

# OPTIMAL STOPPING AND A MARTINGALE APPROACH TO THE PENALTY METHOD

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**1. Introduction.** Let  $(\Omega, \mathcal{F}, P)$  be a complete probability space equipped with an increasing, right-continuous family of complete sub- $\sigma$ -fields  $(\mathcal{F}_t)_{t \geq 0}$  such that  $\mathcal{F} = \bigvee_{t \geq 0} \mathcal{F}_t$ . Let  $X$  be an  $(\mathcal{F}_t)$ -adapted and right-continuous process such that

$$(1) \quad E \left[ \sup_t X_t^+ \right] < \infty, \quad E[X_t^-] < \infty \quad (t \geq 0)$$

where  $x^+ = \max(x, 0)$  and  $x^- = \max(-x, 0)$ . We define the following classes of stopping times:

$$(2) \quad \begin{aligned} \bar{C}_t &= \{ \tau \mid \text{stopping time, } \tau \geq t, E[X_\tau^-] < \infty \}, \\ C_t &= \{ \tau \in \bar{C}_t \mid \text{finite} \}, \end{aligned}$$

for each  $t \geq 0$ , where  $X_\infty$  is interpreted as  $\limsup_{t \rightarrow \infty} X_t$ . A stopping time  $\sigma \in \bar{C}_0$  is said to be optimal if  $E[X_\sigma] = \sup_{\tau \in \bar{C}_0} E[X_\tau]$ . Snell's envelopes  $Y$  and  $\bar{Y}$  are defined as follows:

$$Y_t = \operatorname{ess\,sup}_{\tau \in C_t} E[X_\tau \mid \mathcal{F}_t], \quad \bar{Y}_t = \operatorname{ess\,sup}_{\tau \in \bar{C}_t} E[X_\tau \mid \mathcal{F}_t].$$

Our aim is to give some extensions of the results in optimal stopping and the penalty method presented by Fakheev [4] and Stettner and Zabczyk [9]. In Section 2, we show that the right-continuity of  $Y$  follows from that of  $X$ . This fact is not necessarily pointed out in other articles [4], [7] and [10], and it plays an important role in Sections 3 and 4. In Section 5, we introduce the generator  $A$  associated with non-Markov processes. In Section 6, we assume that  $X$  is of the form  $X_t = e^{-at}f_t + \int_0^t e^{-as}g_s ds$  for some  $f, g$ , and we give a martingale treatment of variational inequalities presented by Bensoussan and Lions [2] and approximate Snell's envelope  $Y$  by the penalty method in [9], whose results are reduced to the case of  $g = 0$ .

## 2. Right-continuous Snell's envelope.

**LEMMA 1.** *Let  $t \geq 0, \varepsilon > 0$  and  $A \in \mathcal{F}_t$  be arbitrary. Then there exists  $\tau \in C_t$  such that  $E[Y_t I_A] \leq E[X_\tau I_A] + \varepsilon$ .*

PROOF. Let  $\nu, \mu \in C_t$  and define  $\sigma = \nu I_{B^c} + \mu I_B$  with  $B = \{E[X_\nu | F_t] < E[X_\mu | F_t]\}$ . Then,

$$E[X_\sigma | F_t] = \text{ess sup} (E[X_\nu | F_t], E[X_\mu | F_t]),$$

that is, the class  $\{E[X_\nu | F_t] | \nu \in C_t\}$  is closed under the operation  $\text{sup}$ . By Proposition VI-1-1 of [6], there exists a sequence  $\tau_n \in C_t$  such that  $E[X_{\tau_n} | F_t]$  is increasing and  $\lim_{n \rightarrow \infty} E[X_{\tau_n} | F_t] = Y_t$  a.s. Since  $\{E[X_{\tau_n} | F_t] | n = 1, 2, \dots\}$  dominates the random variable  $E[X_{\tau_1} | F_t]$ , the monotone convergence theorem implies that  $\lim_{n \rightarrow \infty} E[X_{\tau_n} I_A] = E[Y_t I_A]$  for each  $A \in F_t$ . Thus the lemma is proved.

