{\infty}(i) = \{j > i : \sup S_{ij} = \infty\}$.

Step 1. Fix some $j \in J_{\infty}(i)$ and let us prove that S_{ij} contains an interval in the form $[x_{ij}, \infty)$ for some $x_{ij} > 0$. Since $\sup S_{ij} = \infty$, one can find an increasing sequence $(x_n)_n$ in S_{ij} such that $\lim_n x_n = \infty$. We distinguish the following cases:

- Case 1: $\sup C_i = \infty$. Since the continuation region C_i is open, there exists for all n, a nonempty interval $(\tilde{x}_n, \tilde{y}_n)$ included in C_i , with $x_n \leq \tilde{x}_n \in \mathcal{S}_{ij'}, \tilde{y}_n \in \mathcal{S}_{ik}$ for some $j', k \neq i$. By (4.7) and since $\lim_n x_n = \infty$, we may assume by taking n large enough that j' > i and $\tilde{x}_n \geq \inf \mathcal{S}_{ik}$. This, however, contradicts Lemma 4.4 1), and so *Case 1* would never occur. - *Case 2*: $\sup C_i < \infty$. By (4.7) and since $\lim_n x_n = \infty$, it follows that for n large enough, $[x_n, \infty) \subset \mathcal{S}_i^+ = \bigcup_{k > i} \mathcal{S}_{ik}$. We have two possible subcases:

- a) There exists some n such that $[x_n, \infty) \subset S_{ij}$, and we are done.
- b) Otherwise, by Lemma 4.2, there exist some k ≠ j', k, j' > i, and an increasing sequence (y_n^m)_m with x_n ≤ y_n⁰, lim_m y_n^m = ∞ such that S_{ij'} contains ∪_m[y_n^{2m}, y_n^{2m+1}] and S_{ik} contains ∪_m[y_n^{2φ(m)+1}, y_n^{2φ(m)+2}], where φ is an increasing function valued in N. Recalling that S_{ij'} ⊂ C_{j'}, we have ∪_{m≥0}[y_n^{2m}, y_n^{2m+1}] ⊂ C_{j'}. We claim that C_{j'} contains actually [y_n^m, ∞) for some n and m large enough. Otherwise, for all n, m, one could find some nonempty interval (z_n^m, ž_n^m) included in C_{j'}, with y_n^m ≤ z_n^m < ž_n^m and z_n^m ∈ S_{j'k'}, ž_n^m ∈ S_{j'l} for some k', l ≠ j'. Since lim_n y_n^m = ∞, we may assume, by taking n large enough, that z_n^m ≥ inf S_{j'l}. This contradicts again Lemma 4.4 1), and hence [y_n^m, ∞) ⊂ C_{j'} for some m, n. Fix such an n. Then, recalling that S_{ik} ⊂ C_k and lim_m y_n^m = ∞, we show similarly that [y_n^{m'}, ∞) ⊂ C_k for some m' large enough. By setting m₀ = m ∨ m', we then deduce that [y_n^{m0}, ∞) is included both in C_{j'} and C_k. Therefore, on [y_n^{m0}, ∞), v_{j'} and v_k satisfy respectively rv_{j'} − Lv_{j'} − f_j = 0 and rv_k − Lv_k − f_k = 0. Since v_{j'}, v_k satisfy also the linear growth condition at infinity, they are in the form v_{j'}(x) = V̂_{j'}(x) + A_{j'}ψ⁻(x), v_k(x) = V̂_k(x) + A_kψ⁻(x), x ∈ [y_n^{m0}, ∞), for some constants A_{j'}, A_k. Now, by writing that y_n^{2φ(m)+1} ∈ S_{ij'} ∩ S_{ik}, and by the very definition of the switching regions, we have for all m ≥ m₀:

$$v_{i}(y_{n}^{2\phi(m)+1}) = v_{j'}(y_{n}^{2\phi(m)+1}) - g_{ij'}$$

= $\hat{V}_{j'}(y_{n}^{2\phi(m)+1}) + A_{j'}\psi^{-}(y_{n}^{2\phi(m)+1}) - g_{ij'}$
= $v_{k}(y_{n}^{2\phi(m)+1}) - g_{ik}$
(4.14)

$$= \hat{V}_k(y_n^{2\phi(m)+1}) + A_k\psi^-(y_n^{2\phi(m)+1}) - g_{ik}.$$
(4.15)

By sending n to infinity into the r.h.s. of (4.14)-(4.15), and from (2.9), (3.17), we obtain :

$$(f_k - f_{j'})(\infty) = r(g_{ik} - g_{ij'}).$$

Since $k \neq j'$, say, k > j', it follows from **(Hf)** that $(f_k - f_{j'})(x) < r(g_{ik} - g_{ij'})$ for all x > 0. Consequently, in view of (3.15) we have

$$\hat{V}_k(x) - \hat{V}_j(x) < g_{ik} - g_{ij'}, \quad \forall x > 0.$$
 (4.16)

Hence, for all x large enough, we have from (2.9)

$$\hat{V}_k(x) + A_k \psi^-(x) - g_{ik} < \hat{V}_{j'}(x) + A_{j'} \psi^-(x) - g_{ij'}.$$

This contradicts (4.14)-(4.15) for $x = y_n^{2\phi(m)+1}$, and so *Case 2 b*) never occurs.

Step 2. Fix some $j \in J_{\infty}(i)$. By the definition of the switching region and by Step 1, we have $v_i = v_j - g_{ij}$ on $[x_{ij}, \infty)$. Moreover, since S_{ij} is included in the continuation region C_j , we deduce that $rv_j - \mathcal{L}v_j - f_j = 0$ on $[x_{ij}, \infty)$; and so by the linear growth condition of v_j at infinity, we have $v_j(x) = \hat{V}_j(x) + A_j\psi^-(x)$, $x \ge x_{ij}$, for some constant A_j . Therefore,

$$\psi_i(x) = \hat{V}_j(x) + A_j \psi^-(x) - g_{ij}, \quad \forall x \ge x_{ij}.$$

Writing $v_i \ge v_k - g_{ik} \ge \hat{V}_k - g_{ik}$ for all k, we get :

$$\hat{V}_j(x) + A\psi^-(x) - g_{ij} \geq \hat{V}_k(x) - g_{ij}, \quad \forall x \geq x_{ij}, \ \forall k \in \mathbb{I}_d.$$

By sending x to infinity and from (2.9), (3.17), we obtain

$$(f_j - f_k)(\infty) \geq r(g_{ij} - g_{ik}), \quad \forall k \in \mathbb{I}_d$$

This proves that j lies in J(i) defined in (4.12), and so $J_{\infty}(i) \subset J(i)$.

Step 3. We prove that $J_{\infty}(i)$ is reduced to a singleton. For this, consider $j, j' \in J_{\infty}(i)$. From Step 2, there exists some $x'_i > 0$ such that

$$v_i(x) = \hat{V}_j(x) + A_j \psi^-(x) - g_{ij} = \hat{V}_{j'}(x) + A_{j'} \psi^-(x) - g_{ij'}, \quad \forall x \ge x'_i,$$

for some constants A_j , $A_{j'}$. Moreover, by writing that j and j' lie in J(i), we have :

$$(f_{j'} - f_j)(\infty) = r(g_{ij'} - g_{ij}).$$

Thus, if we assume that $j' \neq j$, we obtain a contradiction by the same argument as that in the end of Step 1.

Step 4. We finally prove that the singleton $J_{\infty}(i)$ consists of $j^+(i) := \min J(i)$. Let j be the unique element in $J_{\infty}(i)$. Then, we recall from Step 1 that $v_i = \hat{V}_j + A_j\psi^- - g_{ij}$ on $[x_{ij}, \infty)$ for some $x_{ij} > 0$ and constant A_j . Since $j, j_+(i) \in J(i)$, we have :

$$(f_j - f_{j^+(i)})(\infty) = r(g_{ij} - g_{ij^+(i)}),$$

and obviously $j \ge j^+(i)$. Assume on the contrary that $j \ne j^+(i)$. Then $j > j^+(i)$, and by the same arguments as in (4.16), we obtain :

$$\hat{V}_j(x) - \hat{V}_{j^+(i)}(x) < g_{ij} - g_{ij^+(i)}, \quad \forall x > 0.$$

By (2.9), this implies for x large enough :

$$v_i(x) = \hat{V}_j + A_j \psi^-(x) - g_{ij} < \hat{V}_{j^+(i)}(x) - g_{ij^+(i)} \le v_{j^+(i)} - g_{ij^+(i)} \le v_i(x),$$

a contradiction.

Proposition 4.2 Let $i \in \{1, \ldots, d-1\}$ with $S_i^+ \neq \emptyset$. 1) Suppose that

$$\sup \left[\mathcal{S}_{ik} \setminus \mathcal{S}_{ij^+(i)} \right] \leq \inf \mathcal{S}_{ij^+(i)}, \quad \forall k \neq i, j^+(i).$$
(4.17)

Then, we have

$$\mathcal{S}_{ij^+(i)} = [\bar{x}_{ij^+(i)}, \infty),$$

with $\bar{x}_{ij^+(i)} \in (0,\infty)$.

2) Suppose that there exists $k > i, k \neq j^+(i)$ such that S_{ik} is nonempty and

$$\sup \mathcal{S}_{ik} \leq \inf \mathcal{S}_{ij}, \quad \forall j \neq i, k.$$

$$(4.18)$$

Then, S_{ik} is in the form

$$\mathcal{S}_{ik} = [\bar{x}_{ik}, \bar{y}_{ik}],$$

with $0 < \bar{x}_{ik} \leq \bar{y}_{ik} < \infty$.

Proof. 1). We set $\bar{x}_{ij^+(i)} = \inf \mathcal{S}_{ij^+(i)}$, which is finite since $\mathcal{S}_{ij^+(i)}$ is nonempty (see Proposition 4.1). By (4.6), we also notice that $\bar{x}_{ij^+(i)} > 0$. Suppose that (4.17) holds. Then, $(\bar{x}_{ij^+(i)}, \infty) \subset \mathcal{S}_{ij^+(i)} \cup \mathcal{C}_i$. From (3.3), we then deduce that v_i is a viscosity solution to

$$\min\left[rv_i - \mathcal{L}v_i - f_i, \ v_i - (v_{j^+(i)} - g_{ij^+(i)})\right] = 0 \quad \text{on} \quad (\bar{x}_{ij^+(i)}, \infty).$$
(4.19)

Let us now consider the continuous function $w_i = v_{j^+(i)} - g_{ij^+(i)}$ on $[\bar{x}_{ij^+(i)}, \infty)$. We check that w_i is a viscosity supersolution to

$$rw_i - \mathcal{L}w_i - f_i \geq 0 \quad \text{on} \quad (\bar{x}_{ij^+(i)}, \infty).$$
 (4.20)

For this, take some point $\bar{x} \in (\bar{x}_{ij^+(i)}, \infty)$ and some smooth test function φ s.t. \bar{x} is a local minimum of $w_i - \varphi$. Then, \bar{x} is a local minimum of $v_{j^+(i)} - (\varphi + g_{ij^+(i)})$. By writing the viscosity supersolution property of $v_{j^+(i)}$ to its Bellman PDE, we have :

$$rv_{j^+(i)}(\bar{x}) - \mathcal{L}\varphi(\bar{x}) - f_{j^+(i)}(\bar{x}) \ge 0.$$

By applying inequality (4.4) to $\bar{x} > \bar{x}_{ij^+(i)} \in \mathcal{S}_{ij^+(i)}$, we have :

$$(f_{j+(i)} - f_i)(\bar{x}) - rg_{ij+(i)} \geq 0.$$

By adding these two last inequalities, we obtain the required supersolution inequality :

$$rw_i(\bar{x}) - \mathcal{L}\varphi(\bar{x}) - f_i(\bar{x}) \ge 0,$$

and so (4.20) is proved. Since $w_i = v_{j^+(i)} - g_{ij^+(i)}$, this proves that w_i is a viscosity solution to :

$$\min\left[rw_i - \mathcal{L}w_i - f_i, \ w_i - (v_{j+(i)} - g_{ij+(i)})\right] = 0 \quad \text{on} \quad (\bar{x}_{ij+(i)}, \infty).$$
(4.21)

Moreover, since $\bar{x}_{ij^+(i)} \in S_{ij^+(i)}$, we have $w_i(\bar{x}_{ij^+(i)}) = v_i(\bar{x}_{ij^+(i)})$. Observing also that v_i and w_i satisfy the linear growth condition at infinity, we deduce by uniqueness of viscosity solution to (4.19) that $v_i = w_i$ on $(\bar{x}_{ij^+(i)}, \infty)$, i.e. $S_{ij^+(i)} = [\bar{x}_{ij^+(i)}, \infty)$.

