

PROGRESSIVE ENLARGEMENT OF FILTRATION AND CONTROL PROBLEMS FOR STEP PROCESSES

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Preliminaries

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, $\mathbb{F} = (\mathcal{F}_t)_{t \geq 0}$.

The jump measure of an \mathbb{R}^d -valued \mathbb{F} -semimartingale X is the integer-valued random measure on $\mathbb{R}_+ \times \mathbb{R}^d$

$$\mu^X(\omega, dt, dx) = \sum_{s > 0} 1_{\{\Delta X_s(\omega) \neq 0\}} \delta_{(s, \Delta X_s(\omega))}(dt, dx).$$

X is a *step process with respect to* \mathbb{F} if

$$X = \sum_{n=1}^{\infty} \xi_n 1_{[\tau_n, +\infty)},$$

where

- $(\tau_n)_n$ is a sequence of \mathbb{F} -stopping times s.t. $\tau_n \uparrow +\infty$, $\tau_n < \tau_{n+1}$ on $\{\tau_n < +\infty\}$
- $(\xi_n)_{n \geq 1}$ is a sequence of \mathbb{R}^d -valued random variables s.t. ξ_n is \mathcal{F}_{τ_n} -measurable and $\xi_n \neq 0$ if and only if $\tau_n < +\infty$.

We use the notation (X, \mathbb{F}) to mean that X is \mathbb{F} -adapted, while \mathbb{F}^X denotes the smallest right-continuous filtration such that X is adapted.

Progressive enlargement of filtrations

We consider two step processes (X, \mathbb{F}^X) and (H, \mathbb{F}^H) .

Given a σ -field \mathcal{R}^X , we denote by $\mathbb{F} = (\mathcal{F}_t)_{t \geq 0}$ the filtration \mathbb{F}^X initially enlarged by \mathcal{R}^X , i.e.

$$\mathcal{F}_t := \mathcal{R}^X \vee \mathcal{F}_t^X \quad t \geq 0.$$

Analogously, given a σ -field \mathcal{R}^H , we denote by $\mathbb{H} = (\mathcal{H}_t)_{t \geq 0}$ the filtration \mathbb{F}^H initially enlarged by \mathcal{R}^H , i.e.

$$\mathcal{H}_t := \mathcal{R}^H \vee \mathcal{F}_t^H, \quad t \geq 0.$$

We denote by $\mathbb{G} = (\mathcal{G}_t)_{t \geq 0}$ the *progressive enlargement of \mathbb{F} by \mathbb{H}* :

$$\mathcal{G}_t := \bigcap_{s > t} \mathcal{F}_s \vee \mathcal{H}_s \quad t \geq 0.$$

\mathbb{G} is the smallest right-continuous filtration containing \mathbb{F}^X , \mathbb{F}^H , \mathcal{R}^X and \mathcal{R}^H (i.e., \mathbb{F} and \mathbb{H}).

MRT in the enlargement of a step process filtration

We introduce the $\mathbb{R}^d \times \mathbb{R}^\ell$ -valued \mathbb{G} -semimartingale $Z = (X, H)$.

Theorem

(Z, \mathbb{G}) is a step process and \mathbb{G} is the smallest right-continuous filtration containing $\mathcal{R}^X \vee \mathcal{R}^H$ and such that μ^Z is optional.

Theorem

If $\mathcal{F} = \mathcal{G}_\infty$, every $Y \in \mathcal{H}_{\text{loc}}^1(\mathbb{G})$ can be represented as

$$Y = Y_0 + W * \mu^Z - W * \mathbf{v}^{\mathbb{G}, Z}$$

where $(\omega, t, x_1, x_2) \mapsto W(\omega, t, x_1, x_2)$ is a $\mathcal{P}(\mathbb{G}) \otimes \mathcal{B}(\mathbb{R}^d) \otimes \mathcal{B}(\mathbb{R}^\ell)$ -measurable function such that

$$|W| * \mu^Z \in \mathcal{A}_{\text{loc}}^+(\mathbb{G})$$

and $\mathbf{v}^{\mathbb{G}, Z}$ denotes the \mathbb{G} -dual predictable projection of μ^Z .

