

# OPTIMAL CLOSING OF A MOMENTUM TRADE

ERIK EKSTRÖM<sup>1</sup> AND CARL LINDBERG<sup>2</sup>

ABSTRACT. There is an extensive academic literature that documents that stocks which have performed well in the past often continue to perform well over some holding period - so called momentum. We study the optimal timing for an asset sale for an agent with a long position in a momentum trade. The asset price is modelled as a geometric Brownian motion with a drift that initially exceeds the discount rate, but with the opposite relation after an unobservable and exponentially distributed time. The problem of optimal selling of the asset is then formulated as an optimal stopping problem under incomplete information. Based on the observations of the asset, the agent wants to detect the unobservable change point as accurately as possible. Using filtering techniques and stochastic analysis, we reduce the problem to a one-dimensional optimal stopping problem, which we solve explicitly. We show also that the optimal boundary at which the investor should liquidate the trade depends monotonically on the model parameters.

## 1. INTRODUCTION

Momentum is the notion that an asset which has performed well in the past will continue to do so for some time. Momentum trades may be the most common asset management strategy, and there is a large literature about it. The pioneering paper [15] documents that stocks with high recent performance continue to generate significant positive returns over a 3–12 months holding period. This conclusion was further established in [17], and [31] found that momentum is present also in international markets. These papers also show that the momentum effect decays in time to eventually disappear entirely. Having accepted the existence of momentum as a "stylized fact" of equity markets, recent academic literature focuses on explaining the phenomenon; a non-exhaustive list is [2, 4, 5, 7–9, 11–14, 16, 18, 22–24, 26, 27].

Clearly, one way to benefit from a detected momentum opportunity is to buy the stock and then liquidate it at some pre-determined future time.

---

*Date:* November 1, 2011.

*2000 Mathematics Subject Classification.* Primary 91G10; Secondary 60G40.

*Key words and phrases.* Optimal stopping; momentum; trading; quickest detection problem for Brownian motion.

<sup>1</sup> Support from the Swedish Research Council (VR) is gratefully acknowledged.

<sup>2</sup> carl.lindberg@alumni.chalmers.se. Work supported by the Gothenburg Mathematical Modelling Centre (GMMC).

An obvious risk with such a strategy, however, is that the momentum effect disappears much earlier or much later than the pre-determined liquidation time.

In this paper, we analyze how to choose optimal liquidation strategies for momentum trades. More precisely, we seek to maximize the expected profit from a momentum trade. The expected return of a stock is assumed to be a constant larger than the discount rate up until some random, and unobservable, time  $\theta$ , at which it drops to a constant smaller than the discount rate. The assumption that the momentum trend eventually disappears is in accordance with the consensus in the financial literature, see e.g. [15], [17], and [31] (for an alternative model for momentum, see [36]). Based on this model for the expected return, we seek the optimal liquidation time for an investor with a long position in a momentum trade. Clearly, the investor wants to hold the position as long as the momentum is present, thereby taking advantage of the drift exceeding the discount rate. On the other hand, when the momentum disappears the investor would like to exit the position to avoid an investment in an asset with an unfavorable drift. However, the time point  $\theta$  at which the momentum disappears is not directly observable, so to exit the position exactly at  $\theta$  is not possible in general. Instead, the agent has to make a guess based merely on his observations of the asset price fluctuations.

We formulate the optimal liquidation problem as an optimal stopping problem under incomplete information. Applying the filtering techniques used in the quickest detection problem for a Brownian motion, see [32], we arrive at an optimal stopping problem under complete information. In addition to the underlying asset, this problem also involves the Markov process given by the conditional probability that  $\theta$  has occurred given the asset price observations up to current time. Next, using a Girsanov change of measure and stochastic analysis we reduce the number of spatial dimensions and express the optimal liquidation problem in terms of an auxiliary one-dimensional optimal stopping problem with an affine pay-off function for a diffusion. An explicit solution to this auxiliary problem is found by solving the related free boundary problem involving an ordinary differential equation. We show that for the auxiliary problem, the optimal stopping boundary is a constant barrier which depends monotonically on each of the model parameters. Accordingly, the optimal liquidation strategy for the momentum trade is to sell the asset the first time  $t$  that the conditional probability that  $\theta \leq t$  exceeds a certain level. We also analyze the dependence of the solution on the different model parameters. This is important, both from a mathematical and from a practical perspective, in order to obtain an understanding of the model sensitivity to changes in the parameter input and to possible mis-specifications. In particular, we show that the optimal liquidation level is decreasing in the intensity at which the momentum opportunity disappears, and it is also decreasing with respect to the volatility of the underlying asset.

The problem under study in the present paper was also addressed in [1]. The authors of [1] showed that the optimal liquidation time, for most parameter choices, coincides with the first hitting time of some level for the conditional probability that  $\theta$  has occurred. However, no characterization of this level or means of calculating it were provided. The same problem is also discussed in [34]. The authors of [34] write the liquidation problem as an optimal stopping problem for a two-dimensional Markov process. In view of the results of these references, the main mathematical contribution of the current paper is to show that the two-dimensional optimal stopping problem can be reduced to a one-dimensional problem using the Girsanov theorem, together with a detailed analysis of the resulting one-dimensional problem, thereby providing a full solution of the liquidation problem. For related studies where optimal stopping theory is applied to optimal stock selling problems, see e.g. [6, 10, 30, 33, 35].

We define the asset price model in Section 2, and the optimal liquidation problem is set up. Moreover, filtering techniques and the Girsanov theorem are used to formulate the problem in terms of a one-dimensional auxiliary optimal stopping problem. In Section 3, the auxiliary optimal stopping problem is studied, and the optimal stopping time is determined. In Section 4 we study the dependency of the solution on the different model parameters. Finally, in Section 5 we study a similar case but with an investor who has a short position in a momentum trade.

## 2. THE MODEL

We take as given a complete probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ . On this probability space, let  $\theta$  be a random variable with distribution  $\mathbb{P}(\theta = 0) = \pi$  and  $\mathbb{P}(\theta \geq t | \theta > 0) = e^{-\lambda t}$ . We assume that the intensity  $\lambda$  is positive and that  $\pi$  belongs to  $[0, 1)$ .

