

OPTIMAL HITTING TIME AND PERPETUAL OPTION IN A NON-LÉVY MODEL: APPLICATION TO REAL OPTIONS

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Abstract

We study the perpetual American option characteristics in the case where the underlying dynamics involve a Brownian motion and a point process with a stochastic intensity. No assumption on the distribution of the jump size is made and we work with an arbitrary positive or negative jump. After proving the existence of an optimal stopping time for the problem and characterizing it as the hitting time of an optimal boundary, we provide closed-form formulae for the option value, as well as for the Laplace transform of the optimal stopping time. These results are then applied to the analysis of a real option problem when considering the impact of a fundamental and brutal change in the investment project environment. The consequences of this impact, that can seriously modify, positively or negatively, the project's future cash flows and therefore the investment decision, are illustrated by numerical examples.

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1. Introduction

In this paper, we consider a stochastic process S with dynamics involving a diffusion and a unique jump process. In such a framework, we aim at studying the problem

$$\operatorname{ess\,sup}_{\tau \in \Upsilon_t} \mathbb{E}(e^{-\mu(\tau-t)}(S_\tau - 1)^+ \mid \mathcal{F}_t),$$

where (\mathcal{F}_t) is the considered filtration, which will be made precise later, Υ_t denotes the set of \mathcal{F}_t -stopping times taking values in $(t, +\infty]$, $(\cdot)^+$ denotes the positive part, and μ is a positive real number. This issue comes from real options and, more precisely, from the investor's problem to finance a project, when a (unique) sudden and drastic change can occur in the environment. No assumption is made on the impact of this shock. As a consequence, even if our study includes credit risk issues, it cannot be reduced to this type of problem. In particular, we can also consider positive consequences of the drastic change on the project.

Real options studies are usually written in a continuous framework for the underlying dynamics (see the seminal papers by Brennan and Schwartz (1985), McDonald and Siegel (1986), and Dixit and Pindyck (1993)). But the existence of crises and shocks on investment

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markets generates discontinuities. Some research has been undertaken studying the impact of such instabilities on the investment decision process (see, for example, Gerber and Shiu (1998), Mordecki (2002), or Barrieu and Bellamy (2005)).

The question of the impact of a brutal change in the project environment has not been studied so far, to our knowledge. Such a fundamental change in the regulation can seriously impact, positively or negatively, the project's future cash flows and therefore the investment decision. Many situations can lead to a major brutal change in the project environment; for instance, a currency devaluation, an increase in interest rates, the appearance of brand-new technologies, or the entry of a new competing firm. Mathematically speaking, all these situations are similar and are equivalent to considering a single jump diffusion process.

This work aims at studying the impact of such a market perturbation on the decision taking. More precisely, the investor has a budget allocation problem. He has to determine whether the project is interesting and if so, to have some idea of the optimal investment time to allocate his capital among financial instruments with different maturities. Therefore, we need to specify the optimal benefit/cost ratio, the project investment value, and the optimal time to enter the project for the investor when he considers the project. This problem is similar to the study of a perpetual American call option written on the project ratio with strike 1. However, since this paper aims at analyzing a real option problem, we do not adopt a risk neutral logic, and the calculations are made with respect to the prior probability measure, without inducing any loss of generality.

In Section 2, we introduce the framework of study and present our model. The relationship between a real option problem and the pricing of a perpetual American option is also recalled. The main results on optimal stopping and perpetual options characteristics are given in Section 3. We obtain in this section closed-form formulae without any assumption on the distribution of the jump size and with an arbitrary positive or negative jump. Section 4 is devoted to the study of real option problems in a disrupted framework. We first illustrate the main results of this paper with different investment examples. Then we analyze the importance of the calibration of the model and study the consequences of a misspecification in a bearish environment. Finally, some proofs are presented in Appendix A.

2. The model

2.1. Framework

In this paper, we consider a given economic agent having the opportunity to invest in a particular project. The stochastic framework is described by a probability space (Ω, \mathcal{F}, P) where P is the prior probability measure representing the beliefs and anticipations of the agent.

The project is characterized at any time $t \geq 0$ by its profit/cost ratio S_t , with

$$\begin{aligned} dS_t &= S_{t-}(\alpha dt + \sigma dW_t + \Phi dD_t), \\ S_0 &= s_0, \end{aligned} \tag{1}$$

where the following statements hold.

- Let $(W_t; t \geq 0)$ be a P -Brownian motion and let α and σ be constants.
- The process $(D_t; t \geq 0)$ describes the potential shock of a drastic change in regulation on the project. It is defined as

$$D_t = \mathbf{1}_{\{t \geq T\}}, \tag{2}$$

where T is a random variable of exponential law with parameter λ and $\mathbf{1}_{\{t\}}$ denotes the indicator function. We denote the compensated martingale associated with $(D_t; t \geq 0)$ by $(M_t; t \geq 0)$, where

$$M_t = D_t - \lambda \int_0^t (1 - D_s) ds.$$

- The jump size Φ is a random variable. More precisely, Φ measures the impact of the perturbation on the profit/cost ratio.

Therefore, the solution of the stochastic differential equation (1) is

$$S_t = s_0(1 + \Phi)^{D_t} \exp\left[\left(\alpha - \frac{1}{2}\sigma^2\right)t + \sigma W_t\right].$$

We assume that the Brownian motion and the random variable T are independent. Moreover, the random variable Φ is taken independent of both the Brownian motion and the jump time process. Note that such an assumption is reasonable since the (regulation) shock is exogenous to the project we consider.

Assumption 1. *From now on, we make the following technical assumptions, ensuring that the problem we consider is well-defined and nondegenerate:*

- (i) $-1 < \underline{\Phi} \leq \Phi < \overline{\Phi}$, for two constants $\underline{\Phi}$ and $\overline{\Phi}$,
- (ii) $0 < s_0 < 1$,
- (iii) $\sigma > 0$,
- (iv) $\mu > \max(0, \alpha, \alpha + \lambda\overline{\Phi})$.

In Assumption 1(i), $\Phi > -1$ implies that the ratio S remains strictly positive, therefore the investment opportunity is never worthless. Considering a bounded Φ ensures that the positive impact of a drastic change in the environment cannot be infinite. Therefore, the situation we study is neither catastrophic nor explosive; Φ is also assumed to be not identically equal to zero, otherwise this study is reduced to the standard Black–Scholes framework.

Assumption 1(ii) states that the project is interesting ($s_0 > 0$) but also that the problem is about a ‘true’ decision since s_0 is (strictly) less than one; delaying the project realization is only relevant in the case where the profits/costs ratio is less than one.

Assumption 1(iii) maintains a Brownian source of randomness in the process dynamics. It is indeed reasonable to assume that the ratio profits/costs randomly evolves even without a major modification of the project environment. Assumption 1(iv) is an integrability condition.

The available information structure is characterized by the filtration $(\mathcal{F}_t; t \geq 0)$. At each instant of time t , the agent knows what is the current value of the ratio profit/cost related to the project, and therefore knows that $\sigma(S_s; 0 \leq s \leq t)$. This information includes the value of D_t ; indeed, at each instant t , from the observation of the underlying process S , the agent knows whether the shock has already occurred or not. As a consequence, the filtration \mathcal{F}_t is defined as

$$\mathcal{F}_t = \sigma(S_s, D_s; 0 \leq s \leq t).$$

Note that the approach defined by the framework we consider differs from those of the models in the literature related to both real options and American options. On the contrary to most of the existing literature (see for instance Gerber and Shiu (1998), Mordecki (2002), or Chesney and

Jeanblanc (2004)), the underlying process S cannot be written in terms of a Lévy process X_t as $S_0 \exp(X_t)$. Quite recently, Muroi (2002) considered a general framework involving a possible default for the underlying process of an American option. His approach is however different in the sense that it is focused on the numerical aspects and based upon partial differential equation methods. Moreover he does not provide closed-form characterizations for the American option.