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Progressive enlargement by a random time

From here on we shall concentrate on the special case

$$H = 1_{[\tau, +\infty)},$$

where τ denotes a random time. We denote by $H^{p, \mathbb{F}}$ the \mathbb{F} -dual predictable projection of H . We also set

$$A_t = \mathbb{P}[\tau > t | \mathcal{F}_t] \text{ a.s., for every } t \geq 0.$$

A is a càdlàg \mathbb{F} -supermartingale, called the Azéma supermartingale.

The \mathbb{G} -dual predictable projection of H is

$$\Lambda^{\mathbb{G}} = \int_0^{\tau \wedge \cdot} \frac{1}{A_{s-}} dH_s^{p, \mathbb{F}}.$$

Quasi-left continuity in the enlarged filtration

We are interested in the following question:

if μ^X is an \mathbb{F} -quasi left continuous random measure, is it \mathbb{G} -quasi left continuous?

In general, no: intuitively, the larger filtration \mathbb{G} supports more predictable stopping times than \mathbb{F} .

Example

Let X be a homogeneous Poisson process with respect to \mathbb{F}^X and let $(\tau_n)_{n \geq 1}$ be the sequence of the jump-times of \mathbb{F}^X .

*The process X is **not quasi-left continuous in the filtration \mathbb{G} obtained enlarging \mathbb{F}^X progressively by $\tau = \frac{1}{2}(\tau_1 + \tau_2)$.***

Indeed, the jump time τ_2 of X is announced in \mathbb{G} by $(\vartheta_n)_{n \geq 1}$, $\vartheta_n := \frac{1}{n}\tau + (1 - \frac{1}{n})\tau_2$, and $\vartheta_n > \tau$ is a \mathbb{G} -stopping time for every $n \geq 1$. Hence, τ_2 is a \mathbb{G} -predictable jump-time of X .

Sufficient conditions on τ s.t. μ^X is \mathbb{G} -quasi-left continuous?

Avoidance of \mathbb{F} -stopping times and immersion property

Definition (Assumption (\mathcal{A}))

A random time τ satisfies the **avoidance of \mathbb{F} -stopping times** if $\mathbb{P}[\tau = \sigma < +\infty] = 0$ for every \mathbb{F} -stopping time σ .

Remark. If τ satisfies (\mathcal{A}), then

- the process H is \mathbb{G} -quasi-left continuous;
- $\Delta X \Delta H = 0$.

τ carries an information which is completely exogenous (nothing about τ can be inferred from the information contained in \mathbb{F}).

Definition (Assumption (\mathcal{H}))

A random time τ satisfies the **immersion property** if \mathbb{F} -martingales remain \mathbb{G} -martingales.

Remark. If X is an \mathbb{F} -step process and τ satisfies (\mathcal{H}), then $\mathbf{v}^{\mathbb{G},X} = \mathbf{v}^{\mathbb{F},X}$.

The \mathbb{G} -dual predictable projection of μ^Z under (\mathcal{A}) -(\mathcal{H})

Theorem

Let X be a step process and let τ be a random time satisfying (\mathcal{A}) -(\mathcal{H}). Then,

$$\begin{aligned} v^{\mathbb{G},Z}(\omega, dt, dx_1, dx_2) \\ = v^{\mathbb{F},X}(\omega, dt, dx_1) \delta_0(dx_2) + \delta_1(dx_2) \delta_0(dx_1) d\Lambda_t^{\mathbb{G}}(\omega). \end{aligned}$$

Remark 1. If μ^X is \mathbb{F} -quasi-left continuous and τ satisfies (\mathcal{A}) -(\mathcal{H}), then μ^Z is \mathbb{G} -quasi-left continuous.