Next, let  $W$  be a Brownian motion which is independent of  $\theta$ . The asset price process  $X$  is modeled by a geometric Brownian motion with a drift that drops from  $\mu_2$  to  $\mu_1$  at time  $\theta$ . More precisely,

$$(1) \quad dX_t = \mu(t)X_t dt + \sigma X_t dW_t$$

and  $X_0 = x > 0$ , where

$$\mu(t) = \mu_2 - (\mu_2 - \mu_1)I(t \geq \theta).$$

Here  $\mu_1$ ,  $\mu_2$  and  $\sigma > 0$  are constants with  $\mu_1 < \mu_2$ . Denote by  $\mathcal{F}^X = (\mathcal{F}_t^X)_{t \in [0, \infty)}$  the filtration generated by  $X$ , and note that  $\mathcal{F}^X$  is strictly contained in the filtration generated by  $X$  and  $\theta$ .

Now, let  $\mathcal{T}$  be the set of (possibly infinite)  $\mathcal{F}^X$ -stopping times. We consider the optimal stopping problem

$$(2) \quad V = \sup_{\tau \in \mathcal{T}} \mathbb{E} e^{-r\tau} X_\tau,$$

where the discount rate  $r \geq 0$  is a given constant. Here (and in all other similar situations below) we use the convention that  $e^{-r\tau}X_\tau = 0$  on the event  $\{\tau = \infty\}$ .

Note that if  $r < \mu_1 < \mu_2$ , then  $e^{-rt}X_t$  is an  $\mathcal{F}^X$ -submartingale, and  $V = \infty$  (choose deterministic stopping times  $\tau = t$ , and let  $t \rightarrow \infty$ ). If  $r = \mu_1 < \mu_2$ , then  $e^{-rt}X_t$  is an  $\mathcal{F}^X$ -submartingale and again the supremum is attained along a sequence of deterministic times (the value  $V$  is then infinite if  $\mu_2 \geq r + \lambda$ , and  $V = \pi x + (1 - \pi)x\lambda/(\lambda + r - \mu_2)$  if  $\mu_2 < r + \lambda$ ). Similarly, if  $\mu_1 < \mu_2 \leq r$ , then  $e^{-rt}X_t$  is a supermartingale,  $V = x$ , and the supremum is attained for  $\tau = 0$ . Therefore, throughout the remainder of this article, we assume that  $\mu_1 < r < \mu_2$ .

**Remark** We emphasize that the supremum in (2) is with respect to  $\mathcal{F}^X$ -stopping times. This models a situation in which the agent cannot directly observe the change point  $\theta$ . Instead he merely observes the asset price trajectory, and based on these observations he tries to determine an optimal liquidation strategy.

On the other hand, an agent with complete information about the change point would choose to liquidate exactly at  $\theta$ . The expected pay-off for such an agent would then be

$$\begin{aligned} \mathbb{E}e^{-r\theta}X_\theta &= \pi x + (1 - \pi)x \int_0^\infty e^{(\mu_2 - r)t} \lambda e^{-\lambda t} dt \\ &= \begin{cases} \pi x + (1 - \pi)x\lambda/(\lambda + r - \mu_2) & \text{if } \mu_2 < r + \lambda \\ \infty & \text{if } \mu_2 \geq r + \lambda. \end{cases} \end{aligned}$$

Clearly, this value is an upper bound for  $V$ .

**Remark** Another possible interpretation of the liquidation problem described above would be that  $X$  models the price of an asset bubble. As long as the bubble lives, the agent clearly wants to keep the asset, but once the bubble bursts, it is optimal to sell it.

**Remark** The optimal liquidation problem (2) is also studied in [1]. The authors of [1] show that the optimal liquidation time is the first passage time over some unknown level for the a posteriori probability process  $\Pi$  defined below. However, no characterization of the optimal liquidation level is obtained.

In [34] a similar problem is studied, but with the difference that the asset price is modeled by a linear Brownian motion instead of a geometric Brownian motion. The authors also briefly discuss the case of a geometric Brownian motion as in (2), and derive a representation in terms of a two-dimensional optimal stopping problem. One key observation of the present paper is that, in fact, problem (2) can be reduced to a one-dimensional optimal stopping problem, and therefore also be solved explicitly.

We define the a posteriori probability process  $\Pi$  by

$$(3) \quad \Pi_t = \mathbb{P}(\theta \leq t | \mathcal{F}_t^X).$$

Note that  $\Pi_0 = \pi$ . It is well-known from filtering theory, compare [32], that  $\Pi$  satisfies

$$(4) \quad d\Pi_t = \lambda(1 - \Pi_t) dt - \omega\Pi_t(1 - \Pi_t) d\bar{W}_t,$$

where

$$\omega = (\mu_2 - \mu_1)/\sigma$$

and  $\bar{W}_t$  is a  $\mathbb{P}$ -Brownian motion with respect to  $\mathcal{F}_t^X$  given by

$$(5) \quad \begin{aligned} d\bar{W}_t &= \frac{dX_t}{\sigma X_t} - \frac{1}{\sigma}(\Pi_t\mu_1 + (1 - \Pi_t)\mu_2) dt \\ &= \frac{1}{\sigma}(\mu(t) - \Pi_t\mu_1 - (1 - \Pi_t)\mu_2) dt + dW_t. \end{aligned}$$

Moreover, in terms of  $\bar{W}$  we have

$$(6) \quad dX_t = (\Pi_t\mu_1 + (1 - \Pi_t)\mu_2)X_t dt + \sigma X_t d\bar{W}_t,$$

and the filtration generated by  $\bar{W}$  coincides with  $\mathcal{F}^X$ . Furthermore, if

$$\Phi_t := \frac{\Pi_t}{1 - \Pi_t},$$

then  $\Phi_0 = \phi := \pi/(1 - \pi)$  and Ito's formula gives that

$$d\Phi_t = \left( \frac{\lambda}{1 - \Pi_t} + \omega^2\Pi_t\Phi_t \right) dt - \omega\Phi_t d\bar{W}_t.$$