THEOREM 1. Let  $X$  be the process satisfying (1). Then there exists a right continuous supermartingale  $\hat{Y}$  majorizing  $X$  and satisfying:

- (a)  $\hat{Y}_t = Y_t = \hat{Y}_t$  a.s.,  $t \geq 0$ ,
- (b)  $\limsup_{t \rightarrow \infty} \hat{Y}_t = \limsup_{t \rightarrow \infty} X_t$ ,
- (c)  $E[\hat{Y}_t] = \sup_{\tau \in \bar{C}_t} E[X_\tau] = \sup_{\tau \in \bar{C}_t} E[X_\tau]$ ,
- (d) For any  $\sigma, \tau \in \bar{C}_0$  with  $\sigma \leq \tau$ ,  $E[|\hat{Y}_\tau|] < \infty$  and  $E[\hat{Y}_\tau | F_\sigma] \leq \hat{Y}_\sigma$ .

PROOF. We first note that  $X_t \leq Y_t \leq \bar{Y}_t$  a.s. for all  $t$ . By Lemma 1, for any  $\varepsilon > 0, t \geq s \geq 0$  and  $A \in F_s$ , there exists  $\tau \in C_t$  such that  $E[Y_t I_A] \leq E[X_\tau I_A] + \varepsilon$ . Hence,

$$E[Y_t I_A] \leq E[E[X_\tau | F_s] I_A] + \varepsilon \leq E[Y_s I_A] + \varepsilon.$$

Letting  $\varepsilon \rightarrow 0$ , we see that  $Y$  is a supermartingale. Define the process  $\hat{Y}$  by  $\hat{Y}_t = Y_{t+}$ . Clearly,  $\hat{Y}$  is a right-continuous supermartingale such that  $X_t \leq \hat{Y}_t \leq Y_t$  a.s. for all  $t$ . For each  $u \geq 0$ ,  $\sup_{s \geq u} X_s$  belongs to  $L^1$  by the inequalities  $X_u \leq \sup_{s \geq u} X_s \leq \sup_{s \geq u} X_s^+$ . For any  $t \geq u$  and  $\tau \in C_t$ ,

$$E[X_\tau | F_t] \leq E\left[\sup_{s \geq t} X_s \middle| F_t\right] \leq E\left[\sup_{s \geq u} X_s \middle| F_t\right],$$

from which  $\hat{Y}_t \leq Y_t \leq E[\sup_{s \geq u} X_s | F_t]$ . Letting  $t \rightarrow \infty$ , we see that

$$\limsup_{t \rightarrow \infty} \hat{Y}_t \leq \sup_{s \geq u} X_s < \infty \quad (u \geq 0),$$

and then, letting  $u \rightarrow \infty$ , we obtain the inequality

$$\limsup_{t \rightarrow \infty} \hat{Y}_t \leq \limsup_{t \rightarrow \infty} X_t < \infty.$$

Thus (b) follows from the fact that  $X_t \leq \hat{Y}_t$  for all  $t$ . For any  $\tau \in \bar{C}_t$  and  $A \in F_t$ ,

$$\begin{aligned} (3) \quad E[\hat{Y}_{t \wedge \tau} I_A] &= E[\hat{Y}_\tau I_{A \cap (\tau < t)}] + E[\hat{Y}_t I_{A \cap (\tau \geq t)}] \\ &\leq E[X_\tau I_{A \cap (\tau < t)}] + E[E[X_\tau | F_t] I_{A \cap (\tau \geq t)}] = E[E[X_\tau | F_t] I_A], \end{aligned}$$

that is, we have  $E[X_\tau | F_t] \leq \hat{Y}_{t \wedge \tau}$ . Applying the optional sampling theorem

to the right continuous supermartingale  $(\hat{Y}_{t \wedge \tau})$ , we obtain (d). By (d), for any  $\tau \in \bar{C}_t$ ,  $E[X_\tau | F_t] \leq E[\hat{Y}_\tau | F_t] \leq \hat{Y}_t$ . Thus we have  $\bar{Y}_t \leq \hat{Y}_t$ , which implies (a). From (a) it follows that  $\sup_{\tau \in \bar{C}_t} E[X_\tau] \leq E[\hat{Y}_t]$ . By Lemma 1, for any  $\varepsilon > 0$ , there exists  $\tau \in C_t$  such that  $E[\hat{Y}_t] \leq E[X_\tau] + \varepsilon$ . Thus we have  $E[\hat{Y}_t] \leq \sup_{\tau \in C_t} E[X_\tau] + \varepsilon$ . Letting  $\varepsilon \rightarrow 0$ , we obtain (c). Consequently, the theorem is established.

### 3. Conditions for optimality.

**THEOREM 2.**  $\sigma \in \bar{C}_0$  is optimal if and only if  $E[X_\sigma | F_t] = \hat{Y}_{t \wedge \sigma}$  for all  $t$ .

