Introducing a signal processes
An optimal stopping problem
Maximum principle for singular control

A universal signal process for control and stopping problems

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A representation for quasi-left continuous processes

- Let $X=(X_t)_{t\geq 0}$ be a right continuous process such that $\{X_\tau|\ \tau\in\mathcal{T}\}$ is a uniformly integrable family of random variables and $X_{+\infty}=0$.
- ► The process X is quasi-left continuous so for any $(\tau_n)_{n\geq 1}$ announcing a stopping time τ

$$\limsup_{n\to\infty} X_{\tau_n} = X_{\tau}$$
 ; $P-a.s.$

► Therefore, the process *X* can only have discontinuities at completely inaccessible stopping times.

A representation for quasi-left continuous processes

Let $c\left(t,I\right)$ to be an increasing random function of I then there exists a progressively measurable **signal process** $\left(\xi_{t}\right)_{t>0}$ such that

$$X_{\tau} = E\left[\int_{\tau}^{\infty} c\left(t, \sup_{v \in [\tau, t)} \xi_{v}\right) r e^{-rt} dt \middle| \mathcal{F}_{\tau}\right] \qquad ; \qquad \forall \tau \in \mathcal{T}.$$

The existence and uniqueness of such a representation has been established by Bank & El Karoui.

Remark The term re^{-rt} dt may be replace with a finite random measure $\mu\left(dt\right)$ with full support

Interpretation of signal process

- 1. Deterministic case If c(t, l) = -l and X(t) is a deterministic function then ξ is the derivative of the largest convex minorant of X. Equivalently, ξ is the derivative of X_{**} .
- **2. Stochastic case** The process *Y* defined via

$$Y_{\tau} = E\left[\left.\int_{\tau}^{\infty} c\left(t, \sup_{v \in [0,t)} \xi_{v}\right) r e^{-rt} \, dt \right| \mathcal{F}_{\tau}\right] \qquad ; \qquad \forall \tau \in \mathcal{T}.$$

is the largest semimartingale which does not exceed X and with drift that can be expressed as $c(t, A_t) re^{-rt} dt$ for an increasing process A. That is the flat-off condition is satisfied

$$\int_0^\infty (Y_t - X_t) \, dA_t = 0$$

Interpretation of signal process

3. Let c(t, l) = -l and let $Z = (Z_t)_{t \ge 0}$ be a Markov process such that the process X can be expressed as X = u(Z) where u is an excessive function. Then ξ coincides with the subadditive operator

$$\underline{D}u\left(x\right)\triangleq\inf\frac{u\left(x\right)-E_{X}\left[\left.u\left(X_{T}\right)\right|\tau< T_{r}\right]}{P_{X}\left[\tau=T_{r}\right]}.$$

where $T_r \sim \exp(r)$. The inf is taken over exit times from open neighbourhoods of x.

4. Consider a family of auxiliary stopping problems indexed by $l \in \mathbb{R}$

$$Y^{I}\left(\sigma\right) = \operatorname{ess}\sup_{ au\in\mathcal{T}\left(\left[\sigma,+\infty\right]\right)}E\left[X_{ au}+\int_{\sigma}^{ au}c\left(t,I\right)r\mathrm{e}^{-rt}\,dt\bigg|\,\mathcal{F}_{\sigma}\right]$$

For $(t,\omega)\in\mathbb{R} imes\Omega$ then

$$\xi\left(\omega,t\right)=\sup\left\{ I\in\mathbb{R}|\:Y^{I}\left(\omega,t\right)=X\left(\omega,t\right)\right\} .$$

Optimal stopping using Snell envelope

The Snell envelope is defined as

$$U_{\sigma} = \operatorname{ess} \sup_{\tau \in \mathcal{T}([\sigma,\infty])} E[X_{\tau} | \mathcal{F}_{\sigma}] \quad ; \quad \forall \sigma \in \mathcal{T}.$$