2.2. Perpetual option and investment problem

The investor has neither any obligation to undertake it nor any time constraint to take his decision. Therefore, the investment decision problem is often brought down to the study of a perpetual American option with the profit/cost ratio S as underlying and 1 as striking level (see, for example, McDonald and Siegel (1986)). Solving the real option problem at any time t is equivalent to determine the value of the quantity

$$C_t = \text{ess sup}_{\tau \in \Upsilon_t} E(e^{-\mu(\tau-t)}(S_\tau - 1)^+ | \mathcal{F}_t),$$

where Υ_t denotes the set of \mathcal{F}_t -stopping times taking values in $(t, +\infty[$ and where μ is the discount rate, which can be different from the instantaneous risk-free rate and represents the agent’s preferences for the present. Also, E is the expectation with respect to the prior probability measure P .

Since the pair (D, S) , where D is defined in (2), is Markovian, we can rewrite C_t as

$$\begin{aligned} C_t &= \text{ess sup}_{\tau \in \Upsilon_t} E(e^{-\mu(\tau-t)}(S_\tau - 1)^+ | \mathcal{F}_t) \\ &= \text{ess sup}_{\tau \in \Upsilon_t} E(e^{-\mu(\tau-t)}(S_\tau - 1)^+ | S_t, D_t), \end{aligned}$$

or for all $(t, x, d) \in \mathbb{R}^+ \times \mathbb{R}^+ \times \{0, 1\}$, and adopting the natural notation, we obtain

$$C(t, x, d) = \text{ess sup}_{\tau \in \Upsilon_t} E(e^{-\mu(\tau-t)}(S_\tau - 1)^+ | S_t = x, D_t = d). \tag{3}$$

Note that the computations are made with respect to the prior probability measure P , corresponding to the agent’s beliefs. This does not create any loss of generality. If the underlying asset of the investment option is traded on financial market, a risk-neutral valuation formula can be easily obtained by considering a risk-neutral probability measure, as underlined in the following remark.

Remark 1. If a risk-neutral approach is adopted, then the only changes in the results come from both the modification of the discount rate μ in the instantaneous risk-free rate r , and a modification in the jump process intensity. More precisely, any equivalent martingale measure $P^{(\psi, \gamma)}$ is defined in terms of a pair of processes (ψ, γ) , satisfying the usual regularity conditions and being the solution to

$$\begin{aligned} \alpha + \lambda \Phi(1 - D_t) - r + \sigma \psi_t + \lambda \Phi \gamma_t &= 0, \\ \gamma_t &> -1. \end{aligned}$$

The Radon–Nikodym derivative

$$\left. \frac{dP^{(\psi, \gamma)}}{dP} \right|_{\mathcal{F}_t} = L_t^{(\psi, \gamma)}$$

is given by

$$dL_t^{(\psi,\gamma)} = L_t^{(\psi,\gamma)}(\psi_t dW_t + \gamma_t dM_t), \quad L_0^{(\psi,\gamma)} = 1.$$

Therefore, the intensity of $(D_t; t \geq 0)$ evaluated with respect to $P^{(\psi,\gamma)}$ is equal to $\lambda \int_0^t (1 + \gamma_s)(1 - D_s) ds$.

2.3. Notation

We now give a series of various notations which are used throughout the paper. In spite of being rather heavy, these notations are as simplified as possible and allow us to give explicit formulae for the different quantities characterizing the investment project.

(i) Let \mathcal{N} be the cumulative distribution of the Gaussian law, i.e.

$$\mathcal{N}(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-t^2/2} dt.$$

(ii) For $s_0 > 0, \nu > 0, L \geq 1, \varphi \geq 0$, and $a \in \mathbb{R}$, let $F_\varphi(s_0, \nu, L, a)$ be defined by

$$F_\varphi(s_0, \nu, L, a) = \int_0^{+\infty} \nu e^{-\nu t} \mathcal{N}\left(\frac{1}{\sigma\sqrt{t}} \ln\left(\frac{L}{s_0(1+\varphi)}\right) - a\sqrt{t}\right) dt - \left(\frac{L}{s_0}\right)^{2a/\sigma} \int_0^{+\infty} \nu e^{-\nu t} \mathcal{N}\left(-\frac{1}{\sigma\sqrt{t}} \ln\left(\frac{L(1+\varphi)}{s_0}\right) - a\sqrt{t}\right) dt, \quad (4)$$

and $\Delta F_{0,\varphi}(s_0, \nu, L, a)$ be defined by

$$\Delta F_{0,\varphi}(s_0, \nu, L, a) = F_\varphi(s_0, \nu, L, a) - F_0(s_0, \nu, L, a).$$

(iii) For $\nu > 0, \varphi \geq 0$, and $a \in \mathbb{R}$, let $f_\varphi(\nu, a)$ be defined by

$$f_\varphi(\nu, a) = \int_0^{+\infty} \nu e^{-\nu t} \frac{1}{\sqrt{2\pi t}} \exp\left(-\frac{1}{2}\left(\frac{1}{\sigma\sqrt{t}} \ln(1+\varphi) + a\sqrt{t}\right)^2\right) dt - a \int_0^{+\infty} \nu e^{-\nu t} \mathcal{N}\left(-\frac{1}{\sigma\sqrt{t}} \ln(1+\varphi) - a\sqrt{t}\right) dt, \quad (5)$$

and $\Delta f_{0,\varphi}(\nu, a)$ be defined by

$$\Delta f_{0,\varphi}(\nu, a) = f_\varphi(\nu, a) - f_0(\nu, a).$$

(iv) We denote by a_0, θ_0 , and ν_0 the following parameters

$$a_0 = \frac{1}{\sigma} \sqrt{\left(\alpha - \frac{\sigma^2}{2}\right)^2 + 2\mu\sigma^2}, \quad (6)$$

$$\theta_0 = \frac{1}{\sigma} \left(\frac{\sigma^2}{2} - \alpha + \sqrt{\left(\alpha - \frac{\sigma^2}{2}\right)^2 + 2\mu\sigma^2}\right), \quad (7)$$

$$\nu_0 = \frac{\alpha + \sigma^2/2}{\sigma}. \quad (8)$$

3. Main results on optimal stopping and perpetual options

This section is devoted to the study of the perpetual call option, described in Section 2.2. We first prove the existence of an optimal stopping time in (3), and characterize it as the hitting time of an optimal region. Then, we look at the different characteristics of the perpetual option. More precisely, we provide closed-form formulae for the Laplace transform of the optimal stopping time and for the real option value.

3.1. Existence and characterization of an optimal stopping time

In this section, we consider the question of the existence of an optimal stopping time for the problem (3). The following result holds true.

Theorem 1. *There exists an optimal stopping time for the problem (3). In other words, there exists $\tau_t^* \in \Upsilon_t$ such that*

$$C(t, x, d) = E(e^{-\mu(\tau_t^*-t)}(S_{\tau_t^*} - 1)^+ \mid S_t = x, D_t = d).$$

Moreover, τ_t^* is characterized as the first hitting time by the process S of a region $[L_t^{d,*}, +\infty)$, fully known at time t , i.e.

$$\tau_t^* = \inf\{s \geq t; S_s \geq L_t^{d,*}\}.$$

Here, two very different situations have to be studied separately, depending on whether the jump has already occurred or not (in other words, whether $d = 1$ or $d = 0$) at time t of the study. The arguments used in both cases are of a completely different nature.