2). We set $\bar{x}_{ik} = \inf S_{ik}$, $\bar{y}_{ik} = \sup S_{ik}$. By Proposition 4.1 2), we know that $\bar{y}_{ik} < \infty$. From (4.6), we notice also that $\bar{x}_{ik} > 0$. Suppose that S_{ik} is neither empty nor a singleton, so that $0 < \bar{x}_{ik} < \bar{y}_{ik} < \infty$, and suppose that (4.18) holds. Let us prove that $S_{ik} = [\bar{x}_{ik}, \bar{y}_{ik}]$. For this, we consider the function $w_i = v_k - g_{ik}$ on $[\bar{x}_{ik}, \bar{y}_{ik}]$. The condition (4.18) implies that

$$v_i(x) > v_j(x) - g_{ij}, \quad \forall j \neq i, k, \quad \forall x < \bar{y}_{ik}.$$

From (3.3), we deduce that v_i is a viscosity solution to

$$\min[rv_i - \mathcal{L}v_i - f_i, v_i - (v_k - g_{ik})] = 0 \quad \text{on} \quad (\bar{x}_{ik}, \bar{y}_{ik}).$$
(4.22)

By same arguments as in (4.20)-(4.21), we show that w_i is a viscosity solution to

$$\min\left[rw_i - \mathcal{L}w_i - f_i, w_i - (v_k - g_{ik})\right] = 0 \quad \text{on} \quad (\bar{x}_{ik}, \bar{y}_{ik}).$$

Moreover, since \bar{x}_{ik} and $\bar{y}_{ik} \in S_{ik}$, we have $w_i(\bar{x}_{ik}) = v_i(\bar{x}_{ik})$ and $w_i(\bar{y}_{ik}) = v_i(\bar{y}_{ik})$. By uniqueness of viscosity solution to (4.22), we deduce that $v_i = w_i$ on $[\bar{x}_{ik}, \bar{y}_{ik}]$, i.e. $S_{ik} = [\bar{x}_{ik}, \bar{y}_{ik}]$.

Remark 4.3 In the two-regime case, i.e d = 2, assertion 2) of the preceding proposition is not applicable while assertion 1) means that $S_1 = S_{12}$ is either empty (see condition 1) in Proposition 4.1) or it is in the form $[\bar{x}_{12}, \infty)$ for some $\bar{x}_{12} > 0$. Proposition 4.2 extends this result (already found in [19]) to the multi-regime case, and we shall see in particular in the next section how one can check conditions (4.17) and (4.18) in the three-regime case in order to determine the upward switching regions.

4.2 Analysis of downward switching region

The main results of this paragraph provide a qualitative description of the downward switching regions.

Proposition 4.3 For all i = 2, ..., d, the switching region S_i^- is nonempty. Moreover, inf $S_{i1} = 0$, S_{i1} contains some interval in the form $(0, y_{i1}]$, $y_{i1} > 0$, and $\inf S_{ij} > 0$ for all 1 < j < i.

Economic interpretation. The first assertion means that one always has interest to switch down due to the negative switching costs. Moreover, for small values of the state, one should switch down to the lowest regime i = 1. This is intuitively justified by the fact that for small values of the state, the running profits are close to zero, and so one chooses the regime with the largest compensation fee, i.e. regime 1, see (4.3).

Proof of Proposition 4.3

1. We first prove that $\inf \mathcal{S}_i^- = 0$. If not, since $\inf \mathcal{S}_i^+ > 0$ by (4.6), we then have $\inf \mathcal{S}_i > 0$. Therefore, the continuation region \mathcal{C}_i contains $(0, y_i)$ for some $y_i > 0$, and so $rv_i - \mathcal{L}v_i - f_i = 0$ on $(0, y_i)$. Then, on $(0, y_i)$, v_i is in the form $v_i(x) = \hat{V}_i(x) + A\psi^-(x) + B\psi^+(x)$, for some constants A and B. Recalling that $v_i(0^+)$ is finite, and under (2.9), we have A = 0. By writing that $v_i \ge v_j - g_{ij} \ge \hat{V}_j - g_{ij}$ for all j, we have

$$\hat{V}_i(x) + B\psi^+(x) \geq \hat{V}_j(x) - g_{ij}, \quad \forall x \in (0, y_i).$$

Sending x to zero and from (2.9), we obtain $0 \ge -g_{ij}$. This is a contradiction with (Hg-) for j < i. Therefore, $\inf S_i^- = 0$, and in particular $S_i^- \neq \emptyset$.

2. Let us then consider the nonempty set $J_0(i) = \{j < i : \inf S_{ij} = 0\}$. Take some $j \in J_0(i)$. Then, one can find a decreasing sequence $(x_n)_n$ in S_{ij} such that $\lim_n x_n = 0$. Since S_{ij} is closed, this implies that for n large enough, S_{ij} contains the interval $(0, x_n]$. Then, $v_i = v_j - g_{ij}$ on $(0, x_n]$. Moreover, recalling that S_{ij} is included on C_j , we deduce that $rv_j - \mathcal{L}v_j - f_j = 0$ on $(0, x_n)$ and so $v_i = v_j - g_{ij} = \hat{V}_j + B\psi^+ - g_{ij}$ on $(0, x_n)$. By writing that $v_i \ge v_k - g_{ik} \ge \hat{V}_k - g_{ik}$ for all k, we obtain :

$$\hat{V}_j(x) + B\psi^+(x) - g_{ij} \ge \hat{V}_k(x) - g_{ik}, \quad \forall x \in (0, x_n).$$

By sending x to zero, we conclude $-g_{ij} \ge -g_{ik}$ for all k < i. Under (4.3), this means that j = 1, and the proof is completed.

Proposition 4.4 Let $i \in \{2, \ldots, d\}$.

1) Suppose that

$$\sup \mathcal{S}_{i1} \leq \inf \left[\mathcal{S}_{ij} \setminus \mathcal{S}_{i1} \right], \quad \forall j \neq 1, i.$$
(4.23)

Then, we have

$$\mathcal{S}_{i1} = (0, y_{i1}],$$

with $\underline{y}_{i1} \in (0, \infty)$.

2) Suppose there exists $k < i, k \neq 1$ such that S_{ik} is nonempty and

$$\sup \mathcal{S}_{ij} \leq \inf \mathcal{S}_{ik}, \quad \forall j \neq i, k.$$

$$(4.24)$$

Then, S_{ik} is in the form

$$S_{ik} = [\underline{x}_{ik}, \underline{y}_{ik}]_{ik}$$

with $0 < \underline{x}_{ik} \leq \underline{y}_{ik} < \infty$.

Proof. 1). We set $\underline{y}_{i1} = \sup S_{i1}$, which is positive since S_{i1} is nonempty (see Proposition 4.3). By (4.7), we also notice that $\underline{y}_{i1} < \infty$. Suppose that (4.23) holds. Then, $(0, \underline{y}_{i1}) \subset S_{i1} \cup C_i$. From (3.3), we deduce that v_i is a viscosity solution to

$$\min[rv_i - \mathcal{L}v_i - f_i, v_i - (v_1 - g_{i1})] = 0 \quad \text{on} \quad (0, \underline{y}_{i1}).$$
(4.25)

Let us prove that $S_{i1} = (0, \underline{y}_{i1}]$. To this end, we consider the function $w_i = v_1 - g_{i1}$ on $(0, \underline{y}_{i1}]$. By same arguments as in (4.20)-(4.21), we show that w_i is a viscosity solution to

$$\min [rw_i - \mathcal{L}w_i - f_i, w_i - (v_1 - g_{i1})] = 0 \quad \text{on} \quad (0, \underline{y}_{i1}).$$

Moreover, since $\inf \mathcal{S}_{i1} = 0$ and $\underline{y}_{i1} \in \mathcal{S}_{i1}$, we have $w_i(0^+) = v_i(0^+) (= -g_{i1})$ and $w_i(\underline{y}_{i1}) = v_i(\underline{y}_{i1})$. By uniqueness of viscosity solution to (4.25), we deduce that $v_i = w_i$ on $(0, \underline{y}_{i1}]$, i.e. $\mathcal{S}_{i1} = (0, \underline{y}_{i1}]$.

2). We set $\underline{x}_{ik} = \inf S_{ik}, \underline{y}_{ik} = \sup S_{ik}$. By Proposition 4.3 2), we know that $\underline{x}_{ik} > 0$. From (4.7), we notice also that $\underline{y}_{ik} < \infty$. Suppose that S_{ik} is neither empty nor a singleton, so that $0 < \underline{x}_{ik} < \underline{y}_{ik} < \infty$, and suppose that (4.24) holds. Let us prove that $S_{ik} = [\underline{x}_{ik}, \underline{y}_{ik}]$. For this, we consider the function $w_i = v_k - g_{ik}$ on $[\underline{x}_{ik}, \underline{y}_{ik}]$. The condition (4.24) implies that

$$v_i(x) > v_j(x) - g_{ij}, \quad \forall j \neq i, k, \quad \forall x > \underline{x}_{ik}.$$

From (3.3), we deduce that v_i is a viscosity solution to

$$\min\left[rv_i - \mathcal{L}v_i - f_i, v_i - (v_k - g_{ik})\right] = 0 \quad \text{on} \quad (\underline{x}_{ik}, \underline{y}_{ik}).$$
(4.26)

By same arguments as in (4.20)-(4.21), we show that w_i is a viscosity solution to

$$\min \left[rw_i - \mathcal{L}w_i - f_i, w_i - (v_k - g_{ik}) \right] = 0 \quad \text{on} \quad (\underline{x}_{ik}, \underline{y}_{ik})$$

Moreover, since \underline{x}_{ik} and $\underline{y}_{ik} \in S_{ik}$, we have $w_i(\underline{x}_{ik}) = v_i(\underline{x}_{ik})$ and $w_i(\underline{y}_{ik}) = v_i(\underline{y}_{ik})$. By uniqueness of viscosity solution to (4.26), we deduce that $v_i = w_i$ on $[\underline{x}_{ik}, \underline{y}_{ik}]$, i.e. $S_{ik} = [\underline{x}_{ik}, \underline{y}_{ik}]$.

Remark 4.4 In the two-regime case, i.e, d = 2, assertion 2) of the preceding proposition does not apply while assertion 1) means that $S_2 = S_{21}$ is in the form $(0, \underline{y}_{21}]$ for some $\underline{y}_{21} > 0$. Proposition 4.4 extends this result (already found in [19]) to the multi-regime case, and we shall see in particular in the next section how one can check conditions (4.23) and (4.24) in the three-regime case in order to determine the downward switching regions.