Theorem

Consider the Doob-Meyer decomposition of the Snell envelope U=M-A into a uniformly integrable martingale M and a predictable increasing process A with $A_0=0$. Then

$$\underline{T} = \inf \{ t \ge 0 | X_t = U_t \}$$
 ; $\overline{T} = \inf \{ t \ge 0 | A_t > 0 \}$

are the smallest and largest stopping times which attain the supremum in

$$V = \sup_{\tau \in \mathcal{T}} E\left[X_{\tau}\right].$$

Optimal stopping using a signal process

Theorem (Bank & Föllmer 2005) The stopping times

$$\underline{T} = \inf \left\{ \left. t \ge 0 \right| \xi_t \ge 0 \right\} \qquad ; \qquad \quad \overline{T} = \inf \left\{ \left. t \ge 0 \right| \xi_t > 0 \right\}$$

are the smallest and largest stopping times which attain the supremum in

$$V = \sup_{ au \in \mathcal{T}} E[X_{ au}].$$

Connecting the signal and the Snell envelope

To connect these two theorems consider the process

$$\zeta_s \triangleq \sup_{v \in [0,s)} \xi_v \vee 0.$$

and the supermartingale V defined via

$$V_t \triangleq E\left[\left.\int_t^\infty \zeta_s r e^{-rs} \, ds\right| \mathcal{F}_t\right] = E\left[\left.\int_0^\infty \zeta_s r e^{-rs} \, ds\right| \mathcal{F}_t\right] - \int_0^t \zeta_s r e^{-rs} \, ds$$

then $V \geq X$ because

$$V_t \triangleq E\left[\left.\int_t^\infty \zeta_s r \mathrm{e}^{-rs} \, ds\right| \mathcal{F}_t\right] \geq E\left[\left.\int_t^\infty r \mathrm{e}^{-rs} \sup_{v \in [t,s)} \xi_s \, ds\right| \mathcal{F}_t\right] = X_t$$

so V dominates the Snell envelope of X.

Connecting the signal and the Snell envelope

However, $V_t = U_t$ for all $t \in \left[0, \overline{T}\right)$ because for $t \leq \overline{T}$

$$V_{t} = E\left[\int_{t}^{\overline{T}} \zeta_{s} r e^{-rs} ds + \int_{\overline{T}}^{\infty} \zeta_{s} r e^{-rs} ds \middle| \mathcal{F}_{t}\right]$$

$$= E\left[\int_{\overline{T}}^{\infty} r e^{-rs} \sup_{v \in [0,s)} \zeta_{v} ds \middle| \mathcal{F}_{t}\right]$$

$$= E\left[E\left[\int_{\overline{T}}^{\infty} r e^{-rs} \sup_{v \in [0,s)} \zeta_{v} ds \middle| \mathcal{F}_{\overline{T}}\right] \middle| \mathcal{F}_{t}\right]$$

$$= E\left[X_{\overline{T}} \middle| \mathcal{F}_{t}\right] \leq U_{t}.$$

Link to a BSDE

Recall $\zeta_s = \sup_{v \in [0,s)} \xi_v \lor 0$ then the supermartingale V

$$V_{ au} = E\left[\left.\int_{ au}^{\infty} \zeta_{t} \, r \mathrm{e}^{-rt} \, dt \right| \mathcal{F}_{ au}
ight]$$

solves the BSDE

$$-dY_t = \zeta_t r e^{-rt} dt - Z_t dW_t$$
 ; $\lim_{t \to +\infty} Y_t = 0$ $Y_t \ge X_t \ \forall t \ge 0 \ P$ -a.s. $0 = \int_0^\infty (Y_t - X_t) \ d\zeta_t$

A BSDE for the Snell envelope

Let $Y \in \mathbb{S}^1$, $\int Z \, dW$ is a uniformly integrable martingale, K is an increasing process with $K_{0-} = 0$. If the triplet (Y, Z, K) satisfies

$$-dY_t = -Z_t dW_t + dK_t$$
 ; $\lim_{t \to +\infty} Y_t = 0$ $Y_t \geq X_t \ \forall t \geq 0 \ P$ -a.s. $0 = \int_0^\infty (Y_t - X_t) \ dK_t$

Then, the process Y has the following explicit representation for any t > 0

$$Y_{t} = \operatorname{ess\,sup}_{ au \in \mathcal{T}} E\left[\left.X_{ au}\right| \mathcal{F}_{t}\right].$$