3.1.1. *Case 1: $D_t = 1$.* In the situation where the jump has occurred before time t , the problem

$$C(t, x, 1) = \operatorname{ess\,sup}_{\tau \in \Upsilon_t} E(e^{-\mu(\tau-t)}(S_\tau - 1)^+ \mid S_t = x, D_t = 1) \tag{9}$$

is standard and well known, therefore we simply state the results and sketch the proof. More precisely,

Proposition 1. (i) *The optimal hitting time in the problem (9) exists and it is characterized as the first hitting time of a constant region $[L^{1,*}, +\infty)$, i.e.*

$$\begin{aligned} C(t, x, 1) &= \sup_{L \geq 1} E(e^{-\mu(\tau_L-t)}(S_{\tau_L} - 1)^+ \mid S_t = x, D_t = 1) \\ &= E(e^{-\mu(\tau_{L^{1,*}}-t)}(S_{\tau_{L^{1,*}}} - 1)^+ \mid S_t = x, D_t = 1), \end{aligned}$$

where $\tau_L = \inf\{t \geq 0; S_t \geq L\}$ is the first hitting time of $[L, +\infty)$.

(ii) *The optimal level $L^{1,*}$ is given by*

$$L^{1,*} = \frac{k_1}{k_1 - 1},$$

where k_1 is the unique positive real number such that $\psi(k_1) = \mu$, with ψ , the Lévy exponent of the Lévy process X , defined by

$$X_s = (\alpha - \frac{1}{2}\sigma^2)s + \sigma W_s.$$

Proof. Note that, on the set $\{t \geq T\}$, we necessarily have $\tau_L \geq T$ and, as a consequence, for any s, t such that $0 \leq T \leq t \leq s$ we have

$$S_s = x \exp\left(\left(\alpha - \frac{1}{2}\sigma^2\right)(s - t) + \sigma(W_s - W_t)\right).$$

Therefore, the proof of these results is similar to that of the standard Black–Scholes model (see, for example, Merton (1992, Section 8.8) or Karatzas and Shreve (1998, Section 2.6)), and we omit it.

3.1.2. *Case 2: $D_t = 0$.* The second situation where the jump has not already occurred at time t is not standard. The problem is of a double nature: first we have to prove the existence of an optimal stopping time for

$$C(t, x, 0) = \operatorname{ess\,sup}_{\tau \in \mathcal{Y}_t} \mathbb{E}(e^{-\mu(\tau-t)}(S_\tau - 1)^+ \mid S_t = x, D_t = 0) \tag{10}$$

and then we have to characterize it.

(i) Existence of an optimal stopping time. For any $x \in \mathbb{R}_*^+$, and for $D_t = 0$, we denote by $S^{t,x,0}$ the trajectory of the process S such that $S_t^{t,x,0} = x$. Hence, for any $s \geq t$ we obtain

$$S_s^{t,x,0} = x(1 + \Phi)^{D_s} \exp\left[\left(\alpha - \frac{1}{2}\sigma^2\right)(s - t) + \sigma(W_s - W_t)\right]. \tag{11}$$

The following result holds true.

Lemma 1. *The stopping time*

$$\tau_t^* = \inf\{s \geq t \mid C(s, S_s^{t,x,0}, 0) = (S_s^{t,x,0} - 1)^+\}$$

is \mathbb{P} -almost surely finite.

Proof. Recall the constants Φ and $\bar{\Phi}$ defined in Assumption 1. Now, the process S^Φ and the function C^Φ are defined as follows:

$$S_s^\Phi = x(1 + \Phi) \exp\left[\left(\alpha - \frac{1}{2}\sigma^2\right)(s - t) + \sigma(W_s - W_t)\right] \quad \text{for } s \geq t$$

and

$$C^\Phi(t, x) = \operatorname{ess\,sup}_{\tau \in \mathcal{Y}_t} \mathbb{E}(e^{-\mu(\tau-t)}(S_\tau^\Phi - 1)^+ \mid S_t^\Phi = x(1 + \Phi)).$$

The process $S^{\bar{\Phi}}$ and the function $C^{\bar{\Phi}}$ are defined in a similar way.

Because the inequalities $-1 < \underline{\Phi} \leq \Phi < \bar{\Phi}$ hold true, we can write

$$\begin{aligned} \tau_t^* &\leq \inf\{s \geq t \mid C^{\bar{\Phi}}(t, x) = (S_s^{t,x,0} - 1)^+\} \\ &\leq \inf\{s \geq t \mid C^{\bar{\Phi}}(t, x) = (S_s^\Phi - 1)^+\} \\ &\leq \inf\left\{s \geq t \mid C^{\bar{\Phi}}(t, x) = k\left(S_s^{\bar{\Phi}} - \frac{1}{k}\right)^+\right\}, \end{aligned}$$

where $k = (1 + \Phi)/(1 + \bar{\Phi})$. Now, from Assumption 1(iv) we have

$$\inf\left\{s \geq t \mid C^{\bar{\Phi}}(t, x) = k\left(S_s^{\bar{\Phi}} - \frac{1}{k}\right)^+\right\} < \infty \quad \mathbb{P} \text{ almost surely.}$$

Proposition 2. *There exists an optimal stopping time for the problem (10). In other words, there exists $\tau_t^* \in \Upsilon_t$ such that*

$$C(t, x, 0) = E(e^{-\mu(\tau_t^* - t)}(S_{\tau_t^*} - 1)^+ \mid S_t = x, D_t = 0). \tag{12}$$

Proof. From (11) and Assumptions 1(i) and 1(iv), we deduce that the optimal stopping theory can apply (see, for example, El Karoui (1981, Section 2.11) or Peskir and Shiryaev (2006)); therefore we conclude from Lemma 1 that the stopping time τ_t^* is optimal.

(ii) Characterization of the optimal stopping time. In this second part of this case, we want to characterize the optimal stopping time of the problem (10). Using the same type of arguments as in Jacka (1991), we get a characterization similar to the one obtained for the case $D_t = 1$.

Let \mathcal{Y} be the continuation region defined by

$$\mathcal{Y} = \{(t, x) \mid C(t, x, 0) > (x - 1)^+\},$$

and let \mathcal{Y}_t denote the section

$$\mathcal{Y}_t = \{x \mid (t, x) \in \mathcal{Y}\}.$$

We first give the characterization of the section \mathcal{Y}_t of the continuation region in terms of an optimal frontier.

