

# Backward SDEs with constrained jumps and Quasi-Variational Inequalities: applications to impulse controls in finance

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# Introduction

Consider the parabolic **Quasi-Variational Inequality** (QVI) :

$$\min [-\partial_t v - \mathcal{L}v - f, v - \mathcal{H}v] = 0, \quad v(T, \cdot) = g, \quad (1)$$

where  $\mathcal{L}$  is the **second order local operator**

$$\mathcal{L}v(t, x) = b(x) \cdot D_x v(t, x) + \frac{1}{2} \text{tr}(\sigma \sigma'(x) D_x^2 v(t, x))$$

and  $\mathcal{H}$  is the **nonlocal operator**

$$\mathcal{H}v(t, x) = \sup_{e \in E} \mathcal{H}^e v(t, x)$$

with

$$\mathcal{H}^e v(t, x) = v(t, x + \gamma(x, e)) + c(x, e).$$

## Introduction (II)

The QVI (1) is the **dynamic programming equation of the impulse control problem** (see Bensoussan-Lions 82 or Øksendal-Sulem 06) :

$$v(t, x) = \sup_{\alpha} \mathbf{E} \left[ g(X_T^{\alpha}) + \int_t^T f(X_s^{\alpha}) ds + \sum_{t < \tau_i \leq s} c(X_{\tau_i}^{\alpha}, \xi_i) \right]$$

with

- **controls** :  $\alpha = (\tau_i, \xi_i)_i$  where
  - $(\tau_i)_i$  **time decisions** : nondecreasing sequence of stopping times
  - $(\xi_i)_i$  **action decisions** : sequence of r.v. s.t.  $\xi_i \in \mathcal{F}_{\tau_i}$  valued in  $E$ ,
- **controlled process**  $X^{\alpha}$  defined by

$$X_s^{\alpha} = x + \int_t^s b(X_u^{\alpha}) du + \int_t^s \sigma(X_u^{\alpha}) dW_u + \sum_{t < \tau_i \leq s} \gamma(X_{\tau_i}^{\alpha}, \xi_i)$$

## Introduction (III)

### Various applications of impulse controls :

- Financial modelling with discrete transaction dates, due e.g. to fixed transaction costs or liquidity constraints
- Optimal multiple stopping : swing options
- Firm's investment and real options : management of power plants, valuation of gas storage, ...
- $\vdots$
- More generally to models with control policies that do not accumulate in time.

→ Many papers!

## Introduction (IV)

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  - The obstacle term contains the solution itself
  - It is nonlocal

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- ▶ Classical approach : **Decouple** the QVI (1) by defining by iteration the sequence of functions  $(v_n)_n$  :

$$\min [-\partial_t v_{n+1} - \mathcal{L}v_{n+1} - f, v_{n+1} - \mathcal{H}v_n] = 0, v_{n+1}(T, \cdot) = g \quad (2)$$

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→ associated to a **sequence of optimal stopping time problems** (reflected BSDEs)

→ Furthermore, to compute  $v_{n+1}$ , we need to know  $v_n$  on the whole domain → heavy computations : **numerically challenging!**

# Introduction (V)

- **Our basic motivation :**

- ▶ Find a probabilistic representation of QVI using BSDE, i.e. nonlinear Feynman-Kac formula
- ▶ We hope to use such a representation for deriving a direct numerical procedure for QVI



# Idea of the approach

- Instead of viewing the obstacle term as a reflection of  $v$  onto  $\mathcal{H}v$  (or  $v_{n+1}$  onto  $\mathcal{H}v_n$ )
- ▶ consider it as a constraint on the jumps of  $v(t, X_t)$  for some suitable forward jump process  $X$  :

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  - ▶ consider it as a constraint on the jumps of  $v(t, X_t)$  for some suitable forward jump process  $X$  :
- Let us introduce the **uncontrolled** jump diffusion  $X$  :

$$dX_t = b(X_t)dt + \sigma(X_t)dW_t + \int_E \gamma(X_{t-}, e)\mu(dt, de), \quad (3)$$

where  $\mu$  is a **Poisson random measure** whose intensity  $\lambda$  is **finite** and **supports the whole space  $E$** .

## Idea of the approach (II)

Take some smooth function  $v(t, x)$  and define :

$$Y_t := v(t, X_t), \quad Z_t := \sigma(X_{t-})' D_x v(t, X_{t-}),$$

$$\begin{aligned} U_t(e) &:= v(t, X_{t-} + \gamma(X_{t-}, e)) - v(t, X_{t-}) + c(X_{t-}, e) \\ &= (\mathcal{H}^e v - v)(t, X_{t-}) \end{aligned}$$

