CONTROL AND STOPPING OF A DIFFUSION PROCESS ON AN INTERVAL

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Consider a process $X(\cdot) = \{X(t), 0 \le t < \infty\}$ which takes values in the interval I = (0, 1), satisfies a stochastic differential equation

$$dX(t) = \beta(t) dt + \sigma(t) dW(t), \qquad X(0) = x \in I$$

and, when it reaches an endpoint of the interval I, it is absorbed there. Suppose that the parameters β and σ are selected by a controller at each instant $t \in [0,\infty)$ from a set depending on the current position. Assume also that the controller selects a stopping time τ for the process and seeks to maximize $\mathbf{E}u(X(\tau))$, where $u\colon [0,1]\to \Re$ is a continuous "reward" function. If $\lambda:=\inf\{x\in I\colon u(x)=\max u\}$ and $\rho:=\sup\{x\in I\colon u(x)=\max u\}$, then, to the left of λ , it is best to maximize the mean-variance ratio (β/σ^2) or to stop, and to the right of ρ , it is best to minimize the ratio (β/σ^2) or to stop. Between λ and ρ , it is optimal to follow any policy that will bring the process $X(\cdot)$ to a point of maximum for the function $u(\cdot)$ with probability 1, and then stop.

1. Formulation of the problem. Suppose that for every x in the interval I=(l,r) with $-\infty < l < r < \infty$, there is a nonempty subset $\mathscr{K}(x)$ of $\mathfrak{R} \times (0,\infty)$ that specifies the drift-diffusion pairs (β,σ) available for controlling the stochastic process $X(\cdot)$ at any time $t\in[0,\infty)$, when the current position is X(t)=x. We also set $\mathscr{K}(l)=\mathscr{K}(r)=\{(0,0)\}$, meaning that the endpoints of the interval I are absorbing barriers once they are reached.

More formally, consider a controlled diffusion process $X(\cdot)$ on a filtered probability space $(\Omega, \mathcal{F}, \mathbf{P})$, $\mathbf{F} = \{\mathcal{F}(t), 0 \le t < \infty\}$, and such that

(1)
$$dX(t) = \beta(t) dt + \sigma(t) dW(t), \qquad X(0) = x \in I.$$

Here $W(\cdot)$ is an **F**-Brownian motion, and $\beta(\cdot)$, $\sigma(\cdot)$ are **F**-progressively measurable processes that satisfy

$$\int_0^t [|\beta(u)| + \sigma^2(u)] \, du < \infty, \qquad (\beta(t), \sigma(t)) \in \mathcal{K}(X(t))$$

almost surely, for all $0 < t < \infty$.

Given an initial position $X(0) = x \in I$, let us denote by $\mathscr{A}(x)$ the set of all processes $X(\cdot)$ that can be constructed this way (and are thus "available" to the controller at the initial position x). For every such process $X(\cdot) \in \mathscr{A}(x)$, let $\mathbf{F}^X := \{\mathscr{F}^X(t), 0 \le t < \infty\}$ be the filtration generated by the process

Received September 1997; revised April 1998.

¹Supported by U.S. Army Research Office Grant DAAH 04-95-I-0528.

²Supported by NSF Grant DMS-97-03285.

AMS 1991 subject classifications. Primary 93E20, 60G40; secondary 62L15, 60D60. Key words and phrases. Stochastic control, optimal stopping, one-dimensional diffusions.

 $X(\cdot)$, where $\mathscr{F}^X(t) := \sigma(X(s), \ 0 \le s \le t)$ denotes the "history" of the process $X(\cdot)$ up to time t. Also, let \mathscr{S}_X be the class of \mathbf{F}^X -stopping times, namely measurable functions $\tau \colon \Omega \to [0,\infty]$ with the property $\{\tau \le t\} \in \mathscr{F}^X(t)$, for every $0 \le t < \infty$. Finally, let $u \colon [l,r] \to \Re$ be a continuous "reward" function.