Example of a perpectual option

- ► The asset price denoted $P = (P_t)_{t \geq 0}$ is an exponential Lévy process: $P_t = P_0 e^{Y_t}$
- ▶ The payoff of an American put option with strike k > 0 is

$$X_t^k = e^{-rt} \left(k - P_t \right)^+.$$

▶ Let $K_t = P_t/\alpha$ where $\alpha \in (0,1)$ is defined via

$$\alpha \triangleq E\left[\int_0^\infty \inf_{0 \le v \le t} e^{Y_v} r e^{-rt} dt\right] = E\left[\inf_{0 \le v \le T_r} e^{Y_v}\right]$$

where $T_r \sim \exp(r)$ independent of the filtration \mathbb{F} .

► The price process *P* has the representation

$$e^{-r\tau}P_{ au} = E\left[\left.\int_{ au}^{\infty} re^{-rt}\inf_{v\in[au,t)}K_{v}\;dt\right|\mathcal{F}_{ au}
ight]$$

Example of a perpectual option

It can be directly verified that this process $K_t = P_t/\alpha$ satisfies the required representation:

$$E\left[\int_{\tau}^{\infty} r e^{-rt} \inf_{v \in [\tau, t)} K_{v} dt \middle| \mathcal{F}_{\tau}\right]$$

$$= \frac{e^{-r\tau} P_{\tau}}{\alpha} E\left[\int_{\tau}^{\infty} r e^{-r(t-\tau)} \inf_{v \in [\tau, t)} e^{(Y_{v} - Y_{\tau})} dt \middle| \mathcal{F}_{\tau}\right]$$

$$= \frac{e^{-r\tau} P_{\tau}}{\alpha} E\left[\int_{0}^{\infty} r e^{-rs} \inf_{v \in [0, s)} e^{Y_{v}} dt\right]$$

$$= \frac{e^{-r\tau} P_{\tau}}{\alpha} E\left[\inf_{v \in [0, T_{r})} e^{Y_{v}}\right] = e^{-r\tau} P_{\tau}$$

Example of a perpectual option

Let $\xi_t^k = k - K_t$ then the payoff function can be represented as

$$X_{\tau}^{k} = E\left[\int_{\tau}^{\infty} re^{-rt} \sup_{v \in [\tau, t)} \xi_{v}^{k} dt \middle| \mathcal{F}_{\tau}\right].$$

Therefore, for any k > 0 the smallest and largest stopping times which attain

$$\sup_{\tau \in \mathcal{T}} E\left[e^{-r\tau}\left(k - P_{\tau}\right)\right]$$

are

$$\underline{T}^k = \inf \{ t \ge 0 | P_t \le \alpha k \}$$
; $\overline{T}^k = \inf \{ t \ge 0 | P_t < \alpha k \}$.

Provisional example of a perpectual option

Let the asset price denoted $P=(P_t)_{t\geq 0}$ is a Lévy process. The payoff function of an American option $G\left(x\right)$ is such that for a continuous function g

$$G(x) = E_{x} \left[\int_{0}^{\infty} g(P_{t}) r e^{-rt} dt \right].$$

In this case the signal coincides with the solution to the (non-standard) stopping problem

$$\underline{D}G(x) \triangleq \inf_{n \geq 1} \frac{E_{x} \left[\int_{0}^{T_{n}} e^{-rs} g(P_{s}) ds \middle| T_{n} < T_{r} \right]}{E_{x} \left[\int_{0}^{T_{n}} e^{-rs} ds \middle| \right]}.$$

where the inf is taken over exit times from open neighbourhoods of x. A different approach to this problem due to Deligiannidis et. al. (2009) suggests that if g is monotone then it may be possible to validate that

$$\underline{D}u(x) = g(x) \quad \forall x \in \mathcal{D}.$$

Provisional example of a perpectual option

To identify $\underline{D}u\left(x\right)$ we consider a transformation similar to that used for a different purpose by Hobson & Klimmek (2010). Denoted τ_y the exit times from sets of the form $(-\infty,y)$ then

$$E_{x}\left[\left.G\left(X_{\tau_{y}}\right)\right|\tau_{y} < T_{r}\right] = G\left(y\right)E_{x}\left[e^{-r\tau_{x}}\right]$$

so letting

$$p(y) \triangleq E_{x}\left[e^{-r\tau_{x}}\right], \qquad \widetilde{G}(x,y) \triangleq \frac{G(x)}{1-p(y)}, \qquad f(y) \triangleq \frac{G(y)p(y)}{1-p(y)}$$