Next, we note that

$$M_t := \exp \left\{ -\frac{1}{2} \int_0^t (\sigma + \omega\Pi_s)^2 ds + \int_0^t (\sigma + \omega\Pi_s) d\bar{W}_s \right\}$$

is an  $\mathcal{F}^X$ -martingale, and apply a generalized version of Girsanov's theorem (Corollary 3.5.2 in [19]) to find the unique probability measure  $\mathbb{Q}$  such that

$$\mathbb{Q}(A) = \mathbb{E}[1_A M_t],$$

for  $A \in \mathcal{F}_t^X$ ,  $0 \leq t < \infty$ . Under  $\mathbb{Q}$ , the new process  $Z$  defined by

$$dZ_t = -(\sigma + \omega\Pi_t) dt + d\bar{W}$$

is a  $\{\mathbb{Q}, \mathcal{F}^X\}$ -Brownian motion. Furthermore,

$$d\Phi_t = (\lambda + (\lambda - \omega\sigma)\Phi_t) dt - \omega\Phi_t dZ_t.$$

Our first result shows that the Radon-Nikodym derivative  $Y$  can be expressed conveniently in terms of  $\Phi$  and  $X$ . This important observation is motivated by results of Klein who showed a similar result in [21] in a setting with incomplete information about a constant (but random) drift.

**Proposition 2.1.** *Let  $\tau$  be an  $\mathcal{F}^X$ -stopping time. Then*

$$\mathbb{E}e^{-r\tau} X_\tau = \frac{x}{1 + \phi} \mathbb{E}^{\mathbb{Q}} e^{(\mu_2 - \lambda - r)\tau} (1 + \Phi_\tau).$$

*Proof.* Let  $\eta_t = 1/M_t$ . Then

$$\begin{aligned}\eta_t &= \exp \left\{ \frac{1}{2} \int_0^t (\sigma + \omega \Pi_t)^2 dt - \int_0^t (\sigma + \omega \Pi_t) d\bar{W}_t \right\} \\ &= \exp \left\{ -\frac{1}{2} \int_0^t (\sigma + \omega \Pi_t)^2 dt - \int_0^t (\sigma + \omega \Pi_t) dZ_t \right\},\end{aligned}$$

and

$$\frac{d\eta_t}{\eta_t} = -(\sigma u + \omega \Pi_t) dZ_t.$$

Moreover,

$$(7) \quad \mathbb{E} e^{-r\tau} X_\tau = \mathbb{E}^\mathbb{Q} \eta_\tau e^{-r\tau} X_\tau.$$

Now, consider the process

$$Y_t = \frac{(1 + \Phi_t) e^{(\mu_2 - \lambda)t} x}{(1 + \phi) X_t}.$$

Using Ito's formula, it is straightforward to check that

$$\frac{dY_t}{Y_t} = -(\sigma + \omega \Pi_t) dZ_t = \frac{d\eta_t}{\eta_t}.$$

Since  $Y_0 = 1 = \eta_0$ , it follows that  $\eta_t = Y_t$  for all  $t$  almost surely. In view of (7), the result follows.  $\square$

It follows from Proposition 2.1 that

$$(8) \quad V = \frac{x}{1 + \phi} \sup_{\tau \in \mathcal{T}} \mathbb{E}^\mathbb{Q} e^{(\mu_2 - \lambda - r)\tau} (1 + \Phi_\tau).$$

In the next section, we study the optimal stopping problem in the above form. The implications for the problem (2) are summarized in Corollary 3.4.

### 3. THE AUXILIARY OPTIMAL STOPPING PROBLEM

In this section we study the optimal stopping problem in (8). Thus, on some probability space, let  $\Phi_t$  be a diffusion process satisfying

$$d\Phi_t = (\lambda + (\lambda - \omega\sigma)\Phi_t) dt - \omega\Phi_t dZ_t,$$

where  $Z$  is a standard Brownian motion. Define

$$(9) \quad F(y) = \sup_{\tau} \mathbb{E}_\phi e^{(\mu_2 - \lambda - r)\tau} (1 + \Phi_\tau),$$

where the index indicates that  $\Phi_0 = \phi$ , and the supremum is taken over random variables that are stopping times with respect to the filtration generated by  $Z$ . Note that  $V$  defined in (2) satisfies  $V = xF(\phi)/(1 + \phi)$ , compare (8).

The quantity  $F$  is infinite if  $\mu_2 \geq \lambda + r$  since  $\Phi$  will reach any positive level with probability one (choose  $\tau = \inf\{t : \Phi_t \geq b\}$  and let  $b \rightarrow \infty$ ).

We therefore assume throughout this section that  $\mu_2 < \lambda + r$ . An application of Ito's formula then yields that the drift of the process  $e^{(\mu_2 - \lambda - r)t} (1 + \Phi_t)$

is positive if  $\Phi_t < (\mu_2 - r)/(r - \mu_1)$  and negative if  $\Phi_t > (\mu_2 - r)/(r - \mu_1)$ . This suggests that the optimal stopping time in (9) is of the form

$$\tau_B := \inf\{t \geq 0 : \Phi_t \geq B\}$$

for some level  $B > (\mu_2 - r)/(r - \mu_1)$ , compare [37]. Moreover, optimal stopping theory (see [29]) suggests that the pair  $(F, B)$  satisfies the free boundary problem

$$(10) \quad \begin{cases} \mathcal{L}F + (\mu_2 - \lambda - r)F = 0 & 0 < \phi < B \\ F(\phi) = 1 + \phi & \phi \geq B \\ F'(B) = 1 \\ F(0+) < \infty \end{cases}$$

where

$$(11) \quad \mathcal{L}F = \frac{\omega^2 \phi^2}{2} F'' + (\lambda + (\lambda - \omega\sigma)\phi)F'.$$

The ordinary differential equation in (10) has the general solution

$$(12) \quad F(\phi) = C\psi(\phi) + D\varphi(\phi),$$

where

$$(13) \quad \psi(\phi) = \int_0^\infty e^{-at} t^{(b+\gamma-3)/2} (1 + \phi t)^{(\gamma-b+1)/2} dt$$

and

$$\varphi(\phi) = \int_0^{1/\phi} e^{at} t^{(b+\gamma-3)/2} (1 - \phi t)^{(\gamma-b+1)/2} dt,$$

with  $\gamma = \sqrt{(b-1)^2 + 4c}$ ,  $a = 2\lambda/\omega^2$ ,  $b = \frac{2}{\omega}(\frac{\lambda}{\omega} - \sigma)$ , and  $c = 2(\lambda + r - \mu_2)/\omega^2 > 0$ . (The functions  $\varphi$  and  $\psi$  can alternatively be expressed in terms of the confluent hypergeometric functions of the first and second kind, respectively.)