Proposition 3. *For any $t > 0$, there exists $L_t^{0,*} > 1$ such that*

$$\mathcal{Y}_t = [0, L_t^{0,*}[.$$

Proof. From (12), the function $x \mapsto C(t, x, 0)$ is continuous, therefore it suffices to prove the following implication:

$$(x \in \mathcal{Y}_t \text{ and } 0 < y < x) \implies (y \in \mathcal{Y}_t).$$

Let us assume that $x \in \mathcal{Y}_t$. As a consequence, $C(t, x, 0) > (x - 1)^+$. Let τ_t^x be defined by

$$\tau_t^x = \inf\{s \geq t \mid (s, S_s^{t,x,0}) \notin \mathcal{Y}\}.$$

On the one hand, we have,

$$C(t, x, 0) = E(e^{-\mu(\tau_t^x - t)}(S_{\tau_t^x} - 1)^+ \mid S_t = x, D_t = 0) = E(e^{-\mu(\tau_t^x - t)}(S_{\tau_t^x}^{t,x,0} - 1)^+)$$

and, on the other hand, as τ_t^x is not optimal for $C(t, y, 0)$, we have

$$C(t, y, 0) \geq E(e^{-\mu(\tau_t^x - t)}(S_{\tau_t^x}^{t,x,0} - 1)^+).$$

Therefore,

$$C(t, x, 0) - C(t, y, 0) \leq E(e^{-\mu(\tau_t^x - t)}[(S_{\tau_t^x}^{t,x,0} - 1)^+ - (S_{\tau_t^x}^{t,y,0} - 1)^+]).$$

The assumption $0 < y < x$ together with (11) imply

$$\begin{aligned} & C(t, x, 0) - C(t, y, 0) \\ & \leq E(e^{-\mu(\tau_t^x - t)}(S_{\tau_t^x}^{t,x,0} - S_{\tau_t^x}^{t,y,0})) \\ & \leq (x - y) E\left(\exp\left((\alpha - \mu)(\tau_t^x - t) + \lambda \int_t^{\tau_t^x} \ln(1 + \Phi)(1 - D_s) ds\right) \mathcal{E}(\sigma W)_{\tau_t^x, t}^y\right), \end{aligned}$$

where $\mathcal{E}(\sigma W)_{\tau_t^x, t}^y = \exp(\sigma(W_{\tau_t^x}^y - W_t) - \frac{1}{2}\sigma^2(\tau_t^x - t))$.

Let \tilde{P} be the probability measure, equivalent to P , defined by

$$\frac{d\tilde{P}}{dP} \Big|_{\mathcal{F}_t} = \exp\left(\sigma W_t - \frac{1}{2}\sigma^2 t\right).$$

Now,

$$C(t, x, 0) - C(t, y, 0) \leq (x - y) \tilde{E}\left(\exp\left((\alpha - \mu)(\tau_t^x - t) + \lambda \int_t^{\tau_t^x} \ln(1 + \Phi)(1 - D_s) ds\right)\right).$$

From Assumption 1(iv), we have

$$\tilde{E}\left(\exp\left((\alpha - \mu)(\tau_t^x - t) + \lambda \int_t^{\tau_t^x} \ln(1 + \Phi)(1 - D_s) ds\right)\right) \leq 1$$

and

$$C(t, x, 0) - C(t, y, 0) \leq (x - y),$$

or

$$C(t, y, 0) \geq C(t, x, 0) - (x - y) > (x - 1)^+ - (x - y) > y - 1.$$

Let $\xi = \inf\{u \geq t \mid S_u^{t,y,0} \geq 2\}$, then

$$\begin{aligned} C(t, y, 0) &\geq E(e^{-\mu(\xi-t)}(S_\xi^{t,y,0} - 1)^+) \\ &\geq E(e^{-\mu(\xi-t)}) \\ &> 0. \end{aligned}$$

Therefore, for any $(t, y) \in \mathbb{R}^+ \times \mathbb{R}_*^+$, $C(t, y, 0) > 0$. Finally, $C(t, y, 0) > (y - 1)^+$. Hence, the result follows.

We can now characterize the optimal stopping time as the hitting time of a region.

Corollary 1. *The optimal stopping time τ_t^* of the problem (10) is characterized by*

$$\tau_t^* = \inf\{s \geq t \mid S_s^{t,x,0} \geq L_t^{0,*}\},$$

where $L_t^{0,*}$ is defined in Proposition 3.

Proof. The proof is an immediate consequence of Proposition 3.

3.2. Perpetual option characteristics

We are now interested in finding the perpetual option characteristics, in particular its value. As previously, the situations when $D_t = 1$ and $D_t = 0$ are very different. Therefore, we will look at them separately.

3.2.1. *Case 1: $D_t = 1$.* As previously mentioned, when the jump has already occurred before time t , the results are standard in this case.

Proposition 4. (i) *The Laplace transform of the optimal time is*

$$E(e^{-\mu(\tau_{L^{1,*}} - t)} \mid S_t = x, D_t = 1) = \left(\frac{x}{L^{1,*}}\right)^{L^{1,*}/L^{1,*}-1},$$

(ii) The price of the real option is given by

$$C(t, x, 1) = (L^{1,*} - 1) \left(\frac{L^{1,*}}{x} \right)^{-k_1},$$

where the optimal level $L^{1,*}$ is given by Proposition 1.

Proof. These results are similar to those of the standard Black–Scholes framework and can be found, for example, in Merton (1992) or Karatzas and Shreve (1998).

Note that the formulae do not apparently depend on the jump size Φ . This comes from the fact that after the jump, the jump size does not play any role apart from the shift in the ‘initial condition’, which becomes $s_0(1 + \Phi)$ instead of s_0 . This dependency is somehow hidden in the value x of the process at time t .

3.2.2. *Case 2:* $D_t = 0$. As previously, the situation where the jump has not occurred yet at time t , is not standard. We now want to study the characteristics of the perpetual option

$$C(t, x, 0) = \text{ess sup}_{\tau \in \mathcal{T}_t} E(e^{-\mu(\tau-t)} (S_\tau - 1)^+ \mid S_t = x, D_t = 0),$$

which can also be written, using Corollary 1, as

$$C(t, x, 0) = \text{ess sup}_{L_t^0 \geq 1} E(e^{-\mu(\tau_{L_t^0} - t)} (S_{\tau_{L_t^0}} - 1)^+ \mid S_t = x, D_t = 0).$$

For the sake of simplicity in the notation and computation but without loss of generality, we consider this problem at time $t = 0$ (obviously, $D_0 = 0$). Therefore, we look at

$$C(0, s_0, 0) = \text{ess sup}_{L^0 \geq 1} E(e^{-\mu\tau_{L^0}} (S_{\tau_{L^0}} - 1)^+ \mid S_0 = s_0). \tag{13}$$

We now present the main result of this paper, giving the perpetual option characteristics in a general framework where we do not know a priori the sign of the jump size Φ . The following two theorems provide these explicit formulae at time zero. In particular, the optimal frontier and a characterization of the optimal hitting time are now given in Theorem 2.

Theorem 2. (i) The optimal frontier of the problem (13) is given by

$$L_\Phi^{0,*} = \frac{\eta(\Phi)}{\eta(\Phi) - 1 + \Gamma(\Phi)},$$

where

$$\begin{aligned} \eta(\Phi) &= \frac{2}{\sigma} f_0(\lambda, a_0) + \theta_0 \sigma \\ &\quad - \frac{2}{\sigma} E(\mathbf{1}_{\{\Phi < 0\}} (1 + \Phi)^{\theta_0/\sigma}) f_0(\lambda, a_0) \\ &\quad + \frac{2}{\sigma} E \left[\mathbf{1}_{\{\Phi > 0\}} \left[\frac{\lambda}{\lambda + \mu} \Delta f_{0,\Phi}(\lambda + \mu, a_0 - \theta_0) - (1 + \Phi)^{\theta_0/\sigma} f_\Phi(\lambda, a_0) \right] \right] \end{aligned}$$

and

$$\begin{aligned} \Gamma(\Phi) &= \frac{2}{\sigma} E \left[\mathbf{1}_{\{\Phi > 0\}} \left(\frac{\lambda(1 + \Phi)}{\lambda + \mu - \alpha} \Delta f_{0,\Phi}(\lambda + \mu - \alpha, v_0) - \frac{\lambda}{\lambda + \mu} \Delta f_{0,\Phi}(\lambda + \mu, a_0 - \theta_0) \right) \right]. \end{aligned}$$

(ii) For every s_0 such that $0 < s_0 < L_{\Phi}^{0,*}$, the optimal hitting time is characterized by its Laplace transform

$$E(\exp(-\mu\tau_{L_{\Phi}^{0,*}(\Phi)})) = \left(\frac{s_0}{L_{\Phi}^{0,*}}\right)^{\theta_0/\sigma} (1 + K(\Phi)) + \Theta(\Phi),$$

where

$$\begin{aligned} K(\Phi) &= -F_0(s_0, \lambda, L_{\Phi}^{0,*}, a_0) + F_0(s_0, \lambda, L_{\Phi}^{0,*}, a_0) E[\mathbf{1}_{\{\Phi < 0\}}(1 + \Phi)^{\theta_0/\sigma}] \\ &\quad + E[\mathbf{1}_{\{\Phi > 0\}}(1 + \Phi)^{\theta_0/\sigma} F_{\Phi}(s_0, \lambda, L_{\Phi}^{0,*}, a_0)], \\ \Theta(\Phi) &= -\frac{\lambda}{\lambda + \mu} E[\mathbf{1}_{\{\Phi > 0\}} \Delta F_{0,\Phi}(s_0, \lambda + \mu, L_{\Phi}^{0,*}, a_0 - \theta_0)]; \end{aligned}$$

see Section 2.3 for the notation.