The "leavable" stochastic control problem (with discretionary stopping) addressed in this paper, is to find, for each $x \in I$, a process $X^*(\cdot) \in \mathscr{A}(x)$ and an \mathbf{F}^{X^*} -stopping time τ_* that attain the supremum

(2)
$$V(x) := \sup_{X(\cdot) \in \mathscr{A}(x), \ \tau \in \mathscr{I}_X} \mathbf{E}u(X(\tau)).$$

Under mild regularity conditions, the problem of (2) admits a surprisingly simple solution. This is presented in Section 4, after some preliminary material on one-dimensional diffusions (Section 2) and on optimal stopping (Section 3). The solution incorporates features of the theory of optimal stopping for Markov processes [e.g., Dynkin and Yushkevich (1969), Fakeev (1971), Shiryaev (1978)], as well as aspects of "controlling a diffusion process to a goal" [cf. Pestien and Sudderth (1985)], in a rather unexpected way.

2. A brief review of one-dimensional diffusions. Consider a diffusion process

(3)
$$dX(t) = b(X(t)) dt + s(X(t)) dW(t), X(0) = x \in I,$$

where $b: I \to \Re$, $s: I \to \Re$ are Borel-measurable functions that satisfy

(4)
$$s^2(x) > 0$$
, $\int_{x-\varepsilon}^{x+\varepsilon} s^{-2}(y)[1+|b(y)|] dy < \infty$ for some $\varepsilon > 0$

at every $x \in I$. Under these conditions, the stochastic differential equation (S.D.E.) of (3) has a weak solution which is unique in the sense of probability law, up to the "explosion time"

(5)
$$S := \inf \{ t \ge 0 \colon X(t) \notin I \} = \lim_{n \to \infty} \uparrow S_n.$$

We have set $S_n := \inf\{t \geq 0 \colon X(t) \notin (l_n, r_n)\}$, where the sequence $\{l_n\}$ decreases to the left endpoint l of I, and the sequence $\{r_n\}$ increases to the right endpoint r of I [cf. Karatzas and Shreve (1991), page 341, Theorem 5.15]. On $\{S < \infty\}$ we set X(t) := X(S) for $S \leq t < \infty$, so that both l and r are absorbing barriers.

An important tool in the study of diffusion processes and for the solution to our problem is the *scale function*

(6)
$$p(x) := \int_{c}^{x} \exp\left[-2\int_{c}^{\xi} (b/s^{2})(u) du\right] d\xi, \qquad x \in I$$

(with arbitrary but fixed $c \in I$). The function $p(\cdot)$ is a one-to-one mapping of the interval I onto the interval $\tilde{I} = (\tilde{l}, \tilde{r})$ with $\tilde{l} := p(l+), \tilde{r} := p(r-)$; it is continuous, with derivative

(7)
$$p'(x) = \exp\left[-2\int_{c}^{x} (b/s^{2})(u) du\right] > 0,$$

which is absolutely continuous, namely

$$p'(x) = 1 + \int_{c}^{x} p''(\xi) d\xi$$
 where $p''(x) := -\frac{2b(x)}{s^{2}(x)} p'(x)$

for $x \in I$. The inverse mapping $q: \tilde{I} \to I$ satisfies p(q(y)) = y and has derivative

$$q'(y) = \frac{1}{p'(q(y))} > 0, \quad y \in \tilde{I}.$$

It is easy to see then that the process

$$(8) Y(t) := p(X(t)), 0 < t < \infty$$

satisfies the S.D.E. with zero-drift (in "natural scale")

(9)
$$dY(t) = \tilde{s}(Y(t)) dW(t)$$

with $Y(0) = p(x) \in \tilde{I}$, where

(10)
$$\tilde{s}(y) := (p's)(q(y)), \qquad y \in \tilde{I}.$$

Observe that $\tilde{s}^2(y) > 0$ for every $y \in \tilde{I}$, by (4) and (7).

Remark 2.1. Under the conditions (4), the process $X(\cdot)$ exits from every proper subinterval of I with probability 1 [cf. Karatzas and Shreve (1991), page 344]. Hence, the process $Y(\cdot)$ exits from every proper subinterval of \tilde{I} with probability 1.