**Lemma 3.1.** *The function  $\psi$  is increasing and convex on  $(0, \infty)$ . The function  $\varphi$  is decreasing and satisfies  $\varphi(0+) = \infty$ .*

*Proof.* We have

$$\psi'(\phi) = \frac{\gamma - b + 1}{2} \int_0^\infty e^{-at} t^{(b+\gamma-1)/2} (1 + \phi t)^{(\gamma-b-1)/2} dt,$$

which is positive since  $\gamma = \sqrt{(b-1)^2 + 4c} > b - 1$ . Hence  $\psi$  is increasing. Moreover,

$$\frac{2\psi''(\phi)}{\gamma - b + 1} = \frac{(\gamma - b - 1)}{2} \int_0^\infty e^{-at} t^{(b+\gamma+1)/2} (1 + \phi t)^{(\gamma-b-3)/2} dt.$$

Now note that  $\mu_1 < r$  implies  $b < c$ , which in turn implies that  $\gamma > b + 1$ . Consequently,  $\psi$  is convex.

Similarly,

$$\varphi'(\phi) = \frac{b - 1 - \gamma}{2} \int_0^{1/\phi} e^{at} t^{(b+\gamma-1)/2} (1 - \phi t)^{(\gamma-b-1)/2} dt < 0,$$

so  $\varphi$  is decreasing. Moreover, for small  $\phi$  we have

$$\begin{aligned}\varphi(\phi) &= \int_0^{1/\phi} e^{at} t^{(b+\gamma-3)/2} (1-\phi t)^{(\gamma-b+1)/2} dt \\ &\geq (1-\sqrt{\phi})^{(\gamma-b+1)/2} \int_0^{1/\sqrt{\phi}} t^{(b+\gamma-3)/2} dt \\ &\geq \epsilon \phi^{(1-\gamma-b)/4},\end{aligned}$$

where  $\epsilon$  is some positive constant. Since  $\gamma > 1 - b$ , it follows that  $\varphi(0+) = \infty$ .  $\square$

**Remark** The function  $\psi$  is the unique (up to multiplication with a positive constant) positive and increasing solution of  $\mathcal{L}f + (\mu_2 - \lambda - r)f = 0$ , compare pages 18-19 in [3]. Similarly,  $\varphi$  is the unique positive and decreasing solution.

Note that the condition  $F(0+) < \infty$  implies that we must have  $D = 0$ . The two boundary conditions at  $B$  in (10) then imply that the two remaining unknowns  $B$  and  $C$  satisfy

$$\begin{cases} C\psi(B) = 1 + B \\ C\psi'(B) = 1. \end{cases}$$

Using (13) we find that

$$(14) \quad \begin{cases} C \int_0^\infty e^{-at} t^{(b+\gamma-3)/2} (1+Bt)^{(\gamma-b+1)/2} dt = 1 + B \\ C \frac{\gamma-b+1}{2} \int_0^\infty e^{-at} t^{(b+\gamma-1)/2} (1+Bt)^{(\gamma-b-1)/2} dt = 1, \end{cases}$$

and solving for  $B$  gives the free boundary equation

$$(15) \quad 2 \int_0^\infty e^{-at} t^{(b+\gamma-3)/2} (1+Bt)^{(\gamma-b+1)/2} dt = (\gamma-b+1)(1+B) \int_0^\infty e^{-at} t^{(b+\gamma-1)/2} (1+Bt)^{(\gamma-b-1)/2} dt.$$

**Lemma 3.2.** *There exists a unique positive solution  $B$  to the free boundary equation (15).*

*Proof.* Define the function  $g : (0, \infty) \rightarrow \mathbb{R}$  by

$$(16) \quad g(x) := 2 \int_0^\infty e^{-at} t^{(b+\gamma-3)/2} (1+xt)^{(\gamma-b+1)/2} dt - (\gamma-b+1)(1+x) \int_0^\infty e^{-at} t^{(b+\gamma-1)/2} (1+xt)^{(\gamma-b-1)/2} dt.$$

Then

$$\begin{aligned}g(0+) &= 2 \int_0^\infty e^{-at} t^{(b+\gamma-3)/2} dt - (\gamma-b+1) \int_0^\infty e^{-at} t^{(b+\gamma-1)/2} dt \\ &= \left( \frac{4a}{b+\gamma-1} - (\gamma-b+1) \right) \int_0^\infty e^{-at} t^{(b+\gamma-1)/2} dt,\end{aligned}$$



where the last equality follows by integration by parts. Since  $4a > 4c = (\gamma - b + 1)(\gamma + b - 1)$ , it follows that  $g(0+) > 0$ . Moreover,

$$g(x) \sim -x^{(\gamma-b-1)/2}(\gamma-b-1) \int_0^\infty e^{-at} t^{\gamma-1} dt$$

for large  $x$ . Since  $\gamma > b + 1$ , it follows that  $g(x)$  is negative for large  $x$ . Existence of a solution now follows from the continuity of  $g$ .

Uniqueness of solutions follows from the fact that

$$g'(x) = -(\gamma - b + 1)(1 + x) \int_0^\infty e^{-at} t^{(b+\gamma+1)/2} (1 + xt)^{(\gamma-b-3)/2} dt < 0$$

for all  $x \in (0, \infty)$ .  $\square$

We next show that the first hitting time of the level  $B$  is an optimal stopping time.

**Theorem 3.3.** *The function  $F$  defined in (12) satisfies*

$$F(\phi) = \begin{cases} \frac{1+B}{\psi(B)} \psi(\phi) & \phi < B \\ 1 + \phi & \phi \geq B, \end{cases}$$

where  $B$  is the unique solution to the free boundary equation (15). Moreover, we have  $B \geq (\mu_2 - r)/(r - \mu_1)$ , and the stopping time  $\tau_B := \inf\{t \geq 0 : \Phi_t \geq B\}$  is optimal in (12).