Proof. Let Φ be a constant denoted by φ . To prove this result, we first consider the following two cases: φ is positive (i.e. a bullish environment) or φ is negative (i.e. a bearish environment). Finally, the general result is obtained by means of conditional expectation, as it is enough to consider the case where the jump size is a constant, from the independence of the random variable Φ and the filtration generated by the Brownian motion and the jump process.

(a) The bullish environment. Assuming the random variable Φ to be a positive constant φ , we determine the optimal threshold on which the agent bases his decision. We proceed in two steps to obtain $L_{\Phi}^{0,*}$ in a bullish framework (with $\varphi > 0$).

Step 1. The functions F_{φ} and f_{φ} , defined by (4) and (5) are linked by the following relationship:

$$\text{for all } \lambda > 0 \text{ and for all } a \in \mathbb{R}, \quad \left. \frac{\partial}{\partial L} F_{\varphi}(s_0, \lambda, L, a) \right|_{s_0} = L = \frac{2}{\sigma L} f_{\varphi}(\lambda, a).$$

Moreover, the function F_{φ} satisfies $F_{\varphi}(L, \lambda, L, a) = 0$.

Step 2. We denote by c the following function:

$$c(s_0, L) = E[e^{-\mu\tau_L}(S_{\tau_L} - 1)].$$

Then from Lemma 2 we deduce the following:

$$\begin{aligned} \left. \frac{\partial}{\partial L} c(s_0, L) \right|_{s_0} &= L \\ &= \frac{2}{\sigma L} (1 + \varphi)^{\theta_0/\sigma} (L - 1) f_{\varphi}(\lambda, a_0) + 1 - \frac{\theta_0}{\sigma L} (L - 1) - \frac{2}{\sigma L} (L - 1) f_0(\lambda, a_0) \\ &\quad - \frac{2\lambda(1 + \varphi)}{\sigma(\lambda + \mu - \alpha)} \Delta f_{0,\varphi}(\lambda + \mu - \alpha, v_0) + \frac{2\lambda}{\sigma L(\lambda + \mu)} \Delta f_{0,\varphi}(\lambda + \mu, a_0 - \theta_0). \end{aligned}$$

The unique solution of

$$\left. \frac{\partial}{\partial L} c(s_0, L) \right|_{s_0} = L = 0$$

is

$$L = L_{\varphi}^{0,*} = \frac{\eta^+(\varphi)}{\eta^+(\varphi) - 1 + \Gamma^+(\varphi)},$$

where

$$\eta^+(\varphi) = \frac{2}{\sigma} f_0(\lambda, a_0) + \theta_0 \sigma + \frac{2}{\sigma} \left[\frac{\lambda}{\lambda + \mu} \Delta f_{0,\varphi}(\lambda + \mu, a_0 - \theta_0) - (1 + \varphi)^{\theta_0/\sigma} f_\varphi(\lambda, a_0) \right],$$

$$\Gamma^+(\varphi) = \frac{2}{\sigma} \left[\frac{\lambda(1 + \Phi)}{\lambda + \mu - \alpha} \Delta f_{0,\varphi}(\lambda + \mu - \alpha, v_0) - \frac{\lambda}{\lambda + \mu} \Delta f_{0,\varphi}(\lambda + \mu, a_0 - \theta_0) \right].$$

(b) The bearish environment. In the case where the constant φ is nonpositive, the optimal frontier is given by

$$L_\varphi^{0,*} = \frac{\eta^-(\varphi)}{\eta^-(\varphi) - 1},$$

where

$$\eta^-(\varphi) = \frac{2}{\sigma} \left[(1 - (1 + \varphi)^{\theta_0/\sigma}) f_0(\lambda, a_0) + \frac{\theta_0}{2} \right].$$

The proof for this case is similar to that of the previous case and the result follows directly from Lemma 3.

Theorem 2(i) is then a straightforward consequence of parts (a) and (b) of the proof and of the independence of the random variable Φ with respect to the Brownian motion and the jump process. Theorem 2(ii) can be deduced from this independence, Lemma 2, Lemma 3, and the value of the optimal threshold given in Theorem 2(i).

The following theorem gives the perpetual option value.

Theorem 3. For every s_0 such that $0 < s_0 < L_\Phi^{0,*}$, the perpetual option value at time $t = 0$ is

$$C(0, s_0, 0) = (L_\Phi^{0,*} - 1) \left(\frac{s_0}{L_\Phi^{0,*}} \right)^{\theta_0/\sigma} (1 + K(\Phi)) + J(\Phi),$$

where $K(\cdot)$ is defined in Theorem 2 and

$$J(\Phi) = - \frac{\lambda s_0}{\lambda + \mu - \alpha} \mathbb{E}[\mathbf{1}_{\{\Phi > 0\}} (1 + \Phi) \Delta F_{0,\Phi}(s_0, \lambda + \mu - \alpha, L_\Phi^{0,*}, v_0)];$$

see Section 2.3 for the notation.

Proof. The proof follows directly from Theorem 2, Lemma 2, and Lemma 3.

3.3. Comments

The model presented here, in particular the situation $D_t = 0$, is very different from the existing literature, as it corresponds to a non-Lévy model. The results are therefore different. Note, however, that if $\Phi \equiv 0$ P almost surely, then all the formulae of this study are similar to those of the ‘Black–Scholes’ model. The ‘interesting’ situation is when D_t is different from 1. Otherwise, the formulae are similar to a standard Black–Scholes-type framework, where no possible jump of the underlying process is introduced (as it is equivalent to have a shift in the initial condition of our problem). Therefore, we focus on the nonstandard results, when $D_t = 0$.

In this paper, Theorem 2 gives an explicit representation of the optimal boundary without any particular assumption either on the sign of the jump Φ or on its distribution. Whether Φ is positive or negative has a great impact on the structure of the different results since it shapes the optimal frontier differently. In fact, both situations are mathematically very different, as it can

be seen when considering the two particular cases where the random variable Φ has a constant negative versus positive sign. The main reason is that when the jump is negative, the process S necessarily crosses the given frontier in a continuous way. However, this property does not hold any more if the jump is positive. Therefore, the mathematics involved and the results we obtain are dissimilar. Moreover, the results we obtain are strongly different from those of the Lévy model.