Remark 2. One can quite easily construct random times τ satisfying (\mathcal{A}) -(\mathcal{H}), see e.g. Di Tella and Engelbert (2021).

Jacod's absolute continuity property

Let $\eta(du)$ denote the law of τ and $P_t(du)$ a regular version of its conditional distribution, i.e., for every $A \in \mathcal{B}(\mathbb{R})$, $\eta(A) := \mathbb{P}(\tau^{-1}(A))$ and $P_t(A) := \mathbb{P}(\tau \in A | \mathcal{F}_t)$.

Definition (Assumption (JAC))

*A random time τ satisfies **Jacod's absolute continuity condition** if*

$$\eta(du) \text{ is a diffused probability measure,} \quad (1)$$

$$P_t(du) \ll \eta(du). \quad (2)$$

Remark 1. Condition (1) ensures property (\mathcal{A}).

Remark 2. Random times satisfying (JAC) can be constructed following the approach presented by Jeanblanc and Le Cam (2009).

The \mathbb{G} -dual predictable projection of μ^Z under (JAC)

Theorem

Let (X, \mathbb{F}) be an \mathbb{F} -quasi left continuous step process, and τ be a random time satisfying (JAC). Then

$$\begin{aligned} & \nu^{\mathbb{G}, Z}(\omega, dt, dx_1, dx_2) \\ &= \left(1_{[0, \tau]}(\omega, t) \left(1 + \frac{W'(\omega, t, x_1)}{A_{t-}(\omega)} \right) \right. \\ & \quad \left. + 1_{(\tau, +\infty)}(\omega, t) (1 + U(\omega, t, x_1)) \right) \nu^{\mathbb{F}, X}(\omega, dt, dx_1) \delta_0(dx_2) \\ & \quad + 1_{[0, \tau]}(\omega, t) \frac{1}{A_{t-}(\omega)} dH_t^{p, \mathbb{F}}(\omega) \delta_0(dx_1) \delta_1(dx_2), \end{aligned}$$

where W' is an \mathbb{F} -predictable function such that $A_- + W' \geq 0$ and U is a \mathbb{G} -predictable function such that $1 + U \geq 0$ identically.

Remark. If μ^X is \mathbb{F} -quasi left continuous and τ satisfies (JAC), then μ^Z is \mathbb{G} -quasi left continuity.

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Background

Optimization problem for a step process (X, \mathbb{F}) , in presence of an **additional exogenous risk source that cannot be inferred from \mathbb{F}** (as the death of the investor or the default of part of the market), whose occurrence time is modeled by τ .

Setting 1

X is \mathbb{F} -quasi-left continuous and τ satisfies (\mathcal{A}) -(\mathcal{H}).

or

Setting 2

X is \mathbb{F} -quasi-left continuous and τ satisfies (JAC) (hence (\mathcal{A})).

Under Setting 1 or 2,

$$\mathbf{v}^{\mathbb{F},X}(dt, dx_1) = \phi_t^{\mathbb{F},X}(dx_1)dC_t^{\mathbb{F},X},$$

where $\phi^{\mathbb{F},X}$ is a transition probability from $(\Omega \times [0, T], \mathcal{P}(\mathbb{F}))$ into $(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$, and $C^{\mathbb{F},X} \in \mathcal{A}_{\text{loc}}^+(\mathbb{F})$ is a **continuous process**.