*Proof.* Let  $B$  be the unique solution to (15), and define  $G$  by

$$G(\phi) := \begin{cases} \frac{1+B}{\psi(B)} \psi(\phi) & \phi < B \\ 1 + \phi & \phi \geq B. \end{cases}$$

Then  $G(\phi) \geq 1 + \phi$ , and  $G(\phi) > 1 + \phi$  if and only if  $\phi < B$ . By (a generalized version of) Ito's formula, the process  $Y_t = e^{(\mu_2 - \lambda - r)t} G(\Phi_t)$  satisfies

$$dY_t = e^{(\mu_2 - \lambda - r)t} (\mu_2 - r + (\mu_1 - r)\Phi_t) 1_{\{\Phi_t > B\}} dt - e^{(\mu_2 - \lambda - r)t} \omega \Phi_t G'(\Phi_t) dZ_t.$$

Now assume that the solution  $B$  to the free boundary equation (15) satisfies  $B \geq (\mu_2 - r)/(r - \mu_1)$  (this is indeed verified below). Then the drift of  $Y$  is negative, and  $Y$  is a supermartingale. Moreover,  $Y_{t \wedge \tau_B}$  is a martingale with a last element.

Now, let  $\tau$  be a stopping time. An application of the optional sampling theorem (see Problem 1.3.16 and Theorem 1.3.22 in [19]) yields that

$$(17) \quad \mathbb{E}_\phi e^{(\mu_2 - \lambda - r)\tau} (1 + \Phi_\tau) \leq \mathbb{E}_\phi Y_\tau \leq \mathbb{E}_\phi Y_0 = G(\phi).$$

Consequently,  $F(\phi) \leq G(\phi)$ .

Conversely, for  $\tau = \tau_B$ , the inequalities in (17) are in fact equalities. Therefore,

$$F(\phi) = \sup_\tau \mathbb{E}_y^{\mathbb{Q}} Y_\tau \geq \mathbb{E}_y^{\mathbb{Q}} Y_{\tau_B} = G(\phi).$$

It follows that  $F(\phi) = G(\phi)$ .

It remains to show that  $B \geq (\mu_2 - r)/(r - \mu_1) =: d$ . To do so, first recall that, by general optimal stopping theory, see e.g. [20, Theorem D.12], it

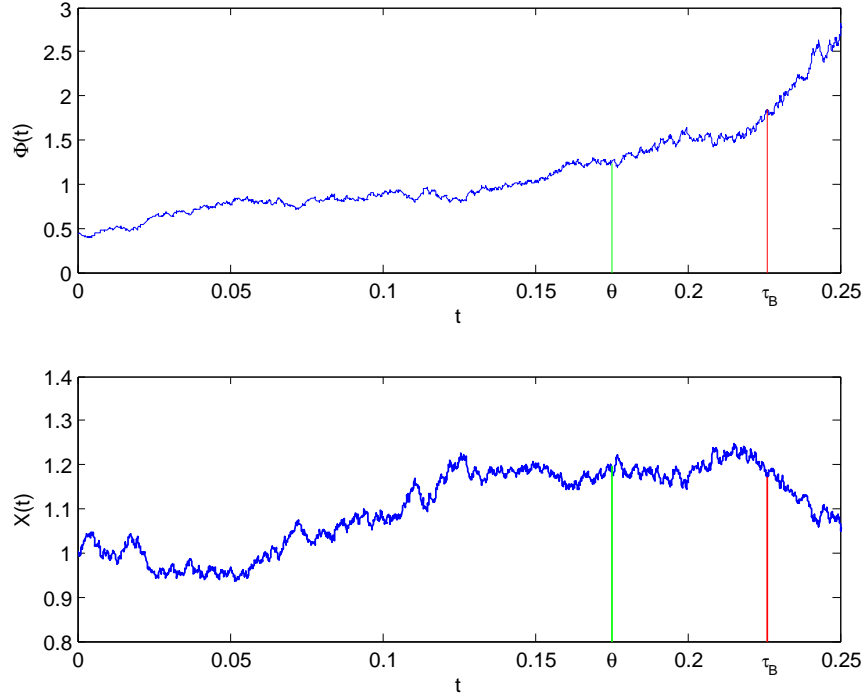


FIGURE 1. *Above:* A simulated path of the process  $\Phi_t$  for  $\mu_1 = -0.2$ ,  $\mu_2 = 0.2$ ,  $\sigma = 0.4$ ,  $r = 0.05$ , and  $\lambda = 3$ . These parameters yield  $B = 1.84$  using Equation (15). *Below:* The stock price  $X_t$  for the same simulation as above. The change point  $\theta$  and the optimal stopping time  $\tau_B$  are marked with green and red, respectively.

is optimal to stop the first time  $\Phi$  enters the stopping region  $\{\phi \in (0, \infty) : F(\phi) = 1 + \phi\}$ . Moreover, points where  $(\mathcal{L} + \mu_2 - \lambda - r)(1 + \phi) = \mu_2 - r + (\mu_1 - r)\phi$  is positive belong to the continuation region  $\{y \in (0, \infty) : F(\phi) > 1 + \phi\}$  (this follows by a local argument, see e.g. the remark after Theorem 10.1.12 in [28]). It follows that  $F$  satisfies

$$(18) \quad F(\phi) = \sup_{\tau: \Phi_\tau \geq d} \mathbb{E}_\phi e^{(\mu_2 - \lambda - r)\tau} (1 + \Phi_\tau),$$

i.e. it suffices to take supremum over stopping times  $\tau$  such that  $\Phi_\tau \geq d$ . Define the function

$$G^d(\phi) := \begin{cases} \frac{1+d}{\psi(d)} \psi(\phi) & \phi < d \\ 1 + \phi & \phi \geq d. \end{cases}$$