When the jump is negative, even if the optimal frontier is still of a standard shape, i.e.

$$L_{\Phi}^{0,*} = \frac{\eta^{-}(\Phi)}{\eta^{-}(\Phi) - 1},$$

both the Laplace transform and the value of the investment opportunity are different. The standard equations for the Laplace transform and the price are perturbed by a multiplicative factor K , this factor K being identical to the one in the bullish environment, we have

$$\begin{aligned} E(\exp(-\mu\tau_{L_0^{0,*}(\Phi)})) &= \left(\frac{s_0}{L_{\Phi}^{0,*}}\right)^{L_{\Phi}^{0,*}/(L_{\Phi}^{0,*}-1)} (1 + K(\Phi)), \\ C(0, s_0, 0) &= (L_{\Phi}^{0,*} - 1) \left(\frac{s_0}{L_{\Phi}^{0,*}}\right)^{L_{\Phi}^{0,*}/(L_{\Phi}^{0,*}-1)} (1 + K(\Phi)). \end{aligned}$$

This multiplicative factor comes from the memory of the process used to model the profit and cost ratio. Obviously, in the limit situation where there is no possible jump ($\Phi \equiv 0$), the results coincide with those of the standard Black–Scholes-type framework as $K(0) = 0$.

When considering a positive jump, we obtain explicit formulae for the three characteristics of the investment opportunity. The results we obtain are fundamentally different from those of Lévy models. In particular, the optimal frontier cannot be written under the previous standard form (if we exclude the particular limit case where $\Phi = 0$). Indeed, an additional term $\Gamma^{+}(\Phi)$ is present in the denominator of the optimal frontier, i.e.

$$L_{\Phi}^{0,*} = \frac{\eta^{+}(\Phi)}{\eta^{+}(\Phi) - 1 + \Gamma^{+}(\Phi)}.$$

Both the Laplace transform and the investment value differ from the standard case by two different effects. There is first a common multiplicative factor K and also two additive factors J' (for the Laplace transform) and J (for the value of the investment), i.e.

$$\begin{aligned} E(\exp(-\mu\tau_{L_0^{0,*}(\Phi)})) &= \Theta(\Phi) + \left(\frac{s_0}{L_{\Phi}^{0,*}}\right)^{L_{\Phi}^{0,*}/(L_{\Phi}^{0,*}-1)} (1 + K(\Phi)), \\ C(0, s_0, 0) &= J(\Phi) + (L_{\Phi}^{0,*} - 1) \left(\frac{s_0}{L_{\Phi}^{0,*}}\right)^{L_{\Phi}^{0,*}/(L_{\Phi}^{0,*}-1)} (1 + K(\Phi)). \end{aligned}$$

These differences can be explained as follows. As for the bearish case, the factor K comes from the memory of the process used to model the profit and cost ratio, whereas the additive factors J' and J translate the fact that in this case the optimal boundary $L_{\Phi}^{0,*}$ is not necessarily crossed in a continuous way.

4. Application to a real option problem

In this section, we look more closely at the investment problem (i.e. the value of the investment project and optimal hitting time) when the investor has no particular knowledge about the type of impact a new regulation, i.e. a sudden shock, may have on the project itself. Namely, we do not know whether the sign of the jump will be positive or negative. This may be the case for investment projects related to new technologies when the prospects are uncertain. In this case, the related market witnesses a rapid evolution. New regulations can be adopted, modifying its structure. They can generate either a positive impact when improving the security and the transparency (bullish environment) or a negative impact when increasing its complexity (bearish environment). New competitors can also enter the market, attracted by these new prospects, and actual competitors can default, victims of a lack of experience and speculation. Finally, we may think of new techniques that can be patented either by the considered firm or by competitors.

In this section, we focus on the situation at time $t = 0$ without any loss of generality. In fact, if a new regulation is expected, or if some disruptive factor for the economic environment is likely to occur, we may think that the investor will delay his investment decision and wait for the actual occurrence of the shock to evaluate the impact on the investment project and to finally take his decision. This delay seems to be all the more important since the potential impact (positive or negative) is unknown. In reality, however, the investor does not necessarily have the opportunity of delaying his decisions, and often has to decide as soon as the investment project appears (i.e. at time $t = 0$). This pressure may be due to several reasons. In the first place, the investor could face some legal constraints, for instance in the search for concessions or licences (for example, the Universal Mobile Telecommunications Systems, where the investors have to take their decisions before a fixed deadline). The investor can also face some market and competition constraints, where the postponement of a decision can give his competitors the opportunity of taking a stand and, as a consequence, invalidate any possibility of investing in the strategic business field for the investor. So considering the study at time $t = 0$ appears to be relevant for the investment problem we analyze.

In the following, we look at a real option problem in a disrupted framework, as previously described. We illustrate the results of Section 3 and provide numerical applications. In such a context, one of the most important questions is certainly related to the calibration of the model. This is all the more important since the shock induces a drastic decrease in the project cash flows. Therefore, we conclude this study by looking at the impact of a model misspecification in a bearish environment, i.e. when the impact of a shock is to be negative.

4.1. Illustration

We now illustrate the main results of this paper (Theorems 2 and 3) through numerical examples. More precisely, we consider an agent facing an investment problem at time zero, in a potentially disturbed environment. We focus especially on the sensitivity with respect to the jump size of both the Laplace transform of the hitting time and the investment value. The set of parameters we consider is

$$s_0 = 0.8, \quad \sigma = 0.2, \quad \alpha = 0.05, \quad \mu = 0.15.$$

In Table 1 we give some values of the optimal profit and cost ratio for different values of the jump intensity λ and the jump size φ . The optimal ratio is monotonic with respect to the jump size φ . Moreover, the optimal ratio is a nonincreasing function of the jump intensity λ for negative values of φ , whereas it is a nondecreasing function of λ for positive values of the

TABLE 1: The optimal profit–cost ratio.

φ	$\lambda = 0.25$	$\lambda = 0.5$	$\lambda = 1$	$\lambda = 2$
−0.99	1.407 617	1.374 67	1.364 417	1.367 871
−0.8	1.415 605	1.382 496	1.372 175	1.375 653
−0.6	1.443 986	1.410 464	1.399 953	1.403 498
−0.5	1.468 846	1.435 171	1.424 556	1.428 139
−0.4	1.503 957	1.470 402	1.459 746	1.463 347
−0.3	1.553 697	1.520 998	1.510 504	1.514 057
−0.2	1.625 893	1.595 907	1.586 132	1.589 45
−0.1	1.735 855	1.713 473	1.705 999	1.708 546
−0.05	1.814 285	1.800 001	1.795 146	1.796 805
0	1.917 891	1.917 891	1.917 891	1.917 891
0.05	2.061 146	2.098 451	2.130 82	2.165 905
0.1	2.259 259	2.387 87	2.545 456	
0.15	2.530 455	2.851 521		
0.2	2.906 61	3.639 258		
0.25	3.446 203			
0.3	4.267 252			
0.35	5.645 627			
0.4	8.406 155			

TABLE 2: The Laplace transform of the optimal time to invest.

φ	$\lambda = 0.25$	$\lambda = 0.5$	$\lambda = 1$	$\lambda = 2$
−0.99	0.117 852	0.066 776	0.025 591	0.005 82
−0.9	0.118 797	0.068 404	0.027 797	0.008 327
−0.8	0.121 859	0.073 688	0.034 962	0.016 465
−0.7	0.127 039	0.082 68	0.047 158	0.030 297
−0.6	0.134 14	0.095 13	0.064 061	0.049 423
−0.5	0.142 747	0.110 47	0.084 925	0.072 96
−0.4	0.152 147	0.127 691	0.108 44	0.099 398
−0.3	0.161 184	0.145 118	0.132 457	0.126 331
−0.2	0.167 994	0.159 997	0.153 506	0.149 993
−0.1	0.169 553	0.167 775	0.165 981	0.164 525
−0.05	0.166 918	0.166 801	0.166 368	0.165 701
0	0.160 903	0.160 903	0.160 903	0.160 903
0.05	0.150 273	0.146 763	0.142 857	0.138 179
0.1	0.135 036	0.123 248	0.108 60	
0.15	0.116 245	0.093 441		
0.2	0.095 067	0.061 498		
0.25	0.072 806			
0.3	0.050 901			
0.35	0.030 943			
0.4	0.014 657			

TABLE 3: The investment value.