Under Setting 1

$$\begin{aligned}\phi_t^{\mathbb{G},X}(\mathrm{d}x_1) &= \phi_t^{\mathbb{F},X}(\mathrm{d}x_1), \\ \mathrm{d}C_t^{\mathbb{G},X} &= \mathrm{d}C_t^{\mathbb{F},X}.\end{aligned}$$

Under Setting 2

$$\begin{aligned}\phi_t^{\mathbb{G},X}(\mathrm{d}x_1) &= \frac{\kappa(t, x_1)}{\int_{\mathbb{R}^d} \kappa(t, x_1) \phi_t^{\mathbb{F},X}(\mathrm{d}x_1)} \phi_t^{\mathbb{F},X}(\mathrm{d}x_1), \\ \mathrm{d}C_t^{\mathbb{G},X} &= \int_{\mathbb{R}^d} \kappa(t, x_1) \phi_t^{\mathbb{F},X}(\mathrm{d}x_1) \mathrm{d}C_t^{\mathbb{F},X},\end{aligned}$$

where

$$\kappa(\omega, t, x_1) := 1_{[0, \tau]}(\omega, t) \left(1 + \frac{W'(\omega, t, x_1)}{A_{t-}(\omega)} \right) + 1_{(\tau, +\infty)}(\omega, t) (1 + U(\omega, t, x_1)).$$

Formulation of the problem

Consider the problem of an agent whose available information is \mathbb{F} (that is, she pursues \mathbb{F} -predictable strategies) but, for some reasons, has **only access to the market up to the occurrence of the exogenous shock event at time τ** .

The Data

(U, \mathcal{U}) is a measurable space.

$r, l : \Omega \times [0, T] \times \mathbb{R}^d \times U \rightarrow \mathbb{R}$ are $\mathcal{P}(\mathbb{F}) \otimes \mathcal{B}(\mathbb{R}^d) \otimes \mathcal{U}$ -measurable and there exist constants $M_r > 1$, $M_l > 0$ such that, \mathbb{P} -a.s.,

$$0 \leq r_t(x_1, u) \leq M_r, \quad |l_t(x_1, u)| \leq M_l, \quad t \in [0, T], x_1 \in \mathbb{R}^d, u \in U.$$

$g : \Omega \times \mathbb{R}^d \rightarrow \mathbb{R}$ is $\mathcal{G}_{T \wedge \tau} \otimes \mathcal{B}(\mathbb{R}^d)$ -measurable, and there exists a constant β such that $\beta > \sup |r - 1|^2$, and

$$\mathbb{E}[e^{\beta C_T}] < +\infty, \quad \mathbb{E}[|g(X_{T \wedge \tau})|^2 e^{\beta C_T}] < +\infty, \quad (3)$$

with $C_t := C_t^{\mathbb{G}, X} + \Lambda_t^{\mathbb{G}}$.

The optimal control problem

For simplicity we consider only the case where Assumption 1 is satisfied, so that, $dC^{\mathbb{F},X} = dC^{\mathbb{G},X}$ and $\phi^{\mathbb{F},X}(dx_1) = \phi^{\mathbb{G},X}(dx_1)$.

Let $\mathcal{C} = \{U\text{-valued } \mathbb{G}\text{-predictable processes } u(\cdot)\}$ and set

$$\hat{\mathcal{C}} := \{u \in \mathcal{C} : 1_{[T \wedge \tau, T]} u = 0\} \subseteq \mathcal{C}.$$

Remark. \mathbb{F} - and \mathbb{G} -predictable processes coincide on $[0, \tau]$, so the set $\hat{\mathcal{C}}$ consists of strategies which are morally \mathbb{F} -predictable.

To every admissible control process $u \in \hat{\mathcal{C}}$ we will associate the cost functional

$$J(u) = \mathbb{E}_u \left[\int_0^{T \wedge \tau} l_t(X_t, u_t) dC_t^{\mathbb{F},X} + g(X_{T \wedge \tau}) \right],$$

where \mathbb{P}_u is a suitable probability measure, absolutely continuous with respect to \mathbb{P} . The control problem will be

$$\inf_{u \in \hat{\mathcal{C}}} J(u) \quad (P)$$

The probability measure \mathbb{P}_u

For any $u \in \mathcal{C}$ we introduce the Doléans-Dade exponential

$$L_t^u = \exp \left(\int_0^t \int_{\mathbb{R}^{d+1}} (1 - R_s(x_1, x_2, u_s)) \mathbf{v}^{\mathbb{G}, Z}(ds, dx_1, dx_2) \right) \prod_{n \geq 1: T_n \leq t} R_{T_n}(X_{T_n}, H_{T_n}, u_{T_n}),$$

with $R_t(x_1, x_2, u) := r_t(x_1, u) 1_{\{x_2=0\}} + 1_{\{x_2 \neq 0\}}$.