**PROOF.** The sufficiency follows immediately from Theorem 1 (c). To prove the necessity, let us show that

$$(4) \quad E[X_\sigma | F_t] = \hat{Y}_t \quad \text{on } \{t \leq \sigma\}.$$

By Theorem 1 (d),  $E[X_{\sigma \vee t} | F_t] \leq E[\hat{Y}_{\sigma \vee t} | F_t] \leq \hat{Y}_t$ , from which  $E[X_\sigma | F_t] \leq \hat{Y}_t$  on  $\{t \leq \sigma\}$ . On the other hand, set  $B = \{E[X_\sigma | F_t] < \hat{Y}_t\}$  and  $A = B \cap \{t \leq \sigma\}$ . Suppose that  $P(A) > 0$ . Then, by Lemma 1, for any  $\varepsilon$  with  $0 < \varepsilon < (E[\hat{Y}_t I_A] - E[X_\sigma I_A])$ , there exists  $\tau \in C_t$  such that

$$E[\hat{Y}_t I_A] \leq E[X_\tau I_A] + \varepsilon < E[X_\tau I_A] + E[\hat{Y}_t I_A] - E[X_\sigma I_A],$$

that is,  $E[X_\sigma I_A] < E[X_\tau I_A]$ . Define  $\rho \in \bar{C}_0$  by

$$\rho = \tau I_A + \sigma I_{A^c} = \tau I_A + \sigma I_{B^c \cap \{\sigma \geq t\}} + \sigma I_{\{\sigma < t\}}.$$

Then,  $E[X_\rho] = E[X_\tau I_A] + E[X_\sigma I_{A^c}] > E[X_\sigma I_A] + E[X_\sigma I_{A^c}] = E[X_\sigma]$ , which is a contradiction. Let  $\sigma$  be optimal. Then we have  $E[X_\sigma] = E[\hat{Y}_0] \leq E[\hat{Y}_\sigma]$  by Theorem 1, and so  $X_\sigma = \hat{Y}_\sigma$ . By (4), for any  $A \in F_t$ ,

$$\begin{aligned} E[\hat{Y}_{t \wedge \sigma} I_A] &= E[\hat{Y}_\sigma I_{A \cap \{\sigma < t\}}] + E[\hat{Y}_t I_{A \cap \{\sigma \geq t\}}] \\ &= E[X_\sigma I_{A \cap \{\sigma < t\}}] + E[E[X_\sigma | F_t] I_{A \cap \{\sigma \geq t\}}] = E[E[X_\sigma | F_t] I_A]. \end{aligned}$$

Consequently, we have  $E[X_\sigma | F_t] = \hat{Y}_{t \wedge \sigma}$ , completing the proof.

Since the process  $(\hat{Y}_t - E[\sup_s X_s^+ | F_t])$  is a negative right-continuous supermartingale, it belongs to the class (DL). By the Doob-Meyer theorem,  $\hat{Y}$  has a unique decomposition

$$\hat{Y}_t = M_t - A_t,$$

where  $M$  is a martingale and  $A$  is a predictable increasing process with  $A_0 = 0$ .

**THEOREM 3.** In order that there exists an optimal stopping time  $\sigma \in C_0$ , it is necessary and sufficient that the stopping time  $\theta = \inf \{t | X_t = M_t\}$  belongs to  $C_0$ . In this case  $\theta$  is optimal in  $C_0$ .

PROOF. By Theorem 2, we have

$$E[X_\sigma | F_t] = \hat{Y}_{t \wedge \sigma} = M_{t \wedge \sigma} - A_{t \wedge \sigma}.$$

Hence,  $(A_{t \wedge \sigma})$  is a predictable increasing martingale, and so  $A_{t \wedge \sigma} = 0$ . Letting  $t \rightarrow \infty$ , we have  $M_\sigma = \hat{Y}_\sigma = X_\sigma$ . Therefore,  $\theta \leq \sigma < \infty$ , and  $X_\theta = \hat{Y}_\theta = E[X_\sigma | F_\theta] \in L^1$ , i.e.,  $\theta \in C_0$ . Conversely, since  $X_\theta = \hat{Y}_\theta = M_\theta$ , we have  $A_\theta = 0$  and then  $\hat{Y}_{t \wedge \theta} = M_{t \wedge \theta}$ . By (3) and the definition of  $\hat{Y}$ ,

$$E[X_\theta | F_t] \leq \hat{Y}_{t \wedge \theta} \leq E\left[\sup_s X_s^+ | F_t\right].$$

Thus, it is easy to see that  $(\hat{Y}_{t \wedge \theta})$  is a uniformly integrable martingale. By the optional sampling theorem, we have

$$E[X_\theta | F_t] = E[\hat{Y}_\theta | F_t] = \hat{Y}_{t \wedge \theta}.$$

By Theorem 2,  $\theta$  is optimal in  $C_0$ .