φ	$\lambda = 0.25$	$\lambda = 0.5$	$\lambda = 1$	$\lambda = 2$
-0.99	0.048 038	0.025 019	0.009 326	0.002 141
-0.9	0.048 642	0.025 752	0.010 179	0.003 078
-0.8	0.050 645	0.028 185	0.013 012	0.006 185
-0.7	0.054 217	0.032 532	0.018 065	0.011 712
-0.6	0.059 556	0.039 048	0.025 622	0.019 942
-0.5	0.066 926	0.048 073	0.036 055	0.031 237
-0.4	0.076 676	0.060 066	0.049 855	0.046 056
-0.3	0.089 247	0.075 606	0.067 620	0.064 941
-0.2	0.105 146	0.095 343	0.089 975	0.088 414
-0.1	0.124 766	0.119 703	0.117 183	0.116 573
-0.05	0.135 919	0.133 441	0.132 287	0.132 032
0	0.147 691	0.147 691	0.147 691	0.147 691
0.05	0.159 882	0.161 456	0.161 618	0.161 111
0.1	0.171 207	0.171 558	0.167 923	
0.15	0.179 599	0.173 483		
0.2	0.183 058	0.162 57		
0.25	0.179 627			
0.3	0.167 352			
0.35	0.144 303			
0.4	0.108 75			

jump size. Note also that whatever the jump intensity is, this ratio tends to

$$\frac{\theta_0}{(\theta_0 - \sigma)}$$

when the jump size tends to zero. This value corresponds to the optimal ratio when the model uncertainty is simply driven by a Brownian motion.

Table 2 provides the variation of the Laplace transform $E(\exp(-\mu\tau_{L^0, \ast}))$ and those of the optimal investment value at time zero with respect to the jump size φ for different values of λ .

Tables 2 and 3 provide highly dissymmetrical values with respect to the sign of the jump. Neither the Laplace transform nor the value of the investment are monotonic with respect to φ . If the jump size tends to zero, then the Laplace transform tends to $(s_0(\theta_0 - \sigma)/\theta_0)^{\theta_0/\sigma}$, and it coincides with the value of the Laplace transform of the optimal hitting time in a Brownian model. The same result holds for the investment value.

4.2. Misspecification impact in a bearish environment

Assume now that the agent who does not know the (negative) random jump size Φ , can however estimate it by its expectation $E(\Phi)$. The question is then if he takes his investment decision by considering only the expected value, how will his decision be impacted, in particular his investment timing: will he invest too early or too late in the project? Note that the value of the investment project is obviously reduced since $C(0, s_0, 0)$ is determined by using an optimal frontier satisfying the supremum, i.e.

$$C(0, s_0, 0) = \text{ess sup}_{L^0 \geq 1} E(e^{-\mu\tau_{L^0}} (S_{\tau_{L^0}} - 1)^+ | S_0 = s_0).$$

Any other frontier (in particular the level obtained by considering the expected value of the jump size instead of the random variable) is suboptimal, and the supremum is not reached for this particular value.

Let us first focus on the optimal level of benefit and cost ratio when the agent considers the expected value as the estimate for the jump size. In the following result we establish that such a misspecification leads the investor to undervalue the optimal level of the benefit/cost ratio.

Proposition 5. *Using previous notation, we have*

$$L_{\Phi}^{0,*} \geq L_{E(\Phi)}^{0,*}.$$

Proof. Let g be the function defined by

$$g(\theta) = \frac{1}{2}\theta^2 + \frac{\theta}{\sigma} \left(\alpha - \frac{\sigma^2}{2} \right) - \mu.$$

The inequalities $0 < \theta_0$ and $0 \leq \alpha < \mu$ and the equalities $g(\theta_0) = 0$ and $g(\sigma) = \alpha - \mu$ imply that $\theta_0/\sigma > 1$. Hence, by Jensen's inequality, we deduce

$$E((1 + \Phi)^{\theta_0/\sigma}) \geq (1 + E(\Phi))^{\theta_0/\sigma}.$$

On the other hand, we have

$$f_0(\lambda, a) \geq 0 \quad \text{for all } \lambda > 0 \text{ and for all } a \in \mathbb{R}.$$

So, from Theorem 2, we may conclude that

$$L_{\Phi}^{0,*} \geq L_{E(\Phi)}^{0,*}.$$

Therefore, assuming that the agent chooses his investment time according to $L_{E(\Phi)}^{0,*}$, he will enter the project as soon as the benefit/cost ratio reaches this threshold (instead of using the optimal threshold $L_{\Phi}^{0,*}$). In that sense, it can be said that the model misspecification leads them to undertake the project too early. More precisely, the agent has some beliefs regarding the dynamics of the random benefit/cost ratio. He knows perfectly its dynamics when his information is perfect or has simply an estimation when his access to the information is limited. Given his beliefs, the agent is able to explicitly determine the optimal ratio and therefore his investment strategy using this ratio as the threshold to take his decision. He has no better way to decide when it is optimal to invest, as the knowledge of the dynamics of S simply gives him the Laplace transform of the optimal investment time and therefore simply a heuristic determination of an average investment time (see, for example, Barrieu and Bellamy (2005)).

5. Conclusion

In this paper, We provide closed-form formulae for the different characteristics of a perpetual option in a framework where the uncertainty is brought by a Brownian motion and a point process with stochastic intensity. An application of these results to a real option problem is then presented. Assuming that a unique shock occurs in the context of an investment project, we study the investment problem when the project's environment can be subject to a brutal and fundamental change (for instance, in its regulation or in its competitors' strengths and numbers). Mathematically speaking, all these situations are similar and are equivalent to

considering a single jump diffusion process. Our aim has been to study the impact of such a market perturbation on the decision. More precisely, assuming that the investor has a budget allocation problem at time zero, he has to determine whether the project is interesting and if so, to have some idea of the optimal investment time as to allocate his capital among financial instruments with different maturities. In this framework, the investor determines his strategy at time zero by characterizing the optimal profit/cost ratio, and then can anticipate the optimal time to invest in the project by looking at the Laplace transform at time zero of the first hitting time of this optimal ratio. We have obtained closed-form formulae for the optimal benefit/cost ratio, the investment value, and the optimal time to enter the project, without any assumption on the distribution of the jump size and with arbitrary positive or negative jump.

The framework of the investment problem studied in this paper can be extended. For example, it is also possible to consider an investment environment that could be altered potentially by different shocks. In this case, the profit/cost ratio is then written in the form

$$dS_t = S_{t-} \left(\alpha dt + \sigma dW_t + \sum_{i=1}^{i=k} \Phi_i dD_{it} \right),$$

where $D_{it} = 1_{t \geq T_i}$ and $T_1, T_2 - T_3, \dots, T_k - T_{k-1}$ is an increasing sequence of independent random variables with exponential law, the parameter of $T_i - T_{i-1}$ being λ_i . Such an extension seems rather natural as it leads to a framework allowing for different impacts of different natures on the investment field. In this sense, the uncertainty prevailing in investment problems could be represented.