5 The three-regime case

In this section, we consider the case of three regimes, i.e. d = 3, and we show how one can use the results of the previous sections to obtain a fairly explicit description of the the switching regions.

We start with the lowest regime 1. Notice that with regime 1, we have $S_1 = S_1^+ = S_{12} \cup S_{13}$.

Theorem 5.1 (Switching regions in Regime 1)

1) Suppose that $(f_2 - f_1)(\infty) \leq rg_{12}$ and $(f_3 - f_1)(\infty) \leq rg_{13}$. Then $S_1^+ = \emptyset$.

2) Suppose that $(f_2 - f_1)(\infty) > rg_{12}$ or $(f_3 - f_1)(\infty) > rg_{13}$.

a) If $(f_3 - f_2)(\infty) \leq r(g_{13} - g_{12})$, then $S_{12} = [\bar{x}_{12}, \infty)$ for some $\bar{x}_{12} > 0$, and $S_{13} = \emptyset$,

b) If $(f_3 - f_2)(\infty) > r(g_{13} - g_{12})$, then $S_{13} = [\bar{x}_{13}, \infty)$ for some $\bar{x}_{13} > 0$. Moreover, S_{12} is either empty or in the form $S_{12} = [\bar{x}_{12}, \bar{y}_{12}]$ for some $0 < \bar{x}_{12} \le \bar{y}_{12} \le \bar{x}_{13}$.

Remark 5.1 The above qualitative description of the switching regions is explicit, depending only on the model parameters, in case 1) and 2a). For example, case 2a) means that when the maximal difference between profit functions in regime 3 and regime 2 is smaller than the difference between the corresponding switching costs, where it is never optimal to switch to regime 2, while it is optimal to wait until the state value reaches some threshold when one has interest to switch to regime 3. In case 2b), the description is fairly explicit, although we are not able to give an explicit characterization when S_{12} is empty. We have an obvious necessary condition $(f_2 - f_1)(\infty) \leq rg_{12}$, which implies under **(Hf)** that $Q_{12} =$ \emptyset and so by Lemma 4.1, $S_{12} = \emptyset$. When the maximal difference between profit functions in regime 3 and regime 2 is greater than the difference between the corresponding switching costs, and $(f_2 - f_1)(\infty) > rg_{12}$, we may have interest to switch to regime 2 for intermediate state values, while for large values of the state, one switches to regime 3.

Proof of Theorem 5.1

1) Suppose first that $(f_2 - f_1)(\infty) \leq rg_{12}$ and $(f_3 - f_1)(\infty) \leq rg_{13}$. This is equivalent in view of Proposition 4.1 1) to $S_1 = S_1^+ = \emptyset$. So $C_1 = (0, \infty)$, $v_1 = \hat{V}_1$; i.e., one always stays in regime 1.

2) Suppose now that $(f_2 - f_1)(\infty) > rg_{12}$ or $(f_3 - f_1)(\infty) > rg_{13}$. Equivalently, $S_1 = S_1^+ \neq \emptyset$. We distinguish two subcases :

a) $(f_3 - f_2)(\infty) \le r(g_{13} - g_{12}).$

By Proposition 4.1, we have $j^+(1) = 2$, and $\bar{y}_{13} := \sup \mathcal{S}_{13} < \infty$.

Step a1). Let us prove that $S_{13} = \emptyset$. First, we claim that $\bar{y}_{13} < \bar{x}_{12} := \inf S_{12}$. We argue by contradiction by assuming $\bar{y}_{13} \ge \bar{x}_{12}$. Recalling from Proposition 4.1 2) that S_{12} contains an interval in the form $[x_{12}, \infty)$, together with Lemma 4.2, we have the following two possible cases:

• \bar{y}_{13} is a crossing boundary or 3-isolated point of $S_{12} \cap S_{13}$.

In this case, we have by Lemma 4.3 that $(f_2 - f_3)(\bar{y}_{13}) \leq r(g_{12} - g_{13})$. This would imply under **(Hf)** that $(f_3 - f_2)(\infty) > r(g_{13} - g_{12})$, a contradiction to the assumption of case 2a).

• there exists $x_{12} > \overline{y}_{13}$, $x_{12} \in \mathcal{S}_{12}$ such that $(\overline{y}_{13}, x_{12}) \subset \mathcal{C}_1$.

In this case, it follows from Lemma 4.4 1) that $\bar{y}_{13} < \bar{x}_{12}$, a contradiction.

Hence, $\bar{y}_{13} < \bar{x}_{12}$ and $(\bar{y}_{13}, \bar{x}_{12}) \subset C_1$. Let us now check that $\bar{y}_{13} = 0$, i.e. $S_{13} = \emptyset$. Otherwise, $\bar{y}_{13} \in S_{13}$, and we have by $(4.4) : (f_3 - f_1)(\bar{y}_{13}) \ge rg_{13}$. Moreover, in view of the assumption of case 2a) as well as under **(Hf)** that $f_3 - f_2 \le r(g_{13} - g_{12})$ on $(0, \infty)$, we deduce that

$$(f_2 - f_1)(\bar{y}_{13}) = (f_2 - f_3)(\bar{y}_{13}) + (f_3 - f_1)(\bar{y}_{13}) \\ \geq r(g_{12} - g_{13}) + rg_{13} = rg_{12},$$

a contradiction to Lemma 4.4 1). Therefore, $S_{13} = \emptyset$.

Step a2). Since $\sup S_{13} = 0$, condition (4.17) is trivially satisfied with $i = 1, j^+(i) = 2$, and we deduce from Proposition 4.2 1) that $S_1 = S_{12}$ is in the form $S_{12} = [\bar{x}_{12}, \infty)$ for some $\bar{x}_{12} \in (0, \infty)$.

b) $(f_3 - f_2)(\infty) > r(g_{13} - g_{12}).$ By Proposition 4.1, we have $j^+(1) = 3$.

Step b1). We claim that S_{13} is in the form $S_1 = [\bar{x}_{13}, \infty)$ for some $\bar{x}_{13} \in (0, \infty)$. We already know from Proposition 4.1 2) that S_{13} contains an interval in the form $[x_{13}, \infty)$. We then set $\bar{x}_{13} = \inf\{x_0 \in S_{13} : [x_0, \infty) \subset S_{13}$, and $\exists x \notin S_{13}, x < x_0\}$. It suffices then to show that $\bar{x}_{13} = \inf S_{13}$. We argue by contradiction by assuming on the contrary that $\inf S_{13} < \bar{x}_{13}$. Recall that $\inf S_{13} > 0$ by (4.6). Then, two possible cases could occur :

- there exists $z_0 \in [\inf \mathcal{S}_{13}, \bar{x}_{13}), z_0 \in \mathcal{S}_{1j}$ for some $j \in \{2, 3\}$ s.t. $(z_0, \bar{x}_{13}) \subset \mathcal{C}_1$. By Lemma 4.4 1), we should have $z_0 < \inf \mathcal{S}_{13}$, a contradiction.
- \bar{x}_{13} is a crossing boundary point of $S_{12} \cap S_{13}$. In this case, we must have by Lemma 4.3 :

$$(f_3 - f_2)(\bar{x}_{13}) = r(g_{13} - g_{12}).$$
 (5.1)

We denote $y_0 = \inf\{0 < x_0 < \bar{x}_{13} : [x_0, \bar{x}_{13}] \subset S_{12}\}$, and we notice that $y_0 < \bar{x}_{13}$ since \bar{x}_{13} is a crossing boundary point of $S_{12} \cap S_{13}$. If $y_0 \leq \inf S_{13}$, then $\inf S_{13}$ would be a 3-isolated point of $S_{12} \cap S_{13}$ by Lemma 4.2, and so by Lemma 4.3, $(f_2 - f_3)(\inf S_{13}) \leq r(g_{12} - g_{13})$. Recalling that $\inf S_{13} < \bar{x}_{13}$, this is in contradiction with (5.1) under **(Hf)**. If, on the other hand, $y_0 > \inf S_{13}$, then we have two possibilities :

* (inf S_{13}, y_0) is included in C_1 . Since $\inf S_{13} \in S_{13}$, we have by (4.4): $(f_3 - f_1)(\inf S_{13}) \ge rg_{13}$. Moreover, from (5.1) and under **(Hf)**, we have $(f_3 - f_2)(x) \le r(g_{13} - g_{12})$ for all $x \le \bar{x}_{13}$, so that

$$(f_2 - f_1)(\inf S_{13}) = (f_2 - f_3)(\inf S_{13}) + (f_3 - f_1)(\inf S_{13}) \\ \leq r(g_{12} - g_{13}) + rg_{13} = rg_{12}.$$

This contradicts Lemma 4.4 1).

* y_0 is a crossing boundary of $S_{12} \cap S_{13}$. By Lemma (4.3), this would imply

$$(f_3 - f_2)(y_0) = r(g_{13} - g_{12}), (5.2)$$

a contradiction with (5.1) under (Hf), since $y_0 < \bar{x}_{13}$.

We have thus proved that $S_{13} = [\bar{x}_{13}, \infty)$ with $\bar{x}_{13} = \inf S_{13} > 0$.

Step b2. Let us now study the structure of S_{12} . We suppose that S_{12} is neither empty, nor a singleton, and set $\bar{x}_{12} = \inf S_{12} < \bar{y}_{12} = \sup S_{12}$. Recall that $\bar{x}_{12} > 0$ from (4.6). We also know from Proposition 4.1 2) that $\bar{y}_{12} < \infty$. Let us first show that $\bar{y}_{12} \leq \bar{x}_{13}$. If it is not true, then by Lemma 4.2 and since $S_{13} = [\bar{x}_{13}, \infty)$, \bar{y}_{12} is a 2-isolated point of $S_{12} \cap S_{13}$, and one can find some $y_0 \in S_{12}$ $y_0 < \bar{y}_{12}$ such that $(y_0, \bar{y}_{12}) \cap S_{12} = \emptyset$. By Lemma 4.3, $(f_3 - f_2)(\bar{y}_{12}) \leq r(g_{13} - g_{12})$, and so under **(Hf)**, we have :

$$(f_3 - f_2)(x) \leq r(g_{13} - g_{12}), \quad \forall x < \bar{y}_{12}.$$
 (5.3)

Let us then consider the function $G = v_1 - (v_2 - g_{12})$. Since $(y_0, \bar{y}_{12}) \cap S_{12} = \emptyset$, i.e. $(y_0, \bar{y}_{12}) \subset C_1 \cup S_{13}$, we notice that G is strictly positive on (y_0, \bar{y}_{12}) . Moreover, v_1 satisfies $rv_1 - \mathcal{L}v_1 = f_1$ on C_1 , $v_1 = v_3 - g_{13}$ on $S_{13} \subset C_3$ satisfies : $rv_1 - \mathcal{L}v_1 = f_3 - rg_{13}$, and v_2 satisfies $rv_2 - \mathcal{L}v_2 - f_2 \ge 0$ on $(0, \infty)$. Hence, we deduce that G is a viscosity subsolution to :

$$rG - \mathcal{L}G \leq \max(f_1 - f_2 + rg_{12}, f_3 - f_2 + r(g_{12} - g_{13}))$$
 on (y_0, \bar{y}_{12}) .