#### 4. Existence of optimal stopping times.

THEOREM 4. Suppose that for any sequence  $\tau_n \in C_0$  increasing to  $\tau$ ,

$$(5) \quad \limsup_{n \rightarrow \infty} X_{\tau_n} \leq X_\tau \quad \text{on } \{\tau < \infty\}.$$

Then  $\gamma = \inf\{t | X_t = \hat{Y}_t\}$  is optimal in  $\bar{C}_0$ , and there exists an optimal stopping time  $\sigma \in C_0$  if and only if  $\gamma < \infty$  a.s. If, in addition,  $\lim_{t \rightarrow \infty} X_t = -\infty$ , then  $\gamma$  is optimal in  $C_0$ .

PROOF. This is proved in [7], except the last assertion, but we briefly sketch its proof. By Theorem 1 (b), for any integer  $n$ , it is possible to show that  $\tau_n = \inf\{t | X_t \geq \hat{Y}_t - 1/n\}$  is finite a.s. and  $X_{\tau_n} \geq \hat{Y}_{\tau_n} - 1/n$ . According to the same arguments as in [10, Chap. 3, Lemma 19], we can prove that  $E[\hat{Y}_{\tau_n}] = E[\hat{Y}_0]$ . Thus, it is clear that  $\tau_n \in C_0$  and

$$(6) \quad E[\hat{Y}_0] = \sup_{\tau \in C_0} E[X_\tau] \leq E[X_{\tau_n}] + 1/n.$$

Let  $\tau = \lim_{n \rightarrow \infty} \tau_n$ . Then  $\tau \leq \gamma$  a.s. and by (5) and (6),

$$E[\hat{Y}_0] = \lim_{n \rightarrow \infty} E[X_{\tau_n}] \leq E\left[\limsup_{n \rightarrow \infty} X_{\tau_n}\right] \leq E[X_\tau].$$

Clearly,  $\tau \in \bar{C}_0$  and  $X_\tau = \hat{Y}_\tau$ . Therefore,  $\tau = \gamma$  is optimal in  $\bar{C}_0$  and if  $\gamma < \infty$  a.s., then  $\gamma$  is optimal in  $C_0$ . Conversely, if there exists an optimal stopping time  $\sigma \in C_0$ , then  $X_\sigma = \hat{Y}_\sigma$  by Theorem 3 and thus  $\gamma \leq \sigma < \infty$  a.s. To prove the last assertion, let us assume that  $P(\gamma = \infty) > 0$ . By Theorem 2,  $\lim_{t \rightarrow \infty} \hat{Y}_{t \wedge \gamma} = \lim_{t \rightarrow \infty} E[X_\gamma | F_t] = X_\gamma$ . Hence,  $\lim_{t \rightarrow \infty} \hat{Y}_t = X_\gamma$  a.s. on  $\{\gamma = \infty\}$ . If  $\lim_{t \rightarrow \infty} X_t = -\infty$ , it follows from Theorem 1 (b)

that  $\lim_{t \rightarrow \infty} \hat{Y}_t = -\infty$ , which is a contradiction. Thus the theorem is established.

**5. Generator  $A$ .** For  $1 < p \leq \infty$  fixed, let  $W^p$  be the Banach space of all right-continuous,  $(F_t)$ -adapted processes  $x$  such that  $\|x\|_p = \|\sup_t |x_t|\|_{L^p} < \infty$ . We set  $T_s x(t) = E[x(t+s)|F_t]$  for each  $s \geq 0$  and  $x \in W^p$ , and define the linear operators  $\{G_\alpha\}_{\alpha>0}$  from  $W^p$  into itself by

$$G_\alpha x(t) = \int_0^\infty e^{-\alpha s} T_s x(t) ds = E \left[ \int_t^\infty e^{-\alpha(s-t)} x_s ds \middle| F_t \right].$$

Then,  $G_\alpha$  is one to one and satisfies the resolvent equation

$$(7) \quad G_\alpha - G_\beta + (\alpha - \beta)G_\alpha G_\beta = 0 \quad (\alpha, \beta > 0).$$