If $\widetilde{G} \in C^{2,2}$ and \widetilde{G}_x and \widetilde{G}_y are both increasing (resp. decreasing) then $\underline{D}G(x)$ coincides with

$$v\left(x\right) \triangleq \inf_{y \geq 0} \left\{ \widetilde{G}\left(x, y\right) - \frac{G\left(y\right) p\left(y\right)}{1 - p\left(y\right)} \right\}.$$

That is $\underline{D}G(x)$ is the \widetilde{G} -concave dual of f(y).

A singular control problem

$$V\left(x
ight) = \inf_{ heta \in \mathcal{A}^2} J\left(heta
ight)$$
 $J\left(heta
ight) = E\left[\int_0^\infty c\left(t, X_t^ heta
ight) dt + \int_0^\infty k_t^1 d heta_t^1 + \int_0^\infty k_t^2 d heta_t^2
ight]$

1. The filtration
$$\mathbb{F}$$
 is assumed to be continuous.

- **2.** Intervention costs k^1 and k^2 are positive processes.
- **3.** Let $M=(M_t)_{t\geq 0}$ be a continuous martingale. The state process $X^{\theta}=(X^{\theta}_t)_{t\geq 0}$ is determined by the SDEs

$$dX_{t}^{\theta}=\left(a_{t}+b_{t}X_{t}^{\theta}\right)\,d\left[M\right]_{t}+\left(c_{t}+d_{t}X_{t}^{\theta}\right)\,dM_{t}+d\theta_{t}^{1}-d\theta_{t}^{2}$$

where $(a_t, b_t, c_t, d_t)_{t>0}$ are predictable bounded processes.

The set of controls

Take
$$heta = \left(heta^1, heta^2
ight)$$
 where $heta^i \in \mathcal{A}$ and

$$\mathcal{A} \triangleq \{ \vartheta : \mathbb{R}_+ \times \Omega \to \mathbb{R}_+ | \vartheta \text{ non-decreasing, RCLL, } \vartheta_{0-} = 0 \}$$
.

No additional constraints need be placed on the set of controls.

- 1. Under some additional assumptions on c(t, x), taking $\theta^2 \equiv 0$ retrieves the problem studied by Bank (2005).
- 2. Taking the martingale M to be Brownian motion retrieves a class of problems studied by Cadenillas & Haussmann (1994) and Bahlali & Merzerdi (2005).

Assumptions

- **1.** For any $\theta_t \equiv x \in R$ the process $c(t, X_t^x)$ is progressively measurable and $c(\cdot, X^x) \in L^1(P \otimes dt)$.
- **2.** The process $c^*\left(t\right)=\inf_{\theta\in\mathbb{R}}c\left(t,X_t^{\theta}\right)$ is such that $c^*\in L^1\left(P\otimes dt\right)$.
- **3.** For fixed $(t, \omega) \in R_+ \times \Omega$ the mapping $\theta \longmapsto c(t, \omega, \theta)$ is a convex functional mapping taking values in $\overline{\mathbb{R}}$.
- **4.** The intervention costs $k=\left(k^1,k^2\right)$ are such that k^i for i=1,2 are positive processes such that the family of random variables $\left\{k_{\tau}^i \left| \tau \right. \text{ is a stopping time} \right.\right\}$ is uniformly integrable and $\lim_{t \to \infty} k_t^i = 0$.

Comment on Assumption 3

Bank substituted assumption 3 was substituted with:

3a. For fixed $(t,\omega) \in R_+ \times \Omega$ the function $c(t,\omega,\theta)$ only depends on $\theta_t(\omega)$ so the mapping $\theta \longmapsto c(t,\omega,\theta)$ is strictly convex function mapping $R_+ \longmapsto R$ with a continuous first derivative $c_\theta(t,\omega,\theta) \triangleq \frac{\partial}{\partial \theta} c(t,\omega,\theta)$ such that

$$c_{\theta}\left(t,\omega,+\infty\right)=+\infty$$
 ; $c_{\theta}\left(t,\omega,-\infty\right)=-\infty$.