Now, assume that the unique solution  $B$  to the free boundary equation satisfies  $B < d$ . In that case, the function  $g$  in (16) satisfies  $g(d) < 0$ . Consequently, the left derivative of  $G^d$  at the point  $d$  is larger than the right derivative. In particular,  $G^d(\phi) < 1 + \phi$  for some  $\phi < d$ . It also follows (again, by

a generalized version of Ito's formula) that the process  $e^{(\mu_2 - \lambda - r)t} G^d(\Phi_t)$  is a supermartingale, and  $e^{(\mu_2 - \lambda - r)(t \wedge \tau_d)} G^d(\Phi_{t \wedge \tau_d})$  is a martingale. For stopping times  $\tau$  satisfying  $\Phi_\tau \geq d$  we therefore obtain

$$\mathbb{E}_\phi e^{(\mu_2 - \lambda - r)\tau} (1 + \Phi_\tau) = \mathbb{E}_\phi e^{(\mu_2 - \lambda - r)\tau} G^d(\Phi_\tau) \leq G^d(\phi).$$

In view of (18), the inequality  $F(\phi) \leq G^d(\phi)$  follows. However, this is a contradiction since there exist points with  $G^d(\phi) < 1 + \phi$ , whereas  $F(\phi) \geq 1 + \phi$  everywhere. This finishes the proof.  $\square$

**Remark** A more direct approach to prove that  $B \geq (\mu_2 - r)/(r - \mu_1)$  would be to show that  $g((\mu_2 - r)/(r - \mu_1)) \geq 0$ , where  $g$  is defined in (16). However, we did not find a short argument for this.

The following result is a consequence of Proposition 2.1 and Theorem 3.3.

**Corollary 3.4.** *Let  $B$  be the unique positive solution to (15), and  $\phi = \pi/(1 - \pi)$ . Then the value  $V$  defined in (2) satisfies*

$$V = \begin{cases} \frac{(1+B)\psi(\phi)}{(1+\phi)\psi(B)} x & \phi < B \\ x & \phi \geq B. \end{cases}$$

Moreover, the stopping time

$$\tau^* = \inf\{t \geq 0 : \Pi_t \geq B/(B + 1)\}$$

is optimal in (2).

#### 4. PARAMETER DEPENDENCIES

In this section we study the dependence of the optimal liquidation strategy and the value  $V$  on the different parameters of the model.

**Theorem 4.1.** *The value  $V$  is decreasing in  $\pi$ ,  $\lambda$  and  $\sigma$ . The optimal liquidation level  $B$  is decreasing in  $\lambda$  and  $\sigma$ .*

*Proof.* Recall that  $\phi \mapsto F(\phi)$  is convex and satisfies  $F(\phi) = 1 + \phi$  for  $\phi \geq B$ . It follows that

$$(19) \quad (1 + \phi)F'(\phi) - F(\phi) \leq 0.$$

Using the relation  $V = x \frac{F(\phi)}{1 + \phi}$  we find that

$$\frac{\partial V}{\partial \phi} = x \frac{\partial \frac{F(\phi)}{1 + \phi}}{\partial \phi} = \frac{(1 + \phi)F'(\phi) - F(\phi)}{(1 + \phi)^2} \leq 0.$$

Since  $\pi = \phi/(1 + \phi)$ ,  $\pi$  is increasing in  $\phi$ , so  $V$  is decreasing also in  $\pi$ .

Next, let  $\lambda_2 \geq \lambda_1 > 0$  and  $\sigma_2 \geq \sigma_1 > 0$ , and let  $F_1$  and  $F_2$  be the corresponding value functions defined by (12). Moreover, let  $B_1$  and  $B_2$  be the corresponding optimal liquidation levels and let  $\mathcal{L}_1$  and  $\mathcal{L}_2$  be the

corresponding linear operators defined by (11). For  $y \leq B_2$  it follows from the fact that  $(\mathcal{L}_2 + \mu - 2 - \lambda_2 - r)F_2 = 0$ , the convexity of  $F_2$  and (19) that

$$\begin{aligned} (\mathcal{L}_1 + \mu_2 - \lambda_1 - r)F_2 &= \frac{\phi^2(\mu_2 - \mu_1)^2}{2}(\sigma_1^{-2} - \sigma_2^{-2})F_2'' \\ &\quad + (\lambda_1 - \lambda_2)((1 + \phi)F_2' - F_2) \geq 0. \end{aligned}$$

If  $\Phi$  is the process with parameters  $\lambda_1$  and  $\sigma_1$ , then

$$F_2(\phi) \leq \mathbb{E}_\phi e^{(\mu_2 - \lambda_1 - r)\tau_{B_2}} F_2(\Phi_{\tau_{B_2}}) = \mathbb{E}_\phi e^{(\mu_2 - \lambda_1 - r)\tau_{B_2}} (1 + \Phi_{\tau_{B_2}}) \leq F_1(\phi).$$

Note that if  $\phi \geq B_2$ , then  $F_2(\phi) = 1 + \phi \leq F_1(\phi)$ , so this inequality holds for all  $\phi$ . Consequently,  $F$  (and hence also  $V$ ) is decreasing in  $\lambda$  and in  $\sigma$ . Since  $B = \inf\{\phi : F(\phi) = 1 + \phi\}$ , it also follows that the optimal liquidation level  $B$  is decreasing in  $\lambda$  and in  $\sigma$ .  $\square$

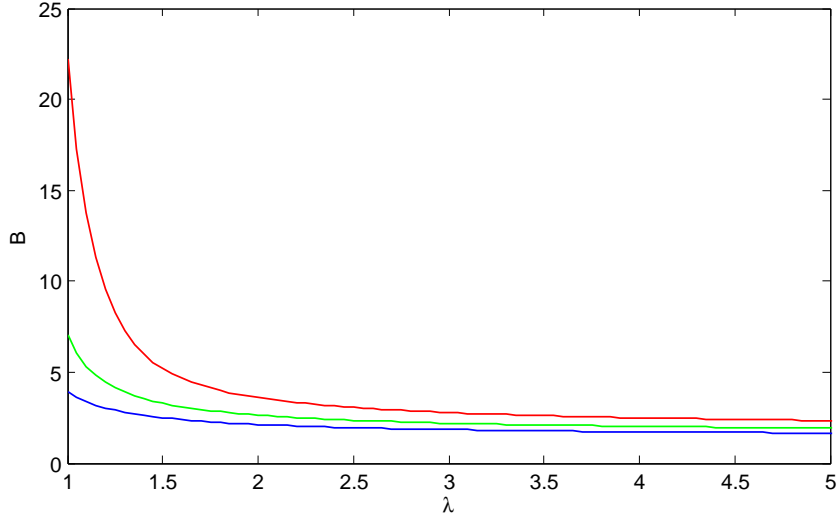


FIGURE 2. The optimal threshold  $B$  as a function of the jump intensity  $\lambda$  for three different values of the volatility  $\sigma$ . The model parameters are  $\mu_1 = -0.2$ ,  $\mu_2 = 0.2$  and  $r = 0.05$ . Blue:  $\sigma = 0.4$ , green:  $\sigma = 0.35$ , red:  $\sigma = 0.3$ .