Indeed, interchanging the orders of integration, we obtain

$$\begin{aligned} G_\alpha G_\beta x(t) &= E \left[ \int_t^\infty e^{-\alpha(s-t)} E \left[ \int_s^\infty e^{-\beta(r-s)} x_r dr \middle| F_s \right] ds \middle| F_t \right] \\ &= E \left[ \int_t^\infty \left( \int_s^\infty e^{-\alpha(s-t)} e^{-\beta(r-s)} x_r dr \right) ds \middle| F_t \right] \\ &= E \left[ \int_t^\infty \left( \int_t^r e^{-\alpha(s-t)} e^{-\beta(r-s)} ds \right) x_r dr \middle| F_t \right] \\ &= E \left[ \int_t^\infty (\beta - \alpha)^{-1} (e^{-\alpha(r-t)} - e^{-\beta(r-t)}) x_r dr \middle| F_t \right] \\ &= (\beta - \alpha)^{-1} (G_\alpha - G_\beta) x(t), \end{aligned}$$

which implies (7). Let  $G_\alpha x(t) = 0$  for each  $t \geq 0$ . Then  $G_\beta x(t) = 0$  for all  $\beta > 0$  by (7). Hence  $T_s x(t) = 0$  for all  $s \geq 0$  by the right-continuity of the mapping  $s \rightarrow T_s x(t)$ . Thus, taking  $s = 0$ , we have  $x_t = 0$ . This implies that  $G_\alpha$  is one to one. Therefore,  $G_\alpha(W^p)$  and  $\alpha - G_\alpha^{-1}$  are independent of  $\alpha$ . Consequently, we can define the subclass  $D(A)$  of  $W^p$  and the generator  $A$  from  $D(A)$  into  $W^p$  by  $D(A) = G_\alpha(W^p)$  and  $A = \alpha - G_\alpha^{-1}$ .

**LEMMA 2.** Let  $x, c \in W^p$  and  $y \in D(A)$ . Then we have:

- (i)  $x_t = \int_0^t c_r dr$  implies  $x \in D(A)$  and  $Ax = c$ .
- (ii)  $A(e^{-\alpha \cdot} y)(t) = e^{-\alpha t} (-\alpha + A)y(t)$ .

**PROOF.** Interchanging the orders of integration, we have

$$\begin{aligned} \alpha G_\alpha x(t) &= \alpha e^{\alpha t} E \left[ \int_t^\infty e^{-\alpha s} \left( \int_0^s c_r dr \right) ds \middle| F_t \right] \\ &= \alpha e^{\alpha t} E \left[ \int_0^t \left( \int_t^\infty e^{-\alpha s} ds \right) c_r dr + \int_t^\infty \left( \int_r^\infty e^{-\alpha s} ds \right) c_r dr \middle| F_t \right] \\ &= x_t + G_\alpha c(t), \end{aligned}$$

which implies (i). Let  $y = G_\alpha x$  for  $x \in W^p$ . Integrating by parts, we obtain

$$\begin{aligned} G_\alpha(e^{-\alpha \cdot} x)(t) &= e^{\alpha t} E \left[ \int_t^\infty e^{-\alpha s} (e^{-\alpha s} x_s) ds \middle| F_t \right] \\ &= e^{\alpha t} E \left[ \left[ e^{-\alpha s} \left( - \int_s^\infty e^{-\alpha r} x_r dr \right) \right]_t^\infty + \alpha \int_t^\infty e^{-\alpha s} \left( - \int_s^\infty e^{-\alpha r} x_r dr \right) ds \middle| F_t \right] \\ &= e^{-\alpha t} G_\alpha x(t) - \alpha G_\alpha (e^{-\alpha \cdot} (G_\alpha x))(t). \end{aligned}$$

Thus,  $(\alpha G_\alpha - I)e^{-\alpha \cdot} y = G_\alpha(e^{-\alpha \cdot}(-\alpha + A)y)$ , which implies (ii).