REMARK 2.2. Let us consider now the function

(11)
$$v(x) := \int_{c}^{x} p'(y) \int_{c}^{y} \frac{2 dz}{p'(z)s^{2}(z)} dy, \qquad x \in I.$$

Feller's (1952) *test for explosions* states that a necessary and sufficient condition for the explosion time S of (5) to be infinite almost surely, is

$$(12) v(l+) = v(r-) = \infty$$

[cf. Karatzas and Shreve (1991), page 348]. Of course, S is also the time of first exit for $Y(\cdot)$ from \tilde{I} ; since this latter process is time-changed Brownian motion, it is clear that (12) is satisfied, if

$$p(l+) = -\infty, \qquad p(r-) = \infty.$$

Condition (12) fails if $\mathbf{P}[S < \infty]$ is positive. We have in fact $\mathbf{P}[S < \infty] = 1$, if and only if one of the following conditions

$$v(l+) < \infty \qquad \text{and} \quad v(r-) < \infty, \quad \text{or}$$

$$p(l+) = -\infty \quad \text{and} \quad v(r-) < \infty, \quad \text{or}$$

$$v(l+) < \infty \qquad \text{and} \quad p(r-) = \infty$$

holds; cf. Karatzas and Shreve (1991), page 350. The first of these conditions implies $\mathbf{E}(S) < \infty$, as well as

(14)
$$p(l+) > -\infty \quad \text{and} \quad p(r-) < \infty.$$

3. Optimal stopping of a diffusion process. Let $X(\cdot)$ be a diffusion process as in (3), and let $u: [l, r] \to \Re$ be a continuous function. Define the value function of the associated *optimal stopping problem*

(15)
$$Q(x) := \sup_{\tau \in \mathscr{S}_X} \mathbf{E}^x u(X(\tau)), \qquad x \in I.$$

Here and in the sequel, the superscript records the initial position of the process under consideration; \mathscr{I}_X denotes the collection of \mathbf{F}^X -stopping times $\tau \colon \Omega \to [0, \infty]$, and we are using the convention $\xi(\infty) := \limsup_{t \to \infty} \xi(t)$.

PROPOSITION. Assume the conditions (4) and (14). Then the function $Q(\cdot)$ is continuous on [l, r] and can be written as the lower envelope of all affine transformations of the scale function $p(\cdot)$ that dominate $u(\cdot)$, namely,

(16)
$$Q(x) = \inf\{\alpha + \beta p(x) : \alpha, \beta \in \Re, \ \alpha + \beta p(\cdot) \ge u(\cdot)\}, \quad x \in I.$$

The stopping time

(17)
$$\tau_* := \inf\{t \ge 0: u(X(t)) = Q(X(t))\}\$$

belongs to \mathcal{S}_X , and attains the supremum in (15).

REMARK 3.1. The optimal stopping region $\Sigma := \{x \in I: u(x) = Q(x)\}$ is a closed subset of the interval I, and τ_* is the time of first entry for the process $X(\cdot)$ into this region. From Remarks 2.2 and 2.1, respectively, it should be clear that τ_* is almost surely finite if condition (13) holds [since τ_* is no greater than the time it takes $X(\cdot)$ to leave the interval I], or if

there exists
$$\varepsilon > 0$$
, such that the intervals $[l, l + \varepsilon]$ and $[r - \varepsilon, r]$ are included in Σ .

Under either of these two conditions, one can restrict attention to stopping times in \mathscr{S}_X that are almost surely finite, without changing the value of the supremum in (15).

For the special case that $X(\cdot)$ is Brownian motion, an elegant and elementary proof of the proposition is given by Dynkin and Yushkevich (1969). More general treatments of optimal stopping problems for continuous-time processes can be found in Shiryaev (1973, 1978), Fakeev (1970, 1971), El Karoui (1981). The arguments in Dynkin and Yushkevich (1969) apply with only minor changes to the diffusion-with-zero-drift process $Y(\cdot)$ of (8) and (9), and it is easy to pass from the optimal stopping problem for $Y(\cdot)$ to that for $X(\cdot)$, as we now demonstrate.

Clearly $X(\cdot)$ and $Y(\cdot)$ generate the same filtration: $\mathbf{F}^X = \mathbf{F}^Y$. Consequently, with \mathscr{S}_Y denoting the class of \mathbf{F}^Y -stopping times and setting

(19)
$$\tilde{u}(y) := u(q(y)),$$

(20)
$$\tilde{Q}(y) := \sup_{\tau \in \mathscr{S}_{Y}} \mathbf{E}^{y} \tilde{u}(Y(\tau))$$

for $y \in \tilde{I}$, we have

(21)
$$Q(x) = \sup_{\tau \in \mathcal{S}_{V}} \mathbf{E}^{p(x)} u(q(Y(\tau))) = \tilde{Q}(p(x)).$$

The proposition is thus immediate from the lemma below.