## 5. OPTIMAL CLOSING OF A SHORT POSITION

In this section we consider the opposite situation in which the agent believes in a small initial drift, and that the small drift is replaced with a larger drift at a random and unobservable change-point. To trade on this belief, the agent has a short position in the asset, and he seeks an optimal time to close the short position. As in the previous case, the optimal closing time would be precisely at the change-point, but this is in general impossible due to the change-point being unobservable.

Let  $\theta$  be as in the previous sections, let  $\mu_1 < r < \mu_2$ , and model  $X$  by

$$dX_t = \mu(t)X_t dt + \sigma X_t dW_t,$$

where the drift now is given by

$$\mu(t) = \mu_1 + (\mu_2 - \mu_1)I(t \geq \theta).$$

We define

$$(20) \quad U = \inf_{\tau \in \mathcal{T}: \tau < \infty} \mathbb{E} e^{-r\tau} X_\tau.$$

Note that the infimum is taken over almost surely finite  $\mathcal{F}^X$ -stopping times (in this way we do not need to define the pay-off for  $\tau = \infty$ ). Define the a posteriori probability process  $\Pi$  by

$$\Pi_t = \mathbb{P}(\theta \leq t | \mathcal{F}_t^X),$$

and let  $\Phi_t = \Pi_t / (1 - \Pi_t)$ . By the same arguments as in the case of a long investor, we find that

$$\mathbb{E} e^{-r\tau} X_\tau = \frac{x}{1 + \phi} \mathbb{E}^\mathbb{Q} e^{(\mu_2 - \lambda - r)\tau} (1 + \Phi_\tau)$$

for finite  $\mathcal{F}^X$ -stopping times  $\tau$ . Here  $\Phi$  satisfies

$$d\Phi_t = (\lambda + (\lambda + \omega\sigma)\Phi_t) dt + \omega\Phi_t dZ_t,$$

where  $Z$  is a  $\{\mathbb{Q}, \mathcal{F}^X\}$ -Brownian motion. Consequently,

$$U = xH(\phi)/(1 + \phi),$$

where

$$(21) \quad H(\phi) = \inf_{\tau} E_{\phi}^{\mathbb{Q}} e^{(\mu_1 - \lambda - r)\tau} (1 + \Phi_\tau).$$

By Ito's formula, the drift of the process  $e^{(\mu_1 - \lambda - r)t} (1 + \Phi_t)$  is positive if  $\Phi_t > (r - \mu_1)/(\mu_2 - r)$  and negative if  $\Phi_t < (r - \mu_1)/(\mu_2 - r)$ . This suggests that the optimal stopping time is the first hitting time of the set  $[A, \infty)$  for some level  $A$ . Moreover, the pair  $(H, A)$  should satisfy the free boundary problem

$$(22) \quad \begin{cases} \mathcal{L}H + (\mu_1 - \lambda - r)H = 0 & 0 < \phi < A \\ H(\phi) = 1 + \phi & \phi \geq A \\ H'(A) = 1 \\ H(0+) < \infty \end{cases}$$

where

$$\mathcal{L}H = \frac{\omega^2 \phi^2}{2} H'' + (\lambda + (\lambda + \omega\sigma)\phi) H'.$$

The differential equation in (22) has the general solution

$$H(\phi) = C\psi(\phi) + D\varphi(\phi),$$

where

$$\psi(\phi) = \int_0^\infty e^{-at} t^{(b+\gamma-3)/2} (1 + \phi t)^{(\gamma-b+1)/2} dt$$

and

$$\varphi(\phi) = \int_0^{1/\phi} e^{at} t^{(b+\gamma-3)/2} (1-\phi t)^{(\gamma-b+1)/2} dt,$$

with  $\gamma = \sqrt{(b-1)^2 + 4c}$ ,  $a = 2\lambda/\omega^2$ ,  $b = \frac{2}{\omega}(\frac{\lambda}{\omega} + \sigma)$ , and  $c = 2(\lambda + r - \mu_1)/\omega^2$ . Using the boundary conditions in (22), we arrive at the free boundary equation

$$(23) \quad 2 \int_0^\infty e^{-at} t^{(b+\gamma-3)/2} (1+At)^{(\gamma-b+1)/2} dt = (\gamma-b+1)(1+A) \int_0^\infty e^{-at} t^{(b+\gamma-1)/2} (1+At)^{(\gamma-b-1)/2} dt.$$

The following result can then be proved along the same lines as Theorem 3.3 above.

**Theorem 5.1.** *Let  $A$  be the unique positive solution to the free boundary equation (23). Then*

$$H(\phi) = \begin{cases} \frac{1+A}{\psi(A)} \psi(\phi) & \phi < A \\ 1 + \phi & \phi \geq A. \end{cases}$$

Moreover, the stopping time  $\tau_A := \inf\{t \geq 0 : \Phi_t \geq A\}$  is optimal in (21).

Furthermore, with  $\varphi = \pi/(1-\pi)$  we have

$$U = \begin{cases} \frac{(1+A)\psi(\varphi)}{(1+\varphi)\psi(A)} x & \varphi < A \\ x & \varphi \geq A, \end{cases}$$

and the supremum in (20) is attained for  $\tau^* := \inf\{t \geq 0 : \Pi_t \geq A/(A+1)\}$ .