**THEOREM 5.** *Let  $y \in D(A)$  and  $s \geq 0$ . Then,*

$$T_t y(s) - y(s) = \int_0^t T_r A y(s) dr \quad \text{for all } t \geq 0.$$

**PROOF.** Let  $y = G_\alpha x$  for  $x \in W^p$ . Integrating by parts, we obtain

$$\begin{aligned} \int_0^t T_r A G_\alpha x(s) dr &= \alpha E \left[ \int_0^t G_\alpha x(s+r) dr \middle| F_s \right] - E \left[ \int_0^t x(s+r) dr \middle| F_s \right] \\ &= E \left[ \int_s^{t+s} \alpha e^{\alpha r} \left( \int_r^\infty e^{-\alpha v} x_v dv \right) dr \middle| F_s \right] + E \left[ \int_s^{t+s} e^{\alpha r} (-e^{-\alpha r} x_r) dr \middle| F_s \right] \\ &= E \left[ \left[ e^{\alpha r} \int_r^\infty e^{-\alpha v} x_v dv \right]_s^{t+s} \middle| F_s \right] \\ &= E \left[ E \left[ \int_{t+s}^\infty e^{-\alpha(r-(t+s))} x_r dr \middle| F_{t+s} \right] \middle| F_s \right] - G_\alpha x(s). \end{aligned}$$

This completes the proof.

**COROLLARY.** *Let  $x \in D(A)$  and  $Ax \leq 0$ . Then  $x$  is a supermartingale.*

**PROOF.** The proof is immediate from Theorem 5.

**REMARK.** Let  $x \in W^p$ . Then  $x$  is a martingale if and only if  $Ax = 0$ . Indeed, the sufficiency is immediate from Theorem 5. Conversely, let  $x \in W^p$  be a martingale. Then  $x$  can be rewritten as  $x_t = E[x_\infty | F_t]$  for some  $x_\infty \in L^p$ . Hence,

$$G_\alpha x(t) = E \left[ \int_t^\infty e^{-\alpha(s-t)} E[x_\infty | F_s] ds \middle| F_t \right] = E \left[ \int_t^\infty e^{-\alpha(s-t)} x_\infty ds \middle| F_t \right] = x_t / \alpha.$$

Thus, we have  $x \in D(A)$  and  $Ax = 0$ .

**6. The penalty method.** Let  $f, g \in W^\infty$  and set  $X_t = e^{-\alpha t} f_t + \int_0^t e^{-\alpha s} g_s ds$  for  $\alpha > 0$ . Let  $U$  be the class of all adapted and right-continuous processes  $z$  such that

$$(8) \quad e^{-\alpha \cdot} z \in W^\infty \quad \text{and} \quad \lim_{t \rightarrow \infty} e^{-\alpha t} z_t = 0 \quad \text{a.s.},$$

$$(9) \quad f_t \leq z_t \quad \text{for all } t,$$

$$(10) \quad \left( e^{-\alpha t} z_t + \int_0^t e^{-\alpha s} g_s ds \right) \text{ is a supermartingale.}$$

We next consider the penalized problem, defined as follows: to find the solution  $z^\varepsilon \in W^\infty$  of the following equation

$$(11) \quad (\alpha - A)z^\varepsilon - \varepsilon^{-1}(f - z^\varepsilon)^+ = g, \quad \varepsilon > 0.$$

Then we can obtain the following theorem.

**THEOREM 6.** *The solution  $z^\varepsilon$  of (11) converges to the minimal element  $z^*$  of  $U$  almost surely for each  $t$  as  $\varepsilon \downarrow 0$  and*

$$(12) \quad z^*(t) = \operatorname{ess\,sup}_{\tau \in \bar{C}_t} E \left[ e^{-\alpha(\tau-t)} f_\tau + \int_t^\tau e^{-\alpha(s-t)} g_s ds \mid F_t \right].$$

Furthermore, if  $f$  satisfies (5), then  $\zeta = \inf \{t \mid z^*(t) = f_t\}$  is an optimal stopping time in  $\bar{C}_0$  with respect to  $X$ .

For the proof, we need the following lemmas.

**LEMMA 3.** *Equation (11) has a unique solution  $z^\varepsilon \in D(A)$ .*

**PROOF.** Let  $x \in W^\infty$ , and define  $z = T_\varepsilon x$  by

$$z(t) = E \left[ \int_t^\infty e^{-(\alpha+\varepsilon^{-1})(s-t)} (g_s + \varepsilon^{-1} f \vee x(s)) ds \mid F_t \right].$$