Lemma

Assume that

$$\mathbb{E}[e^{(3+M_r^4)C_T}] < +\infty. \quad (4)$$

Then, for every $u \in \mathcal{C}$, L^u is a square integrable \mathbb{G} -martingale.

We can then define $\mathbb{P}_u(d\omega) := L_T^u(\omega) \mathbb{P}(d\omega)$.

$$\mathbf{v}^{\mathbb{G}, Z, u}(\omega, dt, dx_1, dx_2) = R_t(\omega, X_t(\omega), H_t(\omega), u_t(\omega)) \mathbf{v}^{\mathbb{G}, Z}(\omega, dt, dx_1, dx_2)$$

is the \mathbb{G} -compensator of μ^Z under \mathbb{P}_u .

The associated BSDE

We consider the following BSDE: \mathbb{P} -a.s., for all $t \in [0, T]$,

$$\begin{aligned} Y_t + \int_t^T \int_{\mathbb{R}^{d+1}} \Theta_s(x_1, x_2) (\mu^Z - \nu^{\mathbb{G}, Z})(ds, dx_1, dx_2) \\ = g(X_{T \wedge \tau}) + \int_t^T f(s, X_s, \Theta_s(\cdot)) \mathbf{1}_{[0, T \wedge \tau]}(s) dC_s^{\mathbb{F}, X} \end{aligned} \quad (5)$$

with

$$\begin{aligned} f(\omega, t, y_1, \theta(\cdot)) \\ := \inf_{u \in U} \left\{ l_t(\omega, y_1, u) + \int_{\mathbb{R}^d} \theta(x_1, 0) (r_t(\omega, x_1, u) - 1) \phi_t^{\mathbb{F}, X}(\omega, dx_1) \right\}. \end{aligned}$$

Assumption 1

$\forall \Theta \in L^1(\mu^Z)$, $\exists \hat{\underline{u}}^\Theta \in \hat{\mathcal{C}}$ s.t., for a.a. (ω, t) wrt $dC_t^{\mathbb{F}, X}(\omega) \mathbb{P}(d\omega)$,

$$\begin{aligned} f(\omega, t, X_{t-}(\omega), \Theta_t(\omega, \cdot)) &= l_t(\omega, X_{t-}(\omega), \hat{\underline{u}}^\Theta(\omega, t)) \\ &+ \int_{\mathbb{R}^d} \Theta_t(\omega, x_1, 0) (r_t(\omega, x_1, \hat{\underline{u}}^\Theta(\omega, t)) - 1) \phi_t^{\mathbb{F}, X}(\omega, dx_1). \end{aligned}$$

$L_{\text{Prog}}^{2,\beta}(\Omega \times [0, T], \mathbb{G})$ is the space of real-valued \mathbb{G} -progr. meas. Y s.t.

$$\mathbb{E} \left[\int_0^T e^{\beta C_t} |Y_t|^2 dC_t \right] < \infty.$$

$L^{2,\beta}(\mu^Z, \mathbb{G})$ is the space of $\mathcal{P}(\mathbb{G}) \otimes \mathcal{B}(\mathbb{R}^{d+1})$ -meas. Θ s.t.