LEMMA. Under the conditions (4) and (14), the function $\tilde{Q}(\cdot)$ is concave and continuous on the interval $[\tilde{l}, \tilde{r}]$ and is the lower envelope of all the affine functions that dominate $\tilde{u}(\cdot)$,

(22)
$$\tilde{Q}(y) = \inf\{\alpha + \beta y : \alpha, \ \beta \in \Re, \ \alpha + \beta \xi \ge \tilde{u}(\xi) \text{ for all } \xi \in \tilde{I}\}, \quad y \in \tilde{I}.$$

The stopping time τ_* of (17) can also be written as

$$\tau_* = \inf \big\{ t \geq 0 \colon \tilde{u}(Y(t)) = \tilde{Q}(Y(t)) \big\},$$

and attains the supremum in (20).

Sketch of the Proof. The process $Y(\cdot)$ of (9) with $Y(\cdot)=y$ is a local martingale, and takes values in the compact interval $[\tilde{l},\tilde{r}]$ [thanks to (9), (14)]. It is thus a bounded martingale and, if $\varphi(\cdot)$ is an affine (or even concave) function dominating $\tilde{u}(\cdot)$, we have by Jensen's inequality and the optional sampling theorem,

(24)
$$\mathbf{E}\tilde{u}(Y(\tau)) \le \mathbf{E}\varphi(Y(\tau)) \le \varphi(\mathbf{E}(Y(\tau))) = \varphi(y),$$

for every $\tau \in \mathscr{I}_Y$. It follows that $\tilde{Q}(\cdot) \leq \varphi(\cdot)$ and that the right-hand side of (22) dominates the left. On the other hand, $\tilde{Q}(\cdot) \geq \tilde{u}(\cdot)$ since we can choose $\tau \equiv 0$ in (20), and $\tilde{Q}(\cdot)$ is concave [cf. Dynkin and Yushkevich (1969), page 115]. Thus $\tilde{Q}(\cdot)$ dominates the right-hand side of (22), as this latter coincides with the smallest concave majorant of $\tilde{u}(\cdot)$, and the equality in (22) is proved.

To see that the stopping time τ_* of (17) is optimal for the problem (20), introduce the function

$$h(y) := \mathbf{E}^y \tilde{u}(Y(\tau_*)), \qquad y \in \tilde{I}$$

and argue [as in Dynkin and Yushkevich (1969), pages 117 and 118] that $h(\cdot)$ is a concave majorant of $\tilde{u}(\cdot)$; hence, $h(\cdot) \geq \tilde{Q}(\cdot)$. The reverse inequality is an immediate consequence of the definition (20) of $\tilde{Q}(\cdot)$. The *continuity* of $\tilde{Q}(\cdot)$ on the open interval $\tilde{I} = (\tilde{l}, \tilde{r})$ follows from concavity; its continuity at the

endpoints \tilde{l} and \tilde{r} requires a special argument, which is provided in Dynkin and Yushkevich [(1969), pages 116 and 117]. \square

From elementary properties of concave functions [e.g., Karatzas and Shreve (1991), Section 3.6], $\tilde{Q}(\cdot)$ has left- and right-derivatives $D_{\pm}\tilde{Q}(\cdot)$ everywhere on \tilde{I} ; these latter functions are decreasing and left- (resp., right-) continuous, and satisfy

(25)
$$D_{+}\tilde{Q}(y) \leq D_{-}\tilde{Q}(y), \qquad y \in \tilde{I}.$$

The inequality is strict for at most countably many points. From (21) and (24), we have

(26)
$$D_{+}Q(x) = p'(x)D_{+}\tilde{Q}(p(x))$$

and thus

$$(27) \quad D_{-}Q(x) - D_{+}Q(x) = p'(x)[D_{-}\tilde{Q}(p(x)) - D_{+}\tilde{Q}(p(x))] \ge 0, \qquad x \in I.$$

Again, strict inequality occurs on a set which is at most countable.