Since $\bar{x}_{12} \in \mathcal{S}_{12}$, we have by (4.4) :

$$(f_1 - f_2)(x) + rg_{12} \leq 0, \quad \forall x \geq \bar{x}_{12}.$$
 (5.4)

Together with (5.3), we obtain

$$rG - \mathcal{L}G \leq 0 \quad \text{on} \quad (y_0, \bar{y}_{12}). \tag{5.5}$$

Now, since y_0 and \bar{y}_{12} lie in S_{12} , we have $G(y_0) = G(\bar{y}_{12}) = 0$. From the maximum principle, G should be nonpositive on (y_0, \bar{y}_{12}) , a contradiction. Hence, $\bar{y}_{12} = \sup S_{12} \leq \bar{x}_{13} = \inf S_{13}$. Condition (4.18) is then satisfied with i = 1, k = 2, and we deduce from Proposition 4.2 2) that $S_{12} = [\bar{x}_{12}, \bar{y}_{12}]$.

We now provide an explicit qualitative description of the switching regions for the intermediate regime 2. Notice that with regime 2, we have $S_2^+ = S_{23}$ and $S_2^- = S_{21}$.

Theorem 5.2 (Switching regions in Regime 2)

1) Suppose that $(f_3 - f_2)(\infty) \leq rg_{23}$. Then $S_2^+ = \emptyset$, and $S_2^- = (0, \underline{y}_{21}]$ for some $\underline{y}_{21} > 0$. 2) Suppose that $(f_3 - f_2)(\infty) > rg_{23}$. Then, $S_2^+ = [\bar{x}_{23}, \infty)$ and $S_2^- = (0, \underline{y}_{21}]$ for some $0 < \underline{y}_{21} \leq \bar{x}_{23} < \infty$.

Remark 5.2 The above qualitative description of the switching regions for regime 2 is explicit. When the maximal difference between the profit functions in the highest regime 3 and the current intermediate regime 2 is not large enough for covering the cost of changing of regime, one never switches up. However, if this maximal difference is larger than the corresponding switching cost, then one has interest to switch up starting from a certain threshold in state. On the other hand, one always switches down to the lowest regime 1 once the state value goes below a certain threshold. Moreover, the upward and downward switching regions may intersect only at some crossing boundary.

Proof of Theorem 5.2

1) Suppose that $(f_3 - f_2)(\infty) \leq rg_{23}$, which means, in view of Proposition 4.1 1), that $S_2^+ = S_{23} = \emptyset$, and so inf $S_{23} = \infty$. Condition (4.23) is then trivially satisfied with i = 2, and we deduce from Proposition 4.4 1) that $S_2^- = S_{21} = (0, \underline{y}_{21}]$ with $\underline{y}_{21} \in (0, \infty)$.

2) Suppose that $(f_3 - f_2)(\infty) > rg_{23}$.

Step A. We first claim that S_{23} is in the form $[\bar{x}_{23}, \infty)$ for some $\bar{x}_{23} > 0$. By Proposition 4.1, $S_2^+ = S_{23}$ is nonempty and contains an interval in the form $[x_{23}, \infty)$. We then set $\bar{x}_{23} = \inf\{x_0 \in S_{23} : [x_0, \infty) \subset S_{23}, \text{ and } \exists x \notin S_{23}, x < x_0\}$. It suffices then to show that $\bar{x}_{23} = \inf S_{23}$. We argue by contradiction by assuming on the contrary that $\inf S_{23} < \bar{x}_{23}$. Then, two possible cases could occur :

- there exists $z_0 \in [\inf S_{23}, \bar{x}_{23}), z_0 \in S_{2j}$ for some $j \in \{1, 3\}$ s.t. $(z_0, \bar{x}_{23}) \subset C_2$.
 - In this case, by Lemma 4.4 1), we must have j = 1 and $z_0 > \inf S_{23}$. Moreover, by using again Lemma 4.4 between $\inf S_{23}$ and z_0 , we deduce that S_{23} should intersect with S_{21} at some crossing boundary point ξ_0 with $[\inf S_{23}, \xi_0] \subset S_{23}$ and $[\xi_0, z_0] \subset$ S_{21} . By Lemma 4.3, we have $(f_3 - f_1)(\xi_0) = r(g_{23} - g_{21})$ and so under (**Hf**) :

$$(f_1 - f_3)(x) + r(g_{23} - g_{21}) \leq 0, \quad \forall x \ge \xi_0.$$
(5.6)

By (4.4), we have

$$(f_2 - f_3)(x) + rg_{23} \leq 0, \quad \forall x \geq \inf \mathcal{S}_{23}.$$
 (5.7)

Let us consider the continuous function $G = v_2 - (v_3 - g_{23})$ on $[\xi_0, \bar{x}_{23}]$. Since $v_2 = v_1 - g_{21}$ on $(\xi_0, z_0) \subset C_1$ satisfies : $rv_2 - \mathcal{L}v_2 = f_1 - rg_{21}$, v_2 satisfies on $(z_0, \bar{x}_{23}) \subset C_2$: $rv_2 - \mathcal{L}v_2 = f_2$, and $rv_3 - \mathcal{L}v_3 - f_3 \ge 0$ on $(0, \infty)$, we deduce that G is a viscosity subsolution to :

$$\begin{aligned} rG - \mathcal{L}G &\leq \max \left[f_1 - f_3 + r(g_{23} - g_{21}) , \ f_2 - f_3 + rg_{23} \right] \\ &\leq 0 \quad \text{on} \ (\xi_0, \bar{x}_{ij^+(i)}) \end{aligned}$$

from (5.6)-(5.7). Since $G(\xi_0) = G(\bar{x}_{23}) = 0$, this implies by the standard maximum principle that G is non-positive on $(\xi_0, \bar{x}_{ij^+(i)})$, a contradiction.

• \bar{x}_{23} is a crossing boundary point of $S_{21} \cap S_{23}$.

By the same arguments as those in proving Theorem 5.1 2b) (Step b1), we can show that this case is impossible.

We have then proved that S_{23} is actually equal to $[\bar{x}_{23}, \infty)$ with $\bar{x}_{23} = \inf S_{23}$.

Step B. Let us now study the structure of the downward switching region $S_2^- = S_{21}$. From Proposition 4.3, S_{21} is nonempty and contains an interval in the form $(0, y_{21}]$. We set \underline{y}_{21} := sup $S_{21} < \infty$ by (4.7). Let us show that $\underline{y}_{21} \leq \overline{x}_{23}$. If it is not true, then by Lemma 4.2 and since $S_{23} = [\overline{x}_{23}, \infty)$, \underline{y}_{21} is a 1-isolated point of $S_{21} \cap S_{23}$, and one can find some $y_0 \in S_{21}$, $y_0 < \underline{y}_{21}$ s.t. $(y_0, \underline{y}_{21}) \cap S_{21} = \emptyset$. By Lemma 4.3, $(f_3 - f_1)(\underline{y}_{21}) \leq r(g_{23} - g_{21})$, and so under **(Hf)**, we have :

$$(f_3 - f_1)(x) + r(g_{21} - g_{23}) \leq 0, \quad \forall \ x < \underline{y}_{21}.$$
 (5.8)

By (4.5), we also have

$$(f_2 - f_1)(x) + rg_{21} \leq 0, \quad \forall x \leq \underline{y}_{21}.$$
 (5.9)

Let us then consider the function $G = v_2 - (v_1 - g_{21})$. Since $(y_0, \underline{y}_{21}) \cap S_{21} = \emptyset$, i.e. $(y_0, \underline{y}_{21}) \subset C_2 \cup S_{23}$, we notice that G is strictly positive on $(y_0, \underline{y}_{21})$. Moreover, v_2 satisfies $rv_2 - \mathcal{L}v_2 = f_2$ on C_2 , $v_2 = v_3 - g_{23}$ on $S_{23} \subset C_3$ satisfies : $rv_2 - \mathcal{L}v_2 = f_3 - rg_{23}$, and v_1 satisfies $rv_1 - \mathcal{L}v_1 - f_1 \ge 0$ on $(0, \infty)$. Hence, we deduce that G is a viscosity subsolution to

$$rG - \mathcal{L}G \leq \max(f_2 - f_1 + rg_{21}, f_3 - f_1 + r(g_{21} - g_{23})) \leq 0 \quad \text{on } (y_0, \underline{y}_{21}),$$

from (5.8)-(5.9). Since $G(y_0) = G(\underline{y}_{21}) = 0$, this implies by the maximum principle that G is non-positive on $(y_0, \underline{y}_{21})$, a contradiction. Therefore, $\underline{y}_{21} = \sup \mathcal{S}_{21} \leq \overline{x}_{23} = \inf \mathcal{S}_{23}$. Condition (4.23) is then satisfied, and we deduce from Proposition 4.4 1) that $\mathcal{S}_{21} = (0, \underline{y}_{21}]$.

We finally provide a quasi-explicit qualitative description of the switching regions in the highest regime 3. Notice that with regime 3 we have $S_3 = S_3^- = S_{31} \cup S_{32}$.

Theorem 5.3 (Switching regions in Regime 3)

We have $S_{31} = (0, \underline{y}_{31}]$ for some $\underline{y}_{31} > 0$. Moreover, S_{32} is either empty or in the form $[\underline{x}_{32}, \underline{y}_{32}]$ for some $\underline{y}_{31} \le \underline{x}_{32} \le \underline{y}_{32} < \infty$.

Remark 5.3 This theorem states that one always switches down to the lowest regime 1 once the state value goes below a certain threshold. Moreover, one may have also interest to switch down to the intermediate regime when the state value lies in some closed bounded interval, which intersects possibly with the switching region for regime 1, but only at a crossing boundary.

Proof of Theorem 5.3

Step 1. By Proposition 4.3, we know that S_{31} contains an interval in the form $(0, y_{31}]$ for some $y_{31} > 0$ and $\inf S_{31} = 0$. We set $\underline{y}_{31} = \sup\{y_0 \in S_{31} : (0, y_0] \subset S_{31}$, and $\exists y \notin S_{31}, y_0 < y\}$. Let us show that $\underline{y}_{31} = \sup S_{31}$ so that $S_{31} = (0, \underline{y}_{31})$. We argue by contradiction by assuming on the contrary that $\underline{y}_{31} < \sup S_{31}$. Then, two possible cases could occur:

• Case 1: there exists $z_0 \in (\underline{y}_{31}, \sup \mathcal{S}_{31}], z_0 \in \mathcal{S}_{3j}$ for some $j \in \{1, 2\}$ such that $(\underline{y}_{31}, z_0) \subset \mathcal{C}_3$.

In this case, by Lemma 4.4 2), we should have $z_0 > \sup S_{31}$, a contradiction.

• Case $2: \underline{y}_{31}$ is a crossing boundary point of $S_{31} \cap S_{32}$.

In this case, we must have by Lemma 4.3 :

$$(f_2 - f_1)(\underline{y}_{31}) = r(g_{32} - g_{31}).$$
 (5.10)

We denote $y_0 = \sup\{\xi_0 > \underline{y}_{31} : [\underline{y}_{31}, \xi_0] \subset S_{32}\}$, and notice that $y_0 > \underline{y}_{31}$ since \underline{x}_{31} is a crossing boundary point of $S_{31} \cap S_{32}$. If $y_0 \ge \sup S_{31}$, then $\sup S_{31}$ would be a 1-isolated point of $S_{31} \cap S_{32}$ by Lemma 4.2, and so by Lemma 4.3, $(f_2 - f_1)(\sup S_{31}) \le r(g_{32} - g_{21})$. Recalling that $\sup S_{31} > \underline{y}_{31}$, this contradicts (5.10) under **(Hf)**. If $y_0 < \sup S_{31}$, then we have two possibilities :

* $(y_0, \sup \mathcal{S}_{31})$ is included in \mathcal{C}_3 . By (4.5), we have $(f_3 - f_1)(\sup \mathcal{S}_{31}) \leq -rg_{31}$. Moreover, from (5.10) and under **(Hf)**, we have $(f_2 - f_1)(x) \geq r(g_{32} - g_{31})$ for all $x \geq \underline{y}_{31}$, so that

$$(f_3 - f_2)(\sup \mathcal{S}_{31}) = (f_3 - f_1)(\sup \mathcal{S}_{31}) + (f_1 - f_2)(\sup \mathcal{S}_{31})$$

$$\leq -rg_{31} + r(g_{31} - g_{32}) = -rg_{32}.$$

This is in contradiction with Lemma $4.4\ 2$).