This assumption restricts us to use $b = d \equiv 0$.

Directional derivative of the running cost functional

Fix $(t,\omega) \in \mathbb{R}_+ \times \Omega$ then the functional $\theta \mapsto c\left(t,X_t^{\theta}\left(\omega\right)\right)$ has a directional derivative at $\theta \in \mathcal{V}$ in the direction μ if $c\left(t,X_t^{\theta}\right) < +\infty$ and the limit

$$\nabla c_{t}\left(\theta;\mu\right) = \lim_{\lambda \downarrow 0} \frac{c\left(t,X_{t}^{\theta+\lambda\mu}\right) - c\left(t,X_{t}^{\theta}\right)}{\lambda}$$

exists, taking values in $\overline{\mathbb{R}}$.

For arbitrary $\eta \in \mathcal{A}^2$ the directional derivative of $c\left(t,X_t^{\theta}\left(\omega\right)\right)$ is

$$\nabla c_t(\theta;\eta) = -c_X(t,X_t^\theta)z_t$$

where z is the solution to the following SDE

$$dz_t = b_t z_t dA_t + d_t z_t dM_t + d\eta_t^1 - d\eta_t^2$$
; $z_0 = 0$

Directional derivative of the running cost functional

In fact,

$$abla c_t\left(\theta;\eta\right) = -c_X\left(t,X_t^{\theta}\right)\Phi_t\int_0^t rac{1}{\Phi_s}\,d\left(\eta_s^1-\eta_s^2\right).$$

where

$$d\Phi_t = b_t \Phi_t d[M]_t + d_t \Phi_t dM_t$$
 ; $\Phi_0 = 1$.

The solution to this SDE coincides with

$$\frac{\partial X^0(t,\omega,x)}{\partial x}$$

where $X^{\theta}\left(t,\omega,x\right)=X^{\theta}_{t}$ with $X^{\theta}_{0}=x$. Recall that

$$dX_{t}^{\theta} = \left(a_{t} + b_{t}X_{t}^{\theta}\right) d\left[M\right]_{t} + \left(c_{t} + d_{t}X_{t}^{\theta}\right) dM_{t} + d\theta_{t}^{1} - d\theta_{t}^{2}$$

A conditional expectation which is a subgradient

Denote by $\nabla J_i\left(\theta\right)_t$ for i=1,2 the continuous adapted process with the property that for each stopping time τ

$$\nabla J_{1}\left(\theta\right)_{\tau}=e^{-r\tau}k_{\tau}^{1}+p_{\tau}^{\theta}\quad;\quad\nabla J_{2}\left(\theta\right)_{\tau}=e^{-r\tau}k_{\tau}^{2}-p_{\tau}^{\theta}$$

where

$$-p_t^{\theta} \triangleq \frac{1}{\Phi_t} E\left[\int_t^T c_X\left(s, X_s^{\theta}\right) \Phi_s \ ds \middle| \mathcal{F}_t\right].$$

Then, for each $\theta \in A^2$ the process $\nabla J(\theta)$ is a subgradient of $J(\theta)$ i.e.

$$J(\varphi) - J(\theta) \ge E \left[\int_0^T \nabla J(\theta)_t \ d(\varphi_t - \theta_t) \right].$$

for all $\varphi \in \mathcal{A}^2$.

The objective function is Gâteaux differentiable

Lemma

The functional $J:\mathcal{V}^2\to\overline{\mathbb{R}}$ is Gâteaux differentiable with Gâteaux differential given by

$$abla J\left(heta
ight) = \left(
abla J_1\left(heta
ight),
abla J_2\left(heta
ight)
ight)'$$

Lemma

Ekeland & Temam Proposition 2.2.1 If $J: \mathcal{V}^2 \to \overline{\mathbb{R}}$ is a proper, convex and Gâteaux differentiable functional then the following two statements are equivalent

- **1.** $\theta^* \in A^2$ is a solution to $\inf_{\theta \in A^2} J(\theta)$
- **2.** For all $\eta \in A^2$

$$E\left[\int_{0}^{\infty} J'\left(\theta^{*}\right)_{t} d\left(\eta_{t} - \theta_{t}^{*}\right)\right] \geq 0.$$

where J' denotes the Gâteaux differential.