Then  $T_\varepsilon$  maps  $W^\infty$  into itself. Moreover, for  $z_i = T_\varepsilon x_i$  with  $x_i \in W^\infty$  ( $i = 1, 2$ ), we have

$$\begin{aligned} |z_1(t) - z_2(t)| &\leq \int_0^\infty e^{-(\alpha+\varepsilon^{-1})s} \varepsilon^{-1} E[|f \vee x_1(s+t) - f \vee x_2(s+t)| \mid F_t] ds \\ &\leq (\alpha\varepsilon + 1)^{-1} E[\sup_r |x_1(r) - x_2(r)| \mid F_t]. \end{aligned}$$

Thus,  $\|z_1 - z_2\|_\infty \leq (\alpha\varepsilon + 1)^{-1} \|x_1 - x_2\|_\infty$ , and so the map  $T_\varepsilon$  is a contraction. A fixed point  $z^\varepsilon$  of  $T_\varepsilon$  satisfies

$$z^\varepsilon(t) = E \left[ \int_t^\infty e^{-(\alpha+\varepsilon^{-1})(s-t)} (g_s + \varepsilon^{-1}(f - z^\varepsilon)^+(s) + \varepsilon^{-1} z^\varepsilon(s)) ds \mid F_t \right].$$

By virtue of Lemma 1 of [9], this equality is equivalent to

$$(13) \quad z^\varepsilon = G_\alpha(g + \varepsilon^{-1}(f - z^\varepsilon)^+),$$

which completes the proof.

Let  $V_\varepsilon$  be the class of all progressively measurable processes  $v = (v_t)$  satisfying the inequalities  $0 \leq v_t \leq \varepsilon^{-1}$  for all  $t$ . For each  $v \in V_\varepsilon$ , we define

$$J_t(v) = E \left[ \int_t^\infty \exp \left( - \int_t^s \alpha + v_r dr \right) (g_s + v_s f_s) ds \middle| F_t \right].$$

Then we can obtain the following lemma:

LEMMA 4. Let  $v^\varepsilon(t) = \varepsilon^{-1}$  if  $z^\varepsilon(t) \leq f_t$ , and  $v^\varepsilon(t) = 0$  if  $z^\varepsilon(t) > f_t$ . Then we have

$$(14) \quad z^\varepsilon(t) = J_t(v^\varepsilon) = \operatorname{ess\,sup}_{v \in V_\varepsilon} J_t(v).$$

PROOF. By virtue of Lemma 1 of [9] and (13),

$$\begin{aligned} z^\varepsilon(t) &= E \left[ \int_t^\infty \exp \left( - \int_t^s \alpha + v_r dr \right) (g_s + \varepsilon^{-1}(f - z^\varepsilon)^+(s) + v_s z^\varepsilon(s)) ds \middle| F_t \right] \\ &= J_t(v) + E \left[ \int_t^\infty \exp \left( - \int_t^s \alpha + v_r dr \right) (v_s(z^\varepsilon - f)(s) + \varepsilon^{-1}(f - z^\varepsilon)^+(s)) ds \middle| F_t \right]. \end{aligned}$$

For any  $v \in V_\varepsilon$ , we have  $v(z^\varepsilon - f) + \varepsilon^{-1}(f - z^\varepsilon)^+ \geq 0$ , and also  $v^\varepsilon(z^\varepsilon - f) + \varepsilon^{-1}(f - z^\varepsilon)^+ = 0$ . Thus we obtain (14).

LEMMA 5. Let  $z \in U$  and  $g \in W^\infty$ . Then

$$\operatorname{ess\,sup}_{v \in V_\varepsilon} J'_t(v) \leq z_t \quad \text{for all } t,$$

$$\text{where } J'_t(v) = E \left[ \int_t^\infty \exp \left( - \int_t^s \alpha + v_r dr \right) (g_s + v_s z_s) ds \middle| F_t \right].$$

PROOF. We denote  $y_t = e^{-\alpha t}(z_t - G_\alpha g(t))$ . Since  $(y_t)$  can be rewritten as

$$y_t = e^{-\alpha t} z_t + \int_0^t e^{-\alpha s} g_s ds - E \left[ \int_0^\infty e^{-\alpha s} g_s ds \middle| F_t \right],$$

$(y_t)$  is a supermartingale such that  $\lim_{t \rightarrow \infty} y_t = 0$ . By virtue of Lemma 4 of [9], we have

$$E \left[ \int_t^\infty \exp \left( - \int_t^s v_r dr \right) (v_s y_s) ds \middle| F_t \right] \leq y_t.$$

Also, by virtue of Lemma 1 of [9],  $G_\alpha g$  can be rewritten as

$$G_\alpha g(t) = E \left[ \int_t^\infty \exp \left( - \int_t^s \alpha + v_r dr \right) (g_s + v_s G_\alpha g(s)) ds \middle| F_t \right].$$

Hence,

$$\begin{aligned} J'_t(v) - G_\alpha g(t) &= e^{\alpha t} E \left[ \int_t^\infty \exp \left( - \int_t^s v_r dr \right) (v_s y_s) ds \middle| F_t \right] \\ &\leq e^{\alpha t} y_t = z_t - G_\alpha g(t), \end{aligned}$$

which completes the proof.