* y_0 is a crossing boundary of $\mathcal{S}_{31} \cap \mathcal{S}_{32}$. By Lemma (4.3), this would imply

$$(f_2 - f_1)(y_0) = r(g_{32} - g_{31}),$$
 (5.11)

a contradiction with (5.10) under (**Hf**), since $y_0 > \underline{y}_{31}$.

We have then proved that $S_{31} = (0, \underline{y}_{31}]$ with $\underline{y}_{31} = \sup S_{31}$.

Step 2. Let us now study the structure of S_{32} . Suppose that S_{32} is neither empty, nor a singleton. We recall that $\underline{y}_{32} := \sup S_{32} < \infty$ by (4.7) and that $\underline{x}_{32} := \inf S_{32} > 0$ thanks to Proposition 4.3. Let us first prove that $\underline{y}_{31} \leq \underline{x}_{32}$. If not, then by Lemma 4.2 and since $S_{31} = (0, \underline{y}_{31}], \underline{x}_{32}$ is a 2-isolated point of $S_{31} \cap S_{32}$, and one can find some $y_0 \in S_{32}, \underline{x}_{32} < y_0$ such that $(\underline{x}_{32}, y_0) \cap S_{32} = \emptyset$. By Lemma 4.3, $(f_1 - f_2)(\underline{x}_{32}) \leq r(g_{31} - g_{32})$, and so under (**Hf**), we have :

$$(f_1 - f_2)(x) + r(g_{32} - g_{31}) \leq 0, \quad \forall \ x \ge \underline{x}_{32}.$$
(5.12)

By (4.5), we also have

$$(f_3 - f_2)(x) + rg_{32} \leq 0, \quad \forall x \leq \underline{y}_{32}.$$
 (5.13)

Let us then consider the function $G = v_3 - (v_2 - g_{32})$. Since $(\underline{x}_{32}, y_0) \cap S_{32} = \emptyset$, i.e. $(\underline{x}_{32}, y_0) \subset C_3 \cup S_{31}$, we notice that G is strictly positive on $(\underline{x}_{32}, y_0)$. Moreover, v_3 satisfies $rv_3 - \mathcal{L}v_3 = f_3$ on C_3 , $v_3 = v_1 - g_{31}$ on $S_{31} \subset C_1$ satisfies : $rv_3 - \mathcal{L}v_3 = f_1 - rg_{31}$, and v_2 satisfies $rv_2 - \mathcal{L}v_2 - f_2 \ge 0$ on $(0, \infty)$. Hence, we deduce that G is a viscosity subsolution to

$$rG - \mathcal{L}G \leq \max(f_3 - f_2 + rg_{32}, f_1 - f_2 + r(g_{32} - g_{32})) \leq 0$$
 on $(\underline{x}_{32}, y_0)$

from (5.12)-(5.13). Since $G(\underline{x}_{32}) = G(y_0) = 0$, this implies by the maximum principle that G is non-positive on $(\underline{x}_{32}, y_0)$, a contradiction. Therefore, $\underline{y}_{31} = \sup \mathcal{S}_{31} \leq \underline{x}_{32} = \inf \mathcal{S}_{32}$. Condition (4.24) is then satisfied with i = 3, k = 2, and we conclude from Proposition 4.4 2) that $\mathcal{S}_{32} = [\underline{x}_{32}, \underline{y}_{32}]$.

We finally summarize the qualitative structure of the switching regions in the threeregime model; see also Figure 1 for a visual aid.

Theorem 5.4 (Switching regions in the three-regime model)

We have the following four cases : A) If $(f_2 - f_1)(\infty) \le rg_{12}$, $(f_3 - f_1)(\infty) \le rg_{13}$, and $(f_3 - f_2)(\infty) \le rg_{23}$, then

$$\begin{split} &\mathcal{S}_{1} = \mathcal{S}_{1}^{+} = \emptyset, \\ &\mathcal{S}_{2}^{-} = (0, \underline{y}_{21}], \qquad \mathcal{S}_{2}^{+} = \emptyset, \\ &\mathcal{S}_{31} = (0, \underline{y}_{31}], \qquad \mathcal{S}_{32} \text{ is either empty or } \mathcal{S}_{32} = [\underline{x}_{32}, \underline{y}_{32}] \end{split}$$

for some $0 < \underline{y}_{31} \leq \underline{x}_{32} \leq \underline{y}_{32} < \infty, \ 0 < \underline{y}_{21} < \underline{x}_{32}.$

B) If
$$(f_2 - f_1)(\infty) > rg_{12}$$
 or $(f_3 - f_1)(\infty) > rg_{13}$, and $(f_3 - f_2)(\infty) \le r(g_{13} - g_{12})$, then

$$\begin{aligned} \mathcal{S}_{12} &= [\bar{x}_{12}, \infty), & \mathcal{S}_{13} &= \emptyset, \\ \mathcal{S}_{2}^{-} &= (0, \underline{y}_{21}], & \mathcal{S}_{2}^{+} &= \emptyset, \\ \mathcal{S}_{31} &= (0, \underline{y}_{31}], & \mathcal{S}_{32} \text{ is either empty or } \mathcal{S}_{32} &= [\underline{x}_{32}, \underline{y}_{32}] \end{aligned}$$

for some $0 < \underline{y}_{21} \le \overline{x}_{12}, 0 < \underline{y}_{31} \le \underline{x}_{32} \le \underline{y}_{32} < \infty, \underline{y}_{31} < \overline{x}_{12} < \infty, 0 < \underline{y}_{21} < \underline{x}_{32}.$ C) If $(f_3 - f_1)(\infty) > rg_{13}$, and $r(g_{13} - g_{12}) < (f_3 - f_2)(\infty) \le rg_{23}$, then

$$\begin{aligned} \mathcal{S}_{13} &= [\bar{x}_{13}, \infty), & \mathcal{S}_{12} \text{ is either empty or } \mathcal{S}_{12} &= [\bar{x}_{12}, \bar{y}_{12}] \\ \mathcal{S}_2^- &= (0, \underline{y}_{21}], & \mathcal{S}_2^+ &= \emptyset, \\ \mathcal{S}_{31} &= (0, \underline{y}_{31}], & \mathcal{S}_{32} \text{ is either empty or } \mathcal{S}_{32} &= [\underline{x}_{32}, \underline{y}_{32}], \end{aligned}$$

for some $0 < \bar{x}_{12} \le \bar{y}_{12} \le \bar{x}_{13} < \infty$, $0 < \underline{y}_{31} \le \underline{x}_{32} \le \underline{y}_{32} < \infty$, $0 < \underline{y}_{21} < \bar{x}_{12}$, $\underline{y}_{31} < \underline{x}_{13}$. D) If $(f_3 - f_1)(\infty) > rg_{13}$, and $(f_3 - f_2)(\infty) > rg_{23}$, then

$$\begin{aligned} \mathcal{S}_{13} &= [\bar{x}_{13}, \infty), & \mathcal{S}_{12} \text{ is either empty or } \mathcal{S}_{12} &= [\bar{x}_{12}, \bar{y}_{12}] \\ \mathcal{S}_2^- &= (0, \underline{y}_{21}], & \mathcal{S}_2^+ &= [\bar{x}_{23}, \infty), \\ \mathcal{S}_{31} &= (0, \underline{y}_{31}], & \mathcal{S}_{32} \text{ is either empty or } \mathcal{S}_{32} &= [\underline{x}_{32}, \underline{y}_{32}], \end{aligned}$$

 $\begin{array}{l} \text{for some } 0 < \bar{x}_{12} \leq \bar{y}_{12} \leq \bar{x}_{13} < \infty, \, 0 < \underline{y}_{31} \leq \underline{x}_{32} \leq \underline{y}_{32} < \bar{x}_{23} < \infty, \, 0 < \underline{y}_{21} < \bar{x}_{12}, \, \bar{y}_{12} < < \bar{x}_{23}, \, \underline{y}_{31} < \underline{x}_{13}. \end{array}$

Proof. Assertion A) follows from Theorem 5.1 1), Theorem 5.2 1) and Theorem 5.3. Assertion B) follows from Theorem 5.1 2a), Theorem 5.2 1) and Theorem 5.3, and observing that $g_{13} - g_{12} < g_{23}$ from (2.13). Assertion C) follows from Theorem 5.1 2b), Theorem 5.2 1) and Theorem 5.3, and observing that $(f_3 - f_1)(\infty) > rg_{13}$ when $(f_2 - f_1)(\infty) > rg_{12}$ and $r(g_{13} - g_{12}) < (f_3 - f_2)(\infty)$. Assertion D) follows from Theorem 5.1 2b), Theorem 5.2 2) and Theorem 5.3. Finally, the other ordering condition on the thresholds of the switching regions follows from the observation that a switching region S_{ij} is included in the continuation region C_j , and hence never intersects with S_{jk} .

6 Numerical procedure

The qualitative structure of optimal switching controls derived in the previous section states that the optimal sequence of stopping times is given by the hitting times of the diffusion process X at a finite number of threshold levels. This is of vital importance in eventually solving (either analytically or numerically) our problem, because it reduces the originally very complex problem into one finding a small number of critical values in state which is a finite-dimensional optimization problem. In this section, we demonstrate how to design algorithms to find these critical values in state. We shall take case B) in Theorem 5.4 to showcase our approach. This is not unduly technical, yet it is rich enough to catch the essential feature and difficulty of such a problem.



Figure 1: Switching regions in the three-regime model

The optimal control of X_t is completely dictated by the feedback law in Theorem 5.4, case B), except we do not yet know the values of the five threshold parameters : \bar{x}_{12} , \underline{y}_{21} , \underline{y}_{31} , \underline{x}_{32} , \underline{y}_{32} , which we are now to find. Notice that these parameters should satisfy :

$$0 < \underline{y}_{21} \le \overline{x}_{12}, \ 0 < \underline{y}_{31} < \overline{x}_{12}, \ 0 < \underline{y}_{21} < \underline{x}_{32}, \ 0 < \underline{y}_{31} \le \underline{x}_{32} < \underline{y}_{32}.$$

In accordance with our notation, we denote by $J_i(x)$ the total profit if one starts with regime i (i = 1, 2, 3) and state x > 0. Bear in mind that $J_i(x)$ would depend on the values of the five threshold parameters, and that the value function $v_i(x)$ is the maximum of $J_i(x)$ over these parameters. Actually, we shall explicitly express $J_i(x)$ and $v_i(x)$ in terms of expectation functionals involving hitting times for the state process X, and derive a procedure to compute the optimal threshold parameters.