## REFERENCES

- [1] BEIBEL, M. AND LERCHE, H.R. A new look at optimal stopping problems related to mathematical finance. *Statistica Sinica* **7** (1997), 93–108.
- [2] BLITZ, D. C., VLIET, P. Global tactical cross-asset allocation: applying value and momentum across asset classes. *Journal of Portfolio Management* **35** (2008), 23–38.
- [3] BORODIN, A. AND SALMINEN, P. *Handbook of Brownian motion - facts and formulae*. Second edition. Probability and its Applications. Birkhäuser Verlag, Basel, 2002.
- [4] COOPER, M. J., GUTIERREZ, C. G. JR., HAMEED, A. Market states and momentum. *Journal of Finance* **59** (2004), 1345–1365.
- [5] CONNOLLY, R., STIVERS, C. Momentum and reversals in equity-index returns during periods of abnormal turnover and return dispersion. *Journal of Finance* **58** (2003), 1521–1555.
- [6] EKSTRÖM, E., LU, B. Optimal selling of an asset under incomplete information. To appear in *International Journal of Stochastic Analysis*, (2012).
- [7] FIGELMAN, I. Stock return momentum and reversal. *Journal of Portfolio Management* **34** (2007), 51–67.
- [8] GEBHARDT, W.R., HVIDKJAER, S., SWAMINATHAN, B. Stock and bond market interaction: does momentum spill over? *Journal of Financial Economics* **75** (2005), 651–690.
- [9] GEORGE T. J., HWANG, C.-Y. The 52-week high and momentum investing. *Journal of Finance* **59** (2004), 2145–2176.

- [10] GRAVERSEN, S. E., PESKIR, G., AND SHIRYAEV, A. N. Stopping Brownian motion without anticipation as close as possible to its ultimate maximum. *Theory of Probability and its Applications* **45** (2001), 125–136.
- [11] GRIFFIN, J. M., JI, X., MARTIN, J. P. Momentum investing and business cycle risk: evidence from pole to pole. *Journal of Finance* **58** (2003), 2515–2547.
- [12] GRUNDY, B. D., MARTIN, J. P. Understanding the nature of the risks and the source of the rewards to momentum investing. *Review of Financial Studies* **14** (2001), 29–78.
- [13] GUTIERREZ, R. C., KELLEY, E. The Long-Lasting Momentum in Weekly Returns. *Journal of Finance* **63** (2008), 415–447.
- [14] HOGAN, S., JARROW, R., MELVYN, T., WARACHKA, M. Testing market efficiency using statistical arbitrage with applications to momentum and value strategies. *Journal of Financial Economics* **73** (2003), 525–565.
- [15] JEGADEESH, N., TITMAN, S. Returns to buying winners and selling losers: Implications for stock market efficiency. *Journal of Finance* **48** (1993), 65–91.
- [16] JEGADEESH, N., TITMAN, S. Overreaction, delayed reaction, and contrarian profits. *Review of Financial Studies* **8** (1995), 973–993.
- [17] JEGADEESH, N., TITMAN, S. Profitability of momentum strategies: an evaluation of alternative explanations. *Journal of Finance* **56** (2001), 699–720.
- [18] JEGADEESH, N., TITMAN, S. Cross-sectional and time-series determinants of momentum returns. *Review of Financial Studies* **15** (2002), 143–157.
- [19] KARATZAS, I., SHREVE, S. *Brownian motion and stochastic calculus, 2nd ed.* Springer-Verlag, New York (2000).
- [20] KARATZAS, I., SHREVE, S. *Methods of mathematical finance.* Applications of Mathematics (New York), 39. Springer-Verlag, New York, 1998.
- [21] KLEIN, M. Comment on "Investment timing under incomplete information". *Mathematics of Operations Research* **34** (2009), 249–254.
- [22] KORAJCZYK, R. A., SADKA, R. Are momentum profits robust to trading costs? *Journal of Finance* **59** (2004), 1039–1082.
- [23] LEE, C. M. C., SWAMINATHAN, B. Price momentum and trading volume. *Journal of Finance* **55** (2000), 2017–2069.
- [24] LEWELLEN, J. Momentum and autocorrelation in stock returns. *Review of Financial Studies* **15** (2002), 533–563.
- [25] LIPTSER, R. S., SHIRYAEV, A. N. *Statistics of Random Processes II.* Springer-Verlag: New York–Heidelberg (1978).
- [26] LIU, L. X., ZHANG, L. Momentum profits, factor pricing, and macroeconomic risk. *Review of Financial Studies* **21** (2009), 2417–2448.
- [27] MOSKOWITZ, T. J., GRINBLATT, M. Do industries explain momentum? *Journal of Finance* **54** (1999), 1249–1290.
- [28] OKSENDAL, B. *Stochastic differential equations. An introduction with applications.* Sixth edition. Universitext. Springer-Verlag, Berlin, 2003.
- [29] PESKIR, G., SHIRYAEV, A. *Optimal stopping and free-boundary problems.* Lectures in Mathematics ETH Zürich, Birkhäuser Verlag: Basel (2006).
- [30] PESKIR, G., DU TOIT, J. Selling a stock at the ultimate maximum. *Annals of Applied Probability* **19** (2009), 983–1014.
- [31] ROUWENHORST, K. G. International momentum strategies. *Journal of Finance* **53** (1998), 267–284.
- [32] SHIRYAEV, A. N. *Optimal Stopping Rules.* Springer Verlag: New York - Heidelberg (1978).
- [33] SHIRYAEV, A. N. Quickest detection problems in the technical analysis of the Financial data. *Math. Finance Bachelier Congress (Paris 2000)*, Springer Verlag: New York - Heidelberg, (2002) 487 – 521.
- [34] SHIRYAEV, A.N. AND NOVIKOV, A.A. On a stochastic version of the trading rule "Buy and Hold". *Statistics and Decisions* **26** (2008), 289–302.

- [35] SHIRYAEV, A. N., XU, Z., ZHOU, X. Y. Thou shalt buy and hold. *Quantitative Finance* **8** (2008), 765–776.
- [36] STOJANOVIC, S. Optimal momentum hedging via hypoelliptic reduced Mone-Ampere PDEs. *SIAM J. Control Optim.* **43** (2004/05), 1151–1173.
- [37] VILLENEUVE, S. On threshold strategies and the smooth-fit principle for optimal stopping problems. *Journal of Applied Probability* **44** (2007), 181–198.

E. EKSTRÖM: UPPSALA UNIVERSITY, BOX 480, 75106 UPPSALA, SWEDEN.

C. LINDBERG: CHALMERS UNIVERSITY OF TECHNOLOGY, SE-412 96 GÖTEBORG, SWEDEN.