Necessary and sufficient conditions

Theorem

The control $\theta^* \in \mathcal{A}^2$ is optimal for the singular stochastic control problem if and only if both

$$\nabla J(\theta^*)_s \ge 0$$
 ; $\forall s \ge 0$

and

$$\int_0^\infty \nabla J(\theta^*)_s \ d\theta_s^* = 0 \qquad P\text{-a.s.}$$

The optimal control for the control problem can be characterised in the usual way:

- 1. 'Initial jump' if $g_0^i p_0 > k_0^i$ for any i, then $\triangle \theta_{0+}^i > 0$.
- **2.** 'No action region' the flat-off condition ensures that $d\theta_t^i = 0$ on the set $\{t : k_t^i > p_t g_t^i\}$ for each i = 1, 2.
- 3. 'Local time' the measure $d\theta^i$ has the same points of increase as the local time of p_t , on the curve k_t^i/g_t^i .

Conditional expectations are linear BSDEs

For $\theta \in \text{dom}(J)$ the conditional expectation

$$-p_{t}^{\theta} = \frac{1}{\Phi_{t}} E\left[\int_{t}^{\infty} c_{X}\left(s, X_{s}^{\theta}\right) \Phi_{s} ds \middle| \mathcal{F}_{t} \right].$$

is the first component to the solution of a linear BSDE. That is the process p^{θ} solves

$$dp_t^{\theta} = \left[c_X \left(t, X_t^{\theta} \right) - b_t p_t^{\theta} - d_t q_t^{\theta} \right] dt + q_t^{\theta} dM_t + dL_t^{\theta}.$$

with terminal condition $\lim_{T\to\infty} p_T^\theta=0$ under the assumption that $q^\theta\in L\left(M\right);\;\int q^\theta\;dM$ and L^θ are uniformly integrable martingales with $\left\lceil M,L^\theta\right\rceil=0.$

A dual problem in terms of a BSDE

Take $\widehat{\theta} \in \mathrm{dom}\,(J)$ and let $(\widehat{\rho},\widehat{q})$ denote the corresponding solution to the adjoint BSDE. Then, $\widehat{\theta} \in \mathcal{A}$ is optimal for the singular stochastic control problem if and only if $(\widehat{\rho},\widehat{q},\widehat{\theta})$ is a solution to the following BSDE with obstacle constraint

$$\begin{array}{rcl} dp_t &=& \left[c_X\left(t,X_t^\theta\right) - b_t p_t - d_t q_t\right] \, dt + q_t \, dM_t + dL_t \\ \lim_{t \to +\infty} p_t &=& 0 \\ e^{-rt} k_t^1 &\leq& p_t \leq e^{-rt} k_t^2 \quad \text{a.s.} \quad \forall t \in [0,T] \\ 0 &=& \int_0^\infty \left(e^{-rt} k_t^1 - p_t\right) \, d\theta_t^1 = \int_0^\infty \left(e^{-rt} k_t^2 + p_t\right) \, d\theta_t^2 \end{array}$$

Recall the obstacle problem linked to the stopping problem

Recall $\zeta_s = \sup_{v \in [0,s)} \xi_v \lor 0$ then the supermartingale V

$$V_{ au} = E\left[\left.\int_{ au}^{\infty} \zeta_{t} \, r \mathrm{e}^{-rt} \, dt\right| \mathcal{F}_{ au}\right]$$

solves the BSDE

$$dY_t = -\zeta_t r e^{-rt} dt + Z_t dM_t$$
 ; $\lim_{t \to +\infty} Y_t = 0$ $Y_t \ge X_t \ \forall t \ge 0 \ P$ -a.s. $0 = \int_0^\infty (Y_t - X_t) \ d\zeta_t$

Literature on this type of representation

- ► Bank (2005) 'Optimal control under a dynamic fuel constraint' SIAM J. Control Optim. 44 pp. 1529–1541.
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- ► Föllmer & Knispel (2006) 'A representation of excessive functions as expected suprema' Prob & Math Stats, 26, Fasc. 2