We first introduce some notations. For x > 0, a, b > 0, let

$$\tau_a^x = \inf\{t \ge 0: \ X_t^x = a\}, \qquad \tau_{ab}^x = \inf\{t \ge \tau_a: \ X_t^x = b\}.$$
(6.1)

Fix $\rho > 0$, and denote

$$R(x,a) = E[e^{-\rho\tau_a^x}], \qquad \hat{R}(x,a,b) = E[e^{-\rho\tau_a^x}], \quad \tilde{R}(x,a,b) = E[e^{-\rho\tau_a^x}\mathbf{1}_{\{\tau_a^x < \tau_b^x\}}]. \quad (6.2)$$

For any measurable function f on \mathbb{R}_+ with linear growth condition, and for x, a, b > 0, we also set

$$F(f;x,a) = E\Big[\int_{0}^{\tau_{a}^{x}} e^{-\rho t} f(X_{t}^{x}) dt\Big], \qquad \hat{F}(f;x,a,b) = E\Big[\int_{\tau_{a}^{x}}^{\tau_{a}^{x}} e^{-\rho t} f(X_{t}^{x}) dt\Big],$$
$$\tilde{F}(f;x,a,b) = E\Big[\int_{0}^{\tau_{a}^{x} \wedge \tau_{b}^{x}} e^{-\rho t} f(X_{t}^{x}) dt\Big].$$
(6.3)

These expectation functionals R, \hat{R} , \tilde{R} , F, \hat{F} , and \tilde{F} , are calculated in closed analytic form when X is a GBM (see Appendix), and may be approximated by Euler scheme discretization and Monte-Carlo simulations for general diffusion X.

We now turn to the computation of the value functions v_i and the optimal threshold parameters. This is achieved in two steps.

Step I : determination of threshold parameters and value functions in regimes 1 and 2

We start with regime 1. Fix some x > 0, and let us compute $J_1(x)$ and $v_1(x)$. Notice that since starting from regime 1 we can only switch eventually to regime 2, $J_1(x)$ depends only on the two threshold parameters \overline{x}_{12} and \underline{y}_{21} . We stress this dependence by writing $J_1(x) = J_1(x, \overline{x}_{12}, \underline{y}_{21})$, and calculate the function $J_1(x, .)$ on the domain $\mathcal{D}_{12} = \{(\overline{x}_{12}, \underline{y}_{21}) \in (0, \infty)^2 : \underline{y}_{21} \leq \overline{x}_{12}\}$ as follows :

• If $\overline{x}_{12} > x$, then the optimal strategy is to let the process diffuse until it hits \overline{x}_{12} at the stopping time $\tau^x_{\overline{x}_{12}}$ when one switches to regime 2 (and paying the cost g_{12}), and then lets it diffuse. One then switches back to regime 1 (paying the cost g_{21}) as soon

as the process hits $\underline{y}_{_{21}}$ at $\tau^x_{\overline{x}_{_{12}}\underline{y}_{_{21}}}$. After $\tau^x_{\overline{x}_{_{12}}\underline{y}_{_{21}}}$, the process repeats itself as described above. Specifically, we have for $\overline{x}_{_{12}} > x$:

$$J_{1}(x, \overline{x}_{12}, \underline{y}_{21}) = E \bigg[\int_{0}^{\tau_{\overline{x}_{12}}^{x}} e^{-\rho t} f_{1}(X_{t}^{x}) dt - e^{-\rho \tau_{\overline{x}_{12}}^{x}} g_{12} + \int_{\tau_{\overline{x}_{12}}^{x}}^{\tau_{\overline{x}_{12}}^{x}} e^{-\rho t} f_{2}(X_{t}^{x}) dt + e^{-\rho \tau_{\overline{x}_{12}}^{x}} \bigg(-g_{21} + J_{1}(\underline{y}_{21}, \overline{x}_{12}, \underline{y}_{21}) \bigg) \bigg] = F(f_{1}; x, \overline{x}_{12}) - R(x, \overline{x}_{12}) g_{12} + \hat{F}(f_{2}; x, \overline{x}_{12}, \underline{y}_{21}) + \hat{R}(x, \overline{x}_{12}, \underline{y}_{21}) \bigg(-g_{21} + J_{1}(\underline{y}_{21}, \overline{x}_{12}, \underline{y}_{21}) \bigg).$$
(6.4)

Notice that this last relation for $x = \underline{y}_{21}$ yields the expression of $J_1(\underline{y}_{21}, \overline{x}_{12}, \underline{y}_{21})$ by :

$$\begin{pmatrix} 1 - \hat{R}(\underline{y}_{21}, \overline{x}_{12}, \underline{y}_{21}) \end{pmatrix} J_1(\underline{y}_{21}, \overline{x}_{12}, \underline{y}_{21})$$

$$= F(f_1; \underline{y}_{21}, \overline{x}_{12}) - R(\underline{y}_{21}, \overline{x}_{12}) g_{12} + \hat{F}(f_2; \underline{y}_{21}, \overline{x}_{12}, \underline{y}_{21}) - \hat{R}(\underline{y}_{21}, \overline{x}_{12}, \underline{y}_{21}) g_{21}$$

$$= (6.5)$$

Hence, plugging the above into (6.4), we obtain an explicit expression of $J_1(x, \overline{x}_{12}, \underline{y}_{21})$ for $\overline{x}_{12} > x$, in terms of the expectation functionals R, \hat{R} , F and \hat{F} .

• If $\overline{x}_{12} \leq x$, then the optimal strategy is to switch immediately at time t = 0 to regime 2, and then as in the first case, let it diffuse until it hits \underline{y}_{21} and so on. Similarly as above, we express $J_1(x, \overline{x}_{12}, y_{21})$ for $\overline{x}_{12} \leq x$ in terms of R, $\hat{R} \in F$ and \hat{F} :

$$J_{1}(x,\overline{x}_{12},\underline{y}_{21}) = -g_{12} + E \Big[\int_{0}^{\tau_{\underline{y}_{21}}^{x}} e^{-\rho t} f_{2}(X_{t}^{x}) dt + e^{-\rho \tau_{\underline{y}_{21}}^{x}} \Big(-g_{21} + J_{1}(\underline{y}_{21},\overline{x}_{12},\underline{y}_{21}) \Big) \Big]$$

$$= -g_{12} + F(f_{2},x,\underline{y}_{21}) + R(x,\underline{y}_{21}) \Big(-g_{21} + J_{1}(\underline{y}_{21},\overline{x}_{12},\underline{y}_{21}) \Big).$$
(6.6)

The value $v_1(x)$, which is the maximum of the expected profit over the threshold parameters, is then determined by :

$$v_1(x) = \max_{(\overline{x}_{12}, \underline{y}_{21}) \in \mathcal{D}_{12}} J_1(x, \overline{x}_{12}, \underline{y}_{21}),$$
(6.7)

and the optimal threshold parameters \overline{x}_{12}^* and \underline{y}_{21}^* are the solutions to the above argmax optimization (which is trivially easy). Notice that the optimal threshold parameters do not depend on the current state value x; so once these have been calculated, the computation of $v_1(x')$ for other state values x' > 0, is directly derived from the relation :

$$v_1(x') = J_1(x', \overline{x}_{12}^*, \underline{y}_{21}^*).$$

Moreover, the value function $v_2(x)$ for regime 2 is derived as follows :

=

• if $x > \underline{y}_{21}^*$, then the optimal strategy is to let the process diffuse until it hits \underline{y}_{21}^* at the stopping time $\tau_{\underline{y}_{12}^*}^x$ when one switchs to regime 1 and finally follows the remaining optimal strategy computed above . Hence,

$$v_{2}(x) = E\left[\int_{0}^{\tau_{\underline{y}_{21}}^{x}} e^{-\rho t} f_{2}(X_{t}^{x}) dt + e^{-\rho \tau_{\underline{y}_{21}}^{x}} \left(-g_{21} + v_{1}(\underline{y}_{21}^{*})\right)\right]$$

$$= F(f_{2}, x, \underline{y}_{21}^{*}) + R(x, \underline{y}_{21}^{*}) \left(-g_{21} + v_{1}(\underline{y}_{21}^{*})\right).$$

• $x \leq \underline{y}_{21}^*$, then the optimal strategy is to switch immediately to regime 1. Hence,

$$v_2(x) = -g_{21} + v_1(x).$$

Step II : determination of threshold parameters and value function in regime 3

The optimal threshold parameters \overline{x}_{12}^* , \underline{y}_{21}^* and the value functions v_1 and v_2 for regimes 1 and 2 are known from Step I, and we now move on to find the threshold parameters in regime 3 : \underline{y}_{31} , and \underline{x}_{32} , \underline{y}_{32} if they exist, together with the value function $v_3(x)$. What is tricky here is that Theorem 5.4 does not decisively tell us whether the interval of switching from regime 3 to regime 2, $S_{32} = [\underline{x}_{32}, \underline{y}_{32}]$, exists. The approach presented here is able to determine such existence, by comparing the two options when starting from regime 3 : one is to let the process diffuse until it hits \underline{y}_{31} , and then follow the remaining optimal strategy, and the other is to switch to regime 2, and then follow the optimal strategy afterwards. If the latter option is better off for some state value x, then it suggests that $[\underline{x}_{32}, \underline{y}_{32}]$ is nonempty. So, we start from regime 3, fix some x > 0, and consider the following two options :

▶ Option 1 : the interval $[\underline{x}_{32}, \underline{y}_{32}]$ does not exist. Denote $J_3^{(1)}(x, \underline{y}_{31})$ the corresponding expected profit in regime 3 with threshold value \underline{y}_{31} . The function $\underline{y}_{31} \to J_3^{(1)}(x, \underline{y}_{31})$ is calculated on the domain $(0, \overline{x}_{12}^*)$ as follows :

• If $\underline{y}_{_{31}} < x$, then the optimal strategy would be to let the process diffuse until it hits $\underline{y}_{_{31}}$ when one switchs to regime 1, and then follow the remaining optimal strategy determined in Step I. Hence, we have for $\underline{y}_{_{21}} < x$:

$$J_{3}^{(1)}(x,\underline{y}_{31}) = E\left[\int_{0}^{\tau_{\underline{y}_{31}}^{x}} e^{-\rho t} f_{3}(X_{t}^{x}) dt + e^{-\rho \tau_{\underline{y}_{31}}^{x}} \left(-g_{31} + v_{1}(\underline{y}_{31})\right)\right]$$

$$= F(f_{3}, x, \underline{y}_{31}) + R(x, \underline{y}_{31}) \left(-g_{31} + v_{1}(\underline{y}_{31})\right).$$
(6.8)

• If $\underline{y}_{_{31}} \ge x$, then the optimal strategy would be to switch immediately to regime 1, and then follow the remaining optimal strategy determined in Step I. Hence, we have for $\underline{y}_{_{31}} \ge x$:

$$J_3^{(1)}(x,\underline{y}_{31}) = -g_{31} + v_1(x).$$
(6.9)

The maximal expected profit in Option 1 is then given by :

$$v_3^{(1)}(x) = \sup_{\underline{y}_{31} \in (0, \overline{x}_{12}^*)} J_3^{(1)}(x, \underline{y}_{31}).$$
(6.10)

▶ Option 2: the interval $[\underline{x}_{32}, \underline{y}_{32}]$ exists (possibly a singleton). Denote $J_3^{(2)}(x, \underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32})$ the corresponding expected profit in regime 3 with threshold parameters $\underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}$. The function $J_3^{(2)}(x, .)$ is calculated on the domain $\mathcal{D}_3^{(2)} = \{(\underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) \in (0, \infty)^3 : \underline{y}_{31} < \overline{x}_{12}, \underline{y}_{21}^* < \underline{x}_{32}, \underline{y}_{31} \leq \underline{x}_{32} \leq \underline{y}_{32}\}$ as follows :