Let us introduce now the smallest and largest locations of the maximum

$$(28) u^* := \max_{x \in [l, r]} u(x)$$

of the reward function $u(\cdot)$, namely,

(29)
$$\lambda := \inf\{x \in I: u(x) = u^*\}, \qquad \rho := \sup\{x \in I: u(x) = u^*\},$$

respectively, and their counterparts

$$\tilde{\lambda} := p(\lambda) = \inf\{y \in \tilde{I}: \tilde{u}(y) = u^*\}, \qquad \tilde{\rho} := p(\rho) = \sup\{y \in \tilde{I}: \tilde{u}(y) = u^*\}.$$

From the fact that $\tilde{Q}(\cdot)$ is the smallest concave majorant of $\tilde{u}(\cdot)$ (cf. the Lemma), we have

$$\begin{cases} D_{+}\tilde{Q}(\cdot)\geq 0, & \text{ on } (\tilde{l},\tilde{\lambda}), \\ D_{-}\tilde{Q}(\cdot)\leq 0, & \text{ on } (\tilde{\rho},\tilde{r}), \\ \tilde{Q}(\cdot)=u^{*}, & \text{ on } [\tilde{\lambda},\tilde{\rho}]. \end{cases}$$

By virtue of (21), these lead to

$$\begin{cases} D_+Q(\cdot)\geq 0, & \text{ on } (l,\lambda),\\ D_-Q(\cdot)\leq 0, & \text{ on } (\rho,r),\\ Q(\cdot)=u^*, & \text{ on } [\lambda,\rho]. \end{cases}$$

4. Solution of the problem. Consider again the stochastic control problem of Section 1. We shall assume that there exist two pairs (b_r, s_r) and (b_l, s_l) of Borel-measurable, real-valued functions on I, each of which satisfies the

conditions of (4) and (14), as well as

$$(31) (bl(x), sl(x)) \in \mathcal{X}(x), (br(x), sr(x)) \in \mathcal{X}(x),$$

(32)
$$\frac{b_l(x)}{s_l^2(x)} = \sup \left\{ \frac{\beta}{\sigma^2} : (\beta, \sigma) \in \mathcal{X}(x) \right\},$$

(33)
$$\frac{b_r(x)}{s_r^2(x)} = \inf \left\{ \frac{\beta}{\sigma^2} : (\beta, \sigma) \in \mathcal{K}(x) \right\},$$

for all $x \in I$. Consider the corresponding diffusion processes $X_l(\cdot)$, $X_r(\cdot)$ with these parameters, namely,

$$dX_{l}(t) = b_{l}(X_{l}(t)) dt + s_{l}(X_{l}(t)) dW(t),$$

$$dX_r(t) = b_r(X_r(t)) dt + s_r(X_r(t)) dW(t)$$

as in (3), and let $Q_l(\cdot) = \sup_{\tau \in \mathscr{I}_{X_l}} \mathbf{E}^{\cdot} u(X_l(\tau)), \ Q_r(\cdot) = \sup_{\tau \in \mathscr{I}_{X_r}} \mathbf{E}^{\cdot} u(X_r(\tau))$ be the value functions for the associated optimal stopping problems in (15) with the same reward function $u(\cdot)$.

Theorem. With the above assumptions and notation, the value function $V(\cdot)$ for the stochastic control problem of (2) satisfies

$$V(x) = \begin{cases} Q_l(x), & x \in (l, \lambda), \\ Q_r(x), & x \in (\rho, r), \\ u^*, & x \in [\lambda, \rho]. \end{cases}$$

In other words, the theorem states the following.

- 1. To the left of λ , it is best either to *maximize* the mean-variance ratio (β/σ^2) , or to stop [when this is optimal in the stopping problem (15) for the diffusion process $X_l(\cdot)$].
- 2. To the right of ρ , it is best either to *minimize* the mean-variance ratio (β/σ^2) , or to *stop* [when this is optimal in the stopping problem (15) for the diffusion process $X_r(\cdot)$].
- 3. In the interval $[\lambda, \rho]$, it is optimal to follow any policy that will bring the state-process $X(\cdot)$ arbitrarily close to a point of maximum of the function $u(\cdot)$, with probability 1—for example, by following the dynamics of either $X_l(\cdot)$ or $X_r(\cdot)$ —and then stop.