Appendix A

Lemma 2. *Assuming that the jump size is a positive constant φ , the following equalities hold:*

- (i) $E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L=T\}}) = -\frac{\lambda}{\lambda + \mu} \Delta F_{0,\varphi}(s_0, \lambda + \mu, L, a_0 - \theta_0),$
- (ii) $E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L>T\}}) = \left(\frac{(1 + \varphi)s_0}{L}\right)^{\theta_0/\sigma} F_\varphi(s_0, \lambda, L, a_0),$
- (iii) $E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L<T\}}) = \left(\frac{s_0}{L}\right)^{\theta_0/\sigma} [1 - F_0(s_0, \lambda, L, a_0)],$
- (iv) $c(s_0, L) = E[e^{-\mu\tau_L} (S_\tau - 1)]$
 $= (L - 1) \left(\frac{s_0}{L}\right)^{\theta_0/\sigma} [1 - F_0(s_0, \lambda, L, a_0)]$
 $+ (L - 1) \left(\frac{(1 + \varphi)s_0}{L}\right)^{\theta_0/\sigma} F_\varphi(s_0, \lambda, L, a_0)$
 $- \frac{\lambda s_0(1 + \varphi)}{\lambda + \mu - \alpha} \Delta F_{0,\varphi}(s_0, \lambda + \mu - \alpha, L, v_0)$
 $+ \frac{\lambda}{\lambda + \mu} \Delta F_{0,\varphi}(s_0, \lambda + \mu, L, a_0 - \theta_0).$

Proof. (i) We have

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L=T\}}) = \frac{\lambda}{\lambda + \mu} E^{\psi_1}(\mathbf{1}_{\{\tau_L=T\}}),$$

where $\psi_1 = \mu/\lambda$ and where P^{ψ_1} is the P-equivalent probability measure specified by its Radon–Nikodym derivative

$$\left. \frac{dP^{\psi_1}}{dP} \right|_{\mathcal{F}_t} = \mathcal{E}(\psi_1 M)_t = \exp\left(\ln(1 + \psi_1)D_t - \lambda\psi_1 \int_0^t (1 - D_s) ds\right),$$

and E^{ψ_1} is the P^{ψ_1} -expectation. Hence, we obtain

$$E^{\psi_1}(\mathbf{1}_{\{\tau_L=T\}}) = \int_0^{+\infty} (\lambda + \mu) e^{-(\lambda+\mu)t} \left[\mathcal{N}\left(\frac{y - u_0t}{\sqrt{t}}\right) - e^{2u_0y} \mathcal{N}\left(\frac{-y - u_0t}{\sqrt{t}}\right) \right] dt \\ - \int_0^{+\infty} (\lambda + \mu) e^{-(\lambda+\mu)t} \left[\mathcal{N}\left(\frac{x - u_0t}{\sqrt{t}}\right) - e^{2u_0y} \mathcal{N}\left(\frac{x - 2y - u_0t}{\sqrt{t}}\right) \right] dt,$$

we finally obtain

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L=T\}}) = \frac{\lambda}{\lambda + \mu} (F_0(s_0, \lambda + \mu, L, u_0) - F_\varphi(s_0, \lambda + \mu, L, u_0)) \\ = -\frac{\lambda}{\lambda + \mu} \Delta F_{0,\varphi}(s_0, \lambda + \mu, L, a_0 - \theta_0).$$

(ii) We have

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L>T\}}) = \left(\frac{(1 + \varphi)s_0}{L}\right)^{\theta_0/\sigma} E^{\theta_0}(\mathbf{1}_{\{\tau_L>T\}}),$$

where P^{θ_0} is the P-equivalent probability measure, such that

$$\left. \frac{dP^{\theta_0}}{dP} \right|_{\mathcal{F}_t} = \mathcal{E}(\theta_0 W)_t = \exp\left(\theta_0 W_t - \frac{1}{2}\theta_0^2 t\right)$$

and where E^{θ_0} is the P^{θ_0} -expectation. We can write

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L>T\}}) = \left(\frac{(1 + \varphi)s_0}{L}\right)^{\theta_0/\sigma} P^{\theta_0}(M_T^{a_0, \theta_0} < y, Y_T^{a_0, \theta_0} < x),$$

where

$$Y_t^{a_0, \theta_0} = a_0t + W_t - \theta_0t, \\ M_t^{a_0, \theta_0} = \sup(Y_s^{a_0, \theta_0}, 0 \leq t \leq s), \\ y = \left(\frac{1}{\sigma}\right) \ln\left(\frac{L}{s_0}\right), \quad \text{and} \\ x = \left(\frac{1}{\sigma}\right) \ln\left(\frac{L}{s_0(1 + \varphi)}\right).$$

Therefore, we obtain

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L > T\}}) = \left(\frac{(1 + \varphi)s_0}{L}\right)^{\theta_0/\sigma} F_\varphi(s_0, \lambda, L, a_0).$$

(iii) The third assertion follows directly from

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L < T\}}) = \left(\frac{s_0}{L}\right)^{\theta_0/\sigma} E^{\theta_0}(\mathbf{1}_{\{\tau_L < T\}}) = \left(\frac{s_0}{L}\right)^{\theta_0/\sigma} (1 - F_0(s_0, \lambda, L, a_0)).$$

(iv) We now consider $c(s_0, L) = E[e^{-\mu\tau_L}(S_{\tau_L} - 1)]$. Let us first study the case where the jump size is a positive constant φ . We have

$$E(S_{\tau_L} e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L = T\}}) = \frac{\lambda s_0(1 + \varphi)}{\lambda + \mu - \alpha} E^{\psi_2}(\mathbf{1}_{\{\tau_L = T\}}),$$

where $\psi_2 = (\mu - \alpha)/\lambda$ and P^{ψ_2} is the probability measure defined by $dP^{\psi_2}/dP^\sigma|_{\mathcal{F}_t} = \mathcal{E}(\sigma W)_t \mathcal{E}(\psi_2 M)_t$. Then we obtain

$$\begin{aligned} E(S_{\tau_L} e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L = T\}}) &= \frac{s_0(1 + \varphi)\lambda}{\lambda + \mu - \alpha} F_0(s_0, \lambda + \mu - \alpha, L, v_0) \\ &\quad - \frac{s_0(1 + \varphi)\lambda}{\lambda + \mu - \alpha} F_\varphi(s_0, \lambda + \mu - \alpha, L, v_0), \end{aligned}$$

where v_0 is defined by (8). We conclude the proof by using (i), (ii), (iii), and the equality $c(s_0, L) = (L - 1) E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L < T\}}) + (L - 1) E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L > T\}}) + E(e^{-\mu\tau_L}(S_{\tau_L} - 1) \mathbf{1}_{\{\tau_L = T\}})$.

Lemma 3. *Let L be a frontier such that $L \geq 1$ and $L > s_0$. For any φ in $]-1, 0[$, we have*

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L > T\}}) = \left(\frac{(1 + \varphi)s_0}{L}\right)^{\theta_0/\sigma} F_0(s_0, \lambda, L, a_0),$$

where a_0 , θ_0 , and $F_0(s_0, \lambda, L, a)$ are defined by (6), (7), and (4) respectively.

Proof. For any φ in $]-1, 0[$, we can write

$$E(e^{-\mu\tau_L} \mathbf{1}_{\{\tau_L > T\}}) = \left(\frac{(1 + \varphi)s_0}{L}\right)^{\theta_0/\sigma} E^{\theta_0}(\mathbf{1}_{\{\tau_L > T\}}).$$

In this case we have

$$E^{\theta_0}(\mathbf{1}_{\{\tau_L > T\}}) = F_0(s_0, \lambda, L, a_0).$$

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