• If $x > \underline{y}_{32}$, then the optimal strategy is to let the process diffuse until it hits \underline{y}_{32} when one switchs to regime 2, and then follow the remaining optimal strategy determined in Step I. Hence, we have for $x > \underline{y}_{32}$:

$$J_{3}^{(2)}(x,\underline{y}_{31},\underline{x}_{32},\underline{y}_{32}) = E\Big[\int_{0}^{\tau_{\underline{y}_{32}}^{x}} e^{-\rho t} f_{3}(X_{t}^{x}) dt + e^{-\rho \tau_{\underline{y}_{32}}^{x}} \Big(-g_{32} + v_{2}(\underline{y}_{32})\Big)\Big]$$

$$= F(f_{3},x,\underline{y}_{32}) + R(x,\underline{y}_{32}) \Big(-g_{32} + v_{2}(\underline{y}_{32})\Big).$$
(6.11)

If <u>x</u>₃₂ ≤ x ≤ <u>y</u>₃₂, then the optimal strategy is to switch immediately to regime 2, and then follow the remaining optimal strategy determined in Step I. Hence, we have for <u>x</u>₃₂ ≤ x ≤ <u>y</u>₃₂:

$$J_3^{(2)}(x, \underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) = -g_{32} + v_2(x).$$
(6.12)

• If $\underline{y}_{_{31}} < x < \underline{x}_{32}$, then the optimal strategy is to let the process diffuse until it hits either $\underline{y}_{_{31}}$ (when one switches to regime 1) or \underline{x}_{32} (when one switches to regime 2), and then follow the remaining optimal strategy determined in Step I. Hence, we have for $\underline{y}_{_{31}} < x < \underline{x}_{32}$:

$$\begin{aligned} J_{3}^{(2)}(x,\underline{y}_{31},\underline{x}_{32},\underline{y}_{32}) &= E\Big[\int_{0}^{\tau_{\underline{y}_{31}}^{x}\wedge\tau_{\underline{x}_{32}}^{x}} e^{-\rho t} f_{3}(X_{t}^{x}) dt \\ &+ e^{-\rho \tau_{\underline{y}_{31}}^{x}} \left(-g_{31}+v_{1}(\underline{y}_{31})\right) \mathbf{1}_{\tau_{\underline{y}_{31}}^{x} < \tau_{\underline{x}_{32}}^{x}} \\ &+ e^{-\rho \tau_{\underline{x}_{32}}^{x}} \left(-g_{32}+v_{2}(\underline{x}_{32})\right) \mathbf{1}_{\tau_{\underline{x}_{32}}^{x} < \tau_{\underline{y}_{31}}^{x}}\Big] \\ &= \tilde{F}(f_{3};\underline{y}_{31},\underline{x}_{32}) + \tilde{R}(x,\underline{y}_{31},\underline{x}_{32}) \left(-g_{31}+v_{1}(\underline{y}_{31})\right) \\ &+ \tilde{R}(x,\underline{x}_{32},\underline{y}_{31}) \left(-g_{32}+v_{2}(\underline{x}_{32})\right). \end{aligned}$$
(6.13)

• If $\underline{y}_{_{31}} = x = \underline{x}_{32}$, then the optimal strategy is to switch immediately to regime 1 or 2, and to follow the remaining optimal strategy determined in Step 1. Hence, we have for $\underline{y}_{_{31}} = x = \underline{x}_{32}$:

$$J_3^{(2)}(x, \underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) = -g_{31} + v_1(x) = -g_{32} + v_2(x).$$
(6.14)

• If $x \leq \underline{y}_{_{31}}$, then the optimal strategy would be to switch immediately to regime 1, and then follow the remaining optimal strategy computed in Step I. Hence, we have for $x \leq \underline{y}_{_{31}}$:

$$J_{3}^{(2)}(x, \underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) = -g_{31} + v_1(x).$$
(6.15)

The maximal expected profit in Option 2 is then given by :

$$v_{3}^{(2)}(x) = \sup_{(\underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) \in \mathcal{D}_{3}^{(2)}} J_{3}^{(2)}(x, \underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}).$$
(6.16)

▶ Now, we compare these two options characterized by the values $v_3^{(1)}(x)$ and $v_3^{(2)}(x)$. First, notice that when $x \leq \underline{y}_{31}^*$, no conclusion could be drawn concerning the existence of the switching region S_{32} as we would obtain $v_3^{(1)}(x) = v_3^{(2)}(x)$, and the optimal policy in both Options would be to switch directly to regime 1. We therefore take some $x > \overline{x}_{12}^*$ (computed in Step I), which ensures $x > \underline{y}_{31}^*$ (albeit to be determined).

• If $v_3^{(1)}(x) > v_3^{(2)}(x)$, then the switching region S_{32} does not exist, and there is only one optimal threshold value \underline{y}_{31}^* in regime 3, which is the solution to (6.10). Moreover, as this optimal threshold parameter does not depend on the current state value x, once it has been calculated, the computation of the value function $v_3(x')$ in regime 3 for other state x' > 0, is directly derived from the relation :

$$v_3(x') = J_3^{(1)}(x', \underline{y}_{31}^*).$$

- If $v_3^{(1)}(x) \leq v_3^{(2)}(x)$, then the switching region S_{32} exists. However, solving (6.16) may not yet yield all the three optimal threshold values \underline{y}_{31}^* , \underline{x}_{32}^* , \underline{y}_{32}^* in regime 3. Indeed, by solving the optimisation problem (6.16), we obtain four different possible outcomes depending on the position of x:
 - (a) Argmax $v_3^{(2)}(x) = \{(\underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) : x \leq \underline{y}_{31} \leq \underline{x}_{32} \leq \underline{y}_{32}\},\$
 - (b) Argmax $v_3^{(2)}(x) = \{(\underline{y}_{31}^*, \underline{x}_{32}^*, \underline{y}_{32}) : \underline{y}_{31}^* < x < \underline{x}_{32}^* \le \underline{y}_{32}\},\$
 - (c) Argmax $v_3^{(2)}(x) = \{(\underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) : \underline{y}_{31} \leq \underline{x}_{32} \leq x \leq \underline{y}_{32}\},\$

(d) Argmax
$$v_3^{(2)}(x) = \{(\underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}^*) : \underline{y}_{31} \le \underline{x}_{32} \le \underline{y}_{32}^* < x\},\$$

As such, solving (6.16) provides us with the relative position of x compared to those threshold values, and in some cases, some optimal threshold values. More precisely, while cases (a) and (c) do not yield the optimal threshold values, in cases (b) and (d), we respectively obtain the threshold values $\{\underline{y}_{31}^*, \underline{x}_{32}^*\}$ and $\{\underline{y}_{32}^*\}$. Therefore, we have to solve (6.16) for two different values x_b and x_d such that $\underline{y}_{31}^* < x_b < \underline{x}_{32}^*$ and $\underline{y}_{32}^* < x_d$. The difficult point here is that we do not know a priori those optimal threshold values. As such, to obtain two values x_b and x_d respectively in $(\underline{y}_{31}^*, \underline{x}_{32}^*)$ and $(\underline{y}_{32}^*, \infty)$, we shall complete the following iteration. Starting from our initial value x, we solve (6.16) and discuss the above four different outcomes respectively:

- (a) : actually, since our chosen initial value satisfies $x > \bar{x}_{12}^*$, this case could not happen.
- (b) : an obvious candidate for x_b is x itself. Then, we want to find a candidate x_d such that $x_d > \underline{y}_{32}^*$. Starting with $x_0 = x$, and $x_{j+1} = 2x_j$, we solve successively (6.16) for x_j until we get case (d) as the outcome, in other words, until we get a \overline{j} such that $x_{\overline{j}} > \underline{y}_{32}^*$. We take $x_d = x_{\overline{j}}$.
- (d) : an obvious candidate for x_d is x itself. Then, we want to find a candidate x_b such that $\underline{y}_{31}^* < x_b < \underline{x}_{32}^*$. Setting $x_0 = x$, and $x_1 = \frac{x_0}{2}$, we solve successively (6.16) for x_j , starting with j = 1. If for x_j , we fall into

- \star case (c) or case (d), then we set $x_{j+1} = \frac{x_j}{2}$,
- \star case (b), we stop the iteration,
- * case (a), then we set $x_{j+1} = \frac{x_j + x_{j-1}}{2}$. Once our iteration stops, we obtain a \overline{j} such that $\underline{y}_{31}^* < x_{\overline{j}} < \underline{x}_{32}^*$. We take $x_b = x_{\overline{j}}$.
- (c) starting from our initial value, we apply the iterations used in the two previous cases to obtain two satisfying values x_b and x_d .

The resolutions of (6.16) associated with x_b and x_d give us all the three optimal threshold values.

Moreover, once we have obtained the three optimal threshold values, the computation of the value function $v_3(x')$ in regime 3 for other state values x' > 0, is directly derived from the relation :

$$v_3(x') = J_3^{(2)}(x', \underline{y}_{31}^*, \underline{x}_{32}^*, \underline{y}_{32}^*)$$

Summary. To summarize, we have the following algorithm in computing the (at most) five parameters for case B) in Theorem 5.4.

• Fix x > 0, and compute the function $(\overline{x}_{12}, \underline{y}_{21}) \to J_1(x, \overline{x}_{12}, \underline{y}_{21})$ by (6.4)-(6.5)-(6.6). Then, solve (6.7), which gives the optimal threshold values $\overline{x}_{12}^*, \underline{y}_{21}^*$ in regimes 1 and 2.

• Fix $x > \overline{x}_{12}^* > 0$.

(i) Compute the function $\underline{y}_{31} \to J_3^{(1)}(x, \underline{y}_{31})$ by (6.8)-(6.9), and solve $v_3^{(1)}(x)$ in (6.10). (ii) Compute the function $(\underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32}) \to J_3^{(2)}(x, \underline{y}_{31}, \underline{x}_{32}, \underline{y}_{32})$ by (6.11)-(6.12)-(6.13)-(6.14)-(6.15), and solve $v_3^{(2)}(x)$ in (6.16). (iii) If $v_3^{(1)}(x) > v_3^{(2)}(x)$ then the switching region Sec is empty, and the optimal threshold

(iii) If $v_3^{(1)}(x) > v_3^{(2)}(x)$, then the switching region S_{32} is empty, and the optimal threshold value \underline{y}_{31}^* in regime 3 is the solution to (6.10). Otherwise, S_{32} is nonempty, and the optimal threshold values in regime 3, \underline{y}_{31}^* , \underline{x}_{32}^* , \underline{y}_{32}^* are the solutions to (6.16) (solved for two well-chosen values x_b and then x_d , see the above discussion).

Although we have demonstrated here only one case in the three-regime model, the essence remains the same for other cases or even general multi-regime model. What is important here is that the qualitative structure obtained for a general model makes it possible to turn the problem into a few finite-dimensional mathematical programming problems, which are very easy to solve given the vast choices of optimization software packages. We finally mention that transforming the problem of determining thresholds to mathematical programming problems was also discussed in [1] and [2] for impulse controls.

Appendix : Computation of functionals of expected hitting times for GBM

In this appendix we demonstrate, via the GBM, how to compute the expectation functionals involved in Section 6. Consider the GBM

$$X_t^x = x e^{\mu t + W_t}, \ t \ge 0, \ x > 0, \tag{A.1}$$

with $\mu \in \mathbb{R}$. We compute explicitly the expectation functionals defined in (6.2) and (6.3).