Notice that any one of the intervals (l, λ) , (ρ, r) or (λ, ρ) could be empty.

REMARK 4.1. It was shown by Pestien and Sudderth (1985) that the best way to control a diffusion, so as to maximize the probability of ever reaching the right-hand endpoint of an interval ("goal"), is to *maximize* the mean-variance ratio (β/σ^2). By analogy then, the best way to maximize the chance of reaching a "goal" on the left, is to *minimize* this same ratio. Thus, the strategy described in 1–3 above seeks to maximize the chance of moving toward the

location of a maximum of the reward function $u(\cdot)$, or to stop. For the problem of Section 1, it is not obvious a priori that one should follow such a strategy everywhere on I, even in the vicinity of local maxima of $u(\cdot)$.

PROOF OF THE THEOREM. Let $H(\cdot)$ be the function on the right-hand side of (34). Then $H(\cdot) \leq V(\cdot)$ on I because, for every $x \in I$, H(x) is the expected payoff from the policy described in 1–3 above. Thus, it remains to be shown that $V(\cdot) \leq H(\cdot)$.

Clearly from (34), it suffices to show $V(\cdot) \leq Q_l(\cdot)$ on the interval (l,λ) [a similar methodology will show as well that $V(\cdot) \leq Q_r(\cdot)$ holds on the interval (ρ,r)]. Thus, let us fix $x \in (l,\lambda)$ and observe that the value function of (2) can also be written then as

(35)
$$V(x) = \sup_{X(\cdot) \in \mathscr{A}(x), \ \tau \in \mathscr{S}_X} \mathbf{E}u(X(\tau \wedge \tau_{\lambda}^X)), \qquad x \in (l, \lambda),$$

where we have set

(36)
$$\tau_{\lambda}^{X} := \inf\{t \ge 0: X(t) = \lambda\}$$

for any given process $X(\cdot) \in \mathcal{A}(x)$. Indeed, once the nearest global maximum of the reward function $u(\cdot)$ has been reached, there is no point in proceeding any further.

On the other hand, let us consider the bounded processes

(37)
$$\xi_l(t) := p_l(X(t)), \qquad 0 \le t < \infty$$

and

(38)
$$\eta_l(t) := \tilde{Q}_l(\xi_l(t)) = Q_l(X(t)), \qquad 0 \le t < \infty.$$

From Itô's rule and (6), (32) we conclude that

$$\begin{split} \xi_l(t) &= p_l(x) + \int_0^t p_l'(X(s)) \, dX(s) + \left(\frac{1}{2}\right) \int_0^t p_l''(X(s)) \sigma^2(s) \, ds \\ &= p_l(x) - \int_0^t \left[\frac{b_l(X(s))}{s_l^2(X(s))} - \frac{\beta(s)}{\sigma^2(s)}\right] p_l'(X(s)) \sigma^2(s) \, ds \\ &+ \int_0^t p_l'(X(s)) \sigma(s) \, dW(s), \qquad 0 \le t < \infty \end{split}$$

is a supermartingale. It is an easy consequence of Jensen's inequality that a concave, bounded and increasing function of a supermartingale is again a supermartingale; therefore, from (38) and (30), we deduce that

(39) the process
$$\eta_l(\cdot \wedge \tau_{\lambda}^X)$$
 is a supermartingale.

It follows then, from (39) and the optional sampling theorem, that we have

(40)
$$Q_l(x) = \eta_l(0) \ge \mathbf{E} \eta_l(\tau \wedge \tau_\lambda^X) = \mathbf{E} Q_l(X(\tau \wedge \tau_\lambda^X)) \ge \mathbf{E} u(X(\tau \wedge \tau_\lambda^X))$$

for arbitrary $X(\cdot) \in \mathscr{A}(x)$ and $\tau \in \mathscr{S}_X$. Taking the supremum in (40) over $X(\cdot) \in \mathscr{A}(x)$ and $\tau \in \mathscr{S}_X$, we obtain in conjunction with (35) the inequality

$$Q_l(x) \geq V(x)$$
,

as promised. \Box

Acknowledgment. We are grateful to Vic Pestien for stimulating discussions on this problem.

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