$$\mathbb{E} \left[\int_0^T \int_{\mathbb{R}^d} e^{\beta C_t} |\Theta_t(x_1, 0)|^2 \phi_t^{\mathbb{G}, X}(dx_1) dC_t^{\mathbb{G}, X} \right] + \mathbb{E} \left[\int_0^T e^{\beta C_t} |\Theta_t(0, 1)|^2 d\Lambda_t^{\mathbb{G}} \right] < \infty.$$

Theorem

Let Assumption 1 hold true. Set

$$L := \text{ess sup}_{\omega} \left(\sup \{ |r_t(x_1, u) - 1| : t \in [0, T], x_1 \in \mathbb{R}^d, u \in U \} \right),$$

and let (3) hold true with $\beta > L^2$.

Then BSDE (5) admits a unique solution

$$(Y, \Theta(\cdot)) \in L_{\text{Prog}}^{2,\beta}(\Omega \times [0, T], \mathbb{G}) \times L^{2,\beta}(\mu^Z, \mathbb{G}).$$

In particular, $Y = Y_{\cdot \wedge \tau}$, $\mathbb{P}(d\omega)$ -a.e., and $\Theta = \Theta 1_{[0, T \wedge \tau]}$

$\phi_t(\omega, dx_1, dx_2) dC_t(\omega) \mathbb{P}(d\omega)$ -a.e.

Solution to the optimal control problem (P)

Theorem

Let Assumption 1 hold true, condition (3) hold true with $\beta > L^2$, and condition (4) be satisfied. Let $(Y, \Theta(\cdot))$ be the unique solution to BSDE (5), with corresponding admissible control $\hat{u}^\Theta \in \hat{\mathcal{C}}$.

Then \hat{u}^Θ is optimal and Y_0 is the optimal cost, i.e.

$$Y_0 = J(\hat{u}^\Theta) = \inf_{u \in \mathcal{C}} J(u).$$

Proof. Let $u \in \hat{\mathcal{C}}$. Thank to the previous result, (adding and subtracting $\mathbb{E}_u \left[\int_0^{T \wedge \tau} l_s(X_s, \hat{u}) dC_s^{X, \mathbb{F}} \right]$) we obtain

$$\begin{aligned} Y_0 = J(u) + \mathbb{E}_u \left[\int_0^{T \wedge \tau} [f(s, X_s, \Theta_s(\cdot)) - l(s, X_s, \hat{u}) \right. \\ \left. - \int_{\mathbb{R}^d} \Theta_s(x_1, 0) (\bar{r}_s(x_1, u) - 1) \phi_s^{\mathbb{F}, X}(dx_1)] dC_s^{\mathbb{F}, X} \right]. \end{aligned}$$

The conclusion comes from the definition of f and Assumption 3. \square

An auxiliary control problem

The functional cost in the optimal control problem (P) can be equivalently rewritten as

$$J(u) = \mathbb{E}_u \left[\int_0^T l_t(X_t, u_t) 1_{[0, T \wedge \tau(\omega)]}(t) dC_t^{\mathbb{F}, X} + g(X_{T \wedge \tau}) \right], \quad u \in \hat{\mathcal{C}}.$$

Clearly $l_{[0, T \wedge \tau]}$ is $\mathcal{P}(\mathbb{G}) \otimes \mathcal{B}(\mathbb{R}^d) \otimes \mathcal{U}$ -measurable and $g(X_{T \wedge \tau})$ is \mathcal{G}_T -measurable.