Lemma A.1

$$R(x,a) = \begin{cases} \left(\frac{a}{x}\right)^{\mu-\sqrt{\mu^2+2\rho}}, & \text{if } a \ge x\\ \left(\frac{a}{x}\right)^{\mu+\sqrt{\mu^2+2\rho}}, & \text{if } 0 < a < x, \end{cases}$$
$$\hat{R}(x,a,b) = R(x,a)R(a,b),$$
$$and \qquad \tilde{R}(x,a,b) = \left(\frac{a}{x}\right)^{\mu}\frac{sh\left[\log\frac{b}{x}\sqrt{2\alpha+\mu^2}\right]}{sh\left[\log\frac{b}{a}\sqrt{2\alpha+\mu^2}\right]}, & \text{if } a \le x \le b. \end{cases}$$

Proof. The first relation is evident from the fact that for the GBM (A.1), the hitting time τ_a^x is written as $\tau_a^x = \inf \{t \ge 0 : \mu t + W_t = \ln \left(\frac{a}{x}\right)\}$, and so

$$E\left[e^{-\rho\tau_a^x}\right] = e^{\mu \log\left(\frac{a}{x}\right) - \left|\log\frac{a}{x}\right| \sqrt{\mu^2 + 2\rho}}.$$

(see, e.g. p. 223, 2.0.1 in [5].) For the second one, we use the Lévy property of the Brownian motion, and write

$$E\left[e^{-\rho\tau_{ab}^{x}}\right] = \int_{0}^{\infty} E\left[e^{-\rho\tau_{ab}^{x}}|\tau_{a}^{x}=s\right]p_{\tau_{a}^{x}}(s)ds$$
$$= \int_{0}^{\infty} E\left[e^{-\rho(s+\tau_{b}^{a})}\right]p_{\tau_{a}^{x}}(s)ds,$$

where $p_{\tau_a^x}$ is the probability density function of τ_a^x . Therefore,

$$E[e^{-\rho\tau_{ab}^{x}}] = E[e^{-\rho\tau_{b}^{a}}] \int_{0}^{\infty} e^{-\rho s} p_{\tau_{a}^{x}}(s) ds$$
$$= E[e^{-\rho\tau_{b}^{a}}] E[e^{-\rho\tau_{a}^{x}}].$$

Finally, for the third relation, we refer to p. 233, 3.0.5, in [5].

Lemma A.2

$$F(f;x,a) = \begin{cases} \frac{1}{\sqrt{2\pi t}} \int_0^\infty e^{-\rho t} \int_0^a \frac{f(y)}{y} e^{\frac{-(\log \frac{y}{x} - \mu t)^2}{2t}} \left[1 - e^{\frac{-2\log \frac{a}{x} \cdot \log \frac{a}{y}}{t}} \right] dy dt, & \text{if } x \le a \\ \frac{1}{\sqrt{2\pi t}} \int_0^\infty e^{-\rho t} \int_a^\infty \frac{f(y)}{y} e^{\frac{-(\log \frac{y}{x} - \mu t)^2}{2t}} \left[1 - e^{\frac{-2\log \frac{a}{x} \cdot \log \frac{a}{y}}{t}} \right] dy dt, & \text{if } 0 < a < x, \end{cases}$$

$$\hat{F}(f;x,a,b) = R(x,a)F(f;x,b),$$

$$and \quad \tilde{F}(f;x,a,b) = \frac{1}{\sqrt{2\pi t}} \int_0^\infty e^{-\rho t} \int_a^b \frac{f(y)}{y} e^{\mu \log \frac{y}{x} - \frac{\mu^2 t}{2}} \\ \qquad \qquad \sum_{k=-\infty}^\infty \left(e^{-(\log \frac{y}{x} + 2k \log \frac{b}{a})^2/2t} - e^{-(\log \frac{y}{a} + 2k \log \frac{b}{a})^2/2t} \right) dy, \quad if \ a \le x \le b.$$

Proof. 1) (i) Consider first the case where $x \leq a$, and denote $W_t^{\mu} = \mu t + W_t$, $\overline{M}_t = \max_{0 \leq s \leq t} W_s^{\mu}$. Then for any $t \geq 0$ and $0 \leq y \leq a$:

$$\begin{split} P\left(\tau_{a}^{x} \geq t, X_{t}^{x} \geq y\right) &= P\left(\max_{0 \leq s \leq t} X_{s}^{x} \leq a, X_{t}^{x} \geq y\right) \\ &= P\left(\overline{M}_{t} \leq \log \frac{a}{x}, W_{t}^{\mu} \geq \log \frac{y}{x}\right) \\ &= P\left(W_{t}^{\mu} \geq \log \frac{y}{x}\right) - P\left(\overline{M}_{t} > \log \frac{a}{x}, W_{t}^{\mu} \geq \log \frac{y}{x}\right) \\ &= \frac{1}{\sqrt{2\pi t}} \int_{\log \frac{y}{x}}^{\infty} e^{-\frac{(z-\mu t)^{2}}{2t}} dz - \frac{1}{\sqrt{2\pi t}} \int_{\log \frac{y}{x}}^{\infty} e^{\mu z - \frac{\mu^{2} t}{2} - \frac{(|z-\log \frac{a}{x}| + \log \frac{a}{x})^{2}}{2t}} dz \\ &\quad (\text{see } p.198, 1.1.8 \text{ in } [5]) \\ &= \frac{1}{\sqrt{2\pi t}} \int_{\log \frac{y}{x}}^{\log \frac{a}{x}} e^{-\frac{(z-\mu t)^{2}}{2t}} \left[1 - e^{-\frac{2\log \frac{a}{x}(\log \frac{a}{x}-z)}{t}}\right] dz. \end{split}$$

Next, let $p_t(s, y)$ be the joint density function of (τ_a^x, X_t^x) . Then,

$$E \int_0^{\tau_a^x} e^{-\rho t} f(X_t^x) dt = \int_0^a \int_0^\infty \int_0^s e^{-\rho t} f(y) p_t(s, y) dt ds dy$$
$$= \int_0^a f(y) \int_0^\infty e^{-\rho t} \int_t^\infty p_t(s, y) dt ds dy$$
$$= \int_0^a f(y) \int_0^\infty e^{-\rho t} P\left(\tau_a^x \ge t, X_t^x \in dy\right) dt$$

However, by the earlier calculation

$$P(\tau_a^x \ge t, X_t \in dy) = -dP(\tau_a^x \ge t, X_t^x \ge y)$$

=
$$\frac{1}{\sqrt{2\pi t}y} e^{\frac{-(\log \frac{y}{x} - \mu t)^2}{2t}} \left[1 - e^{\frac{-2\log \frac{a}{x} \cdot \log \frac{a}{y}}{t}}\right] dy,$$

which proves the required result.

(ii) Suppose that 0 < a < x, and let $\underline{M}_t = \inf_{0 \le s \le t} W_s^{\mu}$. Then, for any $t \ge 0$ and $y \ge a$:

$$\begin{split} P\left(\tau_{a}^{x} \geq t, X_{t}^{x} \leq y\right) &= P\left(\underline{M}_{t} \geq \log\frac{a}{x}, W_{t}^{\mu} \leq \log\frac{y}{x}\right) \\ &= P\left(W_{t}^{\mu} \leq \log\frac{y}{x}\right) - P\left(\underline{M}_{t} < \log\frac{a}{x}, W_{t}^{\mu} \geq \log\frac{y}{x}\right) \\ &= \frac{1}{\sqrt{2\pi t}} \int_{-\infty}^{\log\frac{y}{x}} e^{-\frac{(z-\mu t)^{2}}{2t}} dz - \frac{1}{\sqrt{2\pi t}} \int_{-\infty}^{\log\frac{y}{x}} e^{\mu z - \frac{\mu^{2} t}{2} - \frac{(|z-\log\frac{a}{x}| - \log\frac{a}{x})^{2}}{2t}} dz \\ &\quad (\text{see } p.199, 1.2.8 \text{ in } [5]) \\ &= \frac{1}{\sqrt{2\pi t}} \int_{\log\frac{a}{x}}^{\log\frac{y}{x}} e^{-\frac{(z-\mu t)^{2}}{2t}} \left[1 - e^{-\frac{2\log\frac{a}{x}(\log\frac{a}{x}-z)}{t}}\right] dz, \end{split}$$

and we get the required result as in (i).

2) By using the Lévy property of the Brownian motion, we have

$$\begin{split} E\Big[\int_{\tau_a^x}^{\tau_{ab}^x} e^{-\rho t} f(X_t^x) dt\Big] &= \int_0^\infty E\left(\int_s^{s+\tau_b^a} e^{-\rho t} f(X_t^x) dt\right) p_{\tau_a^x}(s) ds \\ &= \int_0^\infty E\left(\int_0^{\tau_b^a} e^{-\rho(t+s)} f(X_t^x) dt\right) p_{\tau_a^x}(s) ds \\ &= \int_0^\infty e^{-\rho t} F(f;x,b) p_{\tau_a^x}(s) ds \\ &= E[e^{-\rho \tau_a^x}] F(f;x,b) = R(x,a) F(f;x,b). \end{split}$$

3) We now consider the third relation, and denote $\underline{M}_t = \min_{0 \le s \le t} W_s^{\mu}$. Then for any $t \ge 0$, $a \le y \le b$, and $a \le x \le b$:

$$\begin{split} P\left(\tau_a^x \wedge \tau_b^x \ge t, X_t^x \ge y\right) &= P\left(\max_{0 \le s \le t} X_s^x \le b, \min_{0 \le s \le t} X_s^x \ge a, X_t^x \ge y\right) \\ &= P\left(\overline{M}_t \le \log \frac{b}{x}, \underline{M}_t \ge \log \frac{a}{x}, W_t^\mu \ge \log \frac{y}{x}\right) \\ &= \frac{1}{\sqrt{2\pi t}} \int_{\log \frac{y}{x}}^{\infty} e^{\mu z - \frac{\mu^2 t}{2}} \\ &\qquad \qquad \sum_{k=-\infty}^{\infty} \left(e^{-(z+2k\log \frac{b}{a})^2/2t} - e^{-(z-\log \frac{a}{x}+2k\log \frac{b}{a})^2/2t}\right) dz; \end{split}$$

see p. 212, 1.15.8 in [5]. Next, let $q_t(s, y)$ be the joint density function of $(\tau_a^x \wedge \tau_b^x, X_t^x)$. Then, we have

$$E\left[\int_{0}^{\tau_{a}^{x} \wedge \tau_{b}^{x}} e^{-\rho t} f(X_{t}^{x}) dt\right] = \int_{a}^{b} \int_{0}^{\infty} \int_{0}^{s} e^{-\rho t} f(y) q_{t}(s, y) dt ds dy$$
$$= \int_{a}^{b} f(y) \int_{0}^{\infty} e^{-\rho t} \int_{t}^{\infty} q_{t}(s, y) ds dt dy$$
$$= \int_{a}^{b} f(y) \int_{0}^{\infty} e^{-\rho t} P\left(\tau_{a}^{x} \wedge \tau_{b}^{x} \ge t, X_{t}^{x} \in dy\right) dt$$

Now, by the earlier calculation

$$P(\tau_a^x \wedge \tau_b^x \ge t, X_t \in dy) = -dP(\tau_a^x \wedge \tau_b^x \ge t, X_t^x \ge y) \\ = \frac{1}{\sqrt{2\pi t y}} e^{\mu \log \frac{y}{x} - \frac{\mu^2 t}{2}} \\ \sum_{k=-\infty}^{\infty} \left(e^{-(\log \frac{y}{x} + 2k \log \frac{b}{a})^2/2t} - e^{-(\log \frac{y}{a} + 2k \log \frac{b}{a})^2/2t} \right),$$

from which we get the required result.

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