PROOF OF THEOREM 6. By virtue of Theorem 1, the right hand side



of (12) admits the right-continuous modification, denoted by  $z'$ . Right-continuous Snell's envelope  $\hat{Y}$  of  $X$  is of the form

$$\hat{Y}_t = \operatorname{ess\,sup}_{\tau \in \bar{C}_t} E \left[ e^{-\alpha\tau} f_\tau + \int_0^\tau e^{-\alpha s} g_s ds \mid F_t \right] = e^{-\alpha t} z'_t + \int_0^t e^{-\alpha s} g_s ds.$$

By using Theorem 1, it is easy to check that  $z'$  belongs to  $U$ . For any  $z \in U$ , by (8)–(10),

$$\hat{Y}_t \leq \operatorname{ess\,sup}_{\tau \in \bar{C}_t} E \left[ e^{-\alpha\tau} z_\tau + \int_0^\tau e^{-\alpha s} g_s ds \mid F_t \right] \leq e^{-\alpha t} z_t + \int_0^t e^{-\alpha s} g_s ds.$$

This implies that  $z'$  is a minimal element of  $U$ . By (9), (14) and Lemma 5,

$$\begin{aligned} z^\varepsilon(t) &= \operatorname{ess\,sup}_{v \in V_t} E \left[ \int_t^\infty \exp \left( - \int_t^s \alpha + v_r dr \right) (g_s + v_s f_s) ds \mid F_t \right] \\ &\leq \operatorname{ess\,sup}_{v \in V_\varepsilon} E \left[ \int_t^\infty \exp \left( - \int_t^s \alpha + v_r dr \right) (g_s + v_s z'^s) ds \mid F_t \right] \leq z'_t, \end{aligned}$$

and  $z^\varepsilon(t)$  is increasing as  $\varepsilon \downarrow 0$ . Thus we can define  $z^*(t) = \lim_{\varepsilon \downarrow 0} z^\varepsilon(t)$  a.s., and we show that  $z^*$  belongs to  $U$ . By Lemma 2,

$$\begin{aligned} A \left( e^{-\alpha t} z^\varepsilon(t) + \int_0^t e^{-\alpha s} g_s ds \right) &= -e^{-\alpha t} (\alpha - A) z^\varepsilon(t) + e^{-\alpha t} g_t \\ &= e^{-\alpha t} (-\varepsilon^{-1} (f - z^\varepsilon)^+(t)) \leq 0. \end{aligned}$$

Hence, by Corollary to Theorem 5,  $\left( e^{-\alpha t} z^\varepsilon(t) + \int_0^t e^{-\alpha s} g_s ds \right)$  is a supermartingale. Thus, by the monotone convergence theorem and Theorem 16 of [5, Chap. VI], it is easily seen that  $\left( e^{-\alpha t} z^*(t) + \int_0^t e^{-\alpha s} g_s ds \right)$  is a right-continuous supermartingale, i.e.,  $z^*$  satisfies (10). By the inequalities  $z^\varepsilon \leq z^* \leq z'$ ,  $z^*$  satisfies (8). By (11), it is clear that  $G_\alpha(f - z^\varepsilon)^+ = \varepsilon(z^\varepsilon - G_\alpha g) \leq \varepsilon(z' - G_\alpha g)$ , which converges to zero as  $\varepsilon \downarrow 0$ . Hence, by the monotone convergence theorem, we have  $G_\alpha(f - z^*)^+ = 0$ , which implies that  $z^*$  satisfies (9). Consequently,  $z^* \in U$  and (12) follows from the minimality of  $z'$ . Finally,  $\hat{Y}$  can be rewritten as

$$\hat{Y}_t = e^{-\alpha t} z^*(t) + \int_0^t e^{-\alpha s} g_s ds.$$

Therefore, we have  $\zeta = \inf \{t \mid X_t = \hat{Y}_t\}$ , which is optimal in  $\bar{C}_0$  by Theorem 4. The theorem is established.

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