Let us now consider the *enlarged* optimal control problem obtained by **taking the infimum over all the \mathbb{G} -predictable processes**:

$$\inf_{u \in \hat{\mathcal{C}}} J(u) = \mathbb{E}_u \left[\int_0^T l_t(X_t, u_t) 1_{[0, T \wedge \tau(\omega)]}(t) dC_t^{\mathbb{F}, X} + g(X_{T \wedge \tau}) \right]. \quad (P')$$

One can solve optimal control problem (P') by considering the following BSDE: \mathbb{P} -a.s., for all $t \in [0, T]$,

$$\begin{aligned} \tilde{Y}_t + \int_t^T \int_{\mathbb{R}^{d+1}} \tilde{\Theta}_s(x_1, x_2) (\mu^Z - \nu^{\mathbb{G}, Z})(ds, dx_1, dx_2) \\ = g(X_{T \wedge \tau}) + \int_t^T \tilde{f}(s, X_s, \tilde{\Theta}_s(\cdot)) dC_t^{\mathbb{F}, X}, \end{aligned} \quad (6)$$

where

$$\begin{aligned} \tilde{f}(\omega, t, y_1, \theta(\cdot)) := \inf_{u \in U} \left\{ l_t(\omega, y_1, u) 1_{[0, T \wedge \tau(\omega)]}(t) \right. \\ \left. + \int_{\mathbb{R}^d} \theta(x_1, 0) (r_t(\omega, x_1, u) - 1) \phi_t^{\mathbb{F}, X}(\omega, dx_1) \right\} \end{aligned}$$

for every $\omega \in \Omega$, $t \in [0, T]$, $y_1 \in \mathbb{R}^d$, and $\theta \in \mathcal{L}^1(\mathbb{R}^{d+1}, \mathcal{B}(\mathbb{R}^{d+1}), \phi_t(\omega, dx_1, dx_2))$.

Assumption 2

$\forall \Theta \in L^1(\mu^Z)$, \exists \mathbb{G} -predictable process $\underline{u}^\Theta : \Omega \times [0, T] \rightarrow U$, s.t., for a.a. (ω, t) wrt $dC_t^{\mathbb{F}, X}(\omega) \mathbb{P}(d\omega)$,

$$\begin{aligned} \tilde{f}(\omega, t, X_{t-}(\omega), \Theta_t(\omega, \cdot)) &= l_t(\omega, X_{t-}(\omega), \underline{u}^\Theta(\omega, t)) 1_{[0, T \wedge \tau(\omega)]}(t) \\ &+ \int_{\mathbb{R}^d} \Theta_t(\omega, x_1, 0) (r_t(\omega, x_1, \underline{u}^\Theta(\omega, t)) - 1) \phi_t^{\mathbb{F}, X}(\omega, dx_1). \end{aligned}$$

Theorem

Let Assumption 2 hold true. Let condition (3) hold true with $\beta > L^2$, and that condition (4) is satisfied. Let $(\tilde{Y}, \tilde{\Theta}(\cdot))$ denote the unique solution to BSDE (6), with corresponding admissible control $\underline{u}^{\tilde{\Theta}} \in \mathcal{C}$. Then

$$\tilde{Y}_0 = J(\underline{u}^{\tilde{\Theta}}) = \inf_{u \in \mathcal{C}} J(u).$$

Let us now go back to our original optimal control problem (P) : we aim at finding an admissible process $\hat{u} \in \mathcal{C}$ such that

$$J(\hat{u}) = \inf_{u \in \mathcal{C}} J(u).$$

Such an optimal control process exists and is provided by $\underline{u}^{\tilde{\Theta}} 1_{[0, T \wedge \tau]}$. Moreover, the value functions of (P) and (P') coincide:

Theorem

*Let Assumption 2 holds true. Assume that (3) holds true with $\beta > L^2$, and that condition (4) holds true. Let $(\tilde{Y}, \tilde{\Theta}(\cdot))$ denote the unique solution to BSDE (6), with corresponding admissible control $\underline{u}^{\tilde{\Theta}} \in \mathcal{C}$. Then $\hat{u} := \underline{u}^{\tilde{\Theta}} 1_{[0, T \wedge \tau]} \in \mathcal{C}$ is an **optimal control** process for (P) and*

$$\tilde{Y}_0 = J(\underline{u}^{\tilde{\Theta}}) = J(\hat{u}) = Y_0.$$

Thank you for your attention!