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► Apply Itô's formula :

$$\begin{aligned} Y_t &= Y_T + \int_t^T f(X_s) ds + K_T - K_t - \int_t^T Z_s \cdot dW_s \\ &\quad + \int_t^T \int_E [U_s(e) - c(X_{s-}, e)] \mu(ds, de), \end{aligned}$$

where

$$K_t := \int_0^t (-\partial_t v - \mathcal{L}v - f)(s, X_s) ds$$

## Idea of the approach (III)

- Now, suppose that  $\min[-\partial_t v - \mathcal{L}v - f, v - \mathcal{H}v] \geq 0$ , and  $v(T, \cdot) = g$  :
- ▶ Then  $(Y, Z, U, K)$  satisfies

$$\begin{aligned}
 Y_t = & g(X_T) + \int_t^T f(X_s) ds + K_T - K_t - \int_t^T Z_s \cdot dW_s \\
 & + \int_t^T \int_E [U_s(e) - c(X_{s-}, e)] \mu(ds, de), \quad (4)
 \end{aligned}$$

$K$  is a **nondecreasing process**, and  $U$  satisfies the nonpositivity constraint :

$$-U_t(e) \geq 0, \quad 0 \leq t \leq T, \quad e \in E. \quad (5)$$

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- ▶ View (4)-(5) as a **Backward Stochastic Equation (BSE) with jump constraints**
- ▶ We expect to retrieve the solution to the QVI (1) by solving the **minimal solution** to this constrained BSE.

## Remark : Another look at this BSE.

The nonnegativity jump-constraint :  $-U_t(e) \geq 0$  can be removed by defining another nondecreasing process :

$$\bar{K}_t := K_t - \int_0^t \int_E U_s(e) \mu(ds, de),$$

so that the BSE for  $Y$  becomes (for simplicity, take  $c = 0$ ) :

$$Y_t + \int_t^T Z_s \cdot dW_s = g(X_T) + \int_t^T f(X_s) ds + \bar{K}_T - \bar{K}_t$$

→ The minimal solution to this BSE corresponds to the **superreplication problem** of the payoff  $g(X_T) + \int_t^T f(X_s) ds$  by means of  $W$  in a jump-diffusion model. (Bouchard 06).

► Here, we shall keep explicitly the jump-constraint → more general jump-constraint on  $U$

► Moreover, by considering general dependence on  $f$ ,  $c$ , we introduce a class of **BSDE with constrained jumps**.

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  - Existence and approximation via penalization
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## Definition

**Minimal Solution** : find a solution

$(Y, Z, U, K) \in \mathcal{S}^2 \times \mathbf{L}^2(\mathbf{W}) \times \mathbf{L}^2(\tilde{\mu}) \times \mathbf{A}^2$  to

$$\begin{aligned}
 Y_t = & g(X_T) + \int_t^T f(X_s, Y_s, Z_s) ds + K_T - K_t - \int_t^T Z_s \cdot dW_s \\
 & - \int_t^T \int_E (U_s(e) - c(X_{s-}, Y_{s-}, Z_s, e)) \mu(ds, de)
 \end{aligned} \tag{6}$$

with

$$h(U_t(e), e) \geq 0, \quad d\mathbf{P} \otimes dt \otimes \lambda(de) \text{ a.e.} \tag{7}$$

such that for any other solution  $(\tilde{Y}, \tilde{Z}, \tilde{U}, \tilde{K})$  to (6)-(7) :

$$Y_t \leq \tilde{Y}_t, \quad 0 \leq t \leq T, \text{ a.s.}$$

# BSDEs with constrained jumps

## Related literature

- Constraints on  $Y \rightarrow$  reflected BSDE : El Karoui et al (97), Hamadène et al, etc ...
- Constraints on  $Z$  : Cvitanic et al (98), Hu and Buckdahn (98), Peng (99), Peng and Xu (07)

## Assumptions on coefficients

- **Forward SDE** :  $b$  and  $\sigma$  Lipschitz continuous,  $\gamma$  bounded and Lipschitz continuous w.r.t.  $x$  uniformly in  $e$  :

$$|\gamma(x, e) - \gamma(x', e)| \leq k|x - x'| \quad \forall e \in E$$

- **Backward SDE** :  $f$ ,  $g$  and  $c$  have linear growth,  $f$  and  $g$  Lipschitz continuous,  $c$  Lipschitz continuous w.r.t.  $y$  and  $z$  uniformly in  $x$  and  $e$

$$|c(x, y, z, e) - c(x, y', z', e)| \leq k_c(|y - y'| + |z - z'|)$$

- **Constraint** :  $h$  Lipschitz continuous w.r.t.  $u$  uniformly in  $e$  :

$$|h(u, e) - h(u', e)| \leq k_h|u - u'|$$

and

$$u \mapsto h(u, e) \text{ nonincreasing. (e.g. } h(u, e) = -u)$$

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# Penalized BSDEs

Consider for each  $n$  the BSDE with jumps :

$$Y_t^n = g(X_T) + \int_t^T f(X_s, Y_s^n, Z_s^n) ds + K_T^n - K_t^n - \int_t^T Z_s^n \cdot dW_s - \int_t^T \int_E [U_s^n(e) - c(X_{s-}, Y_{s-}^n, Z_{s-}^n, e)] \mu(ds, de) \quad (8)$$

with a penalization term

$$K_t^n = n \int_0^t \int_E h^-(U_s^n(e), e) \lambda(de) ds$$

where  $h^- = \max(-h, 0)$ .

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where  $h^- = \max(-h, 0)$ .

→ For each  $n$ , existence and uniqueness of  $(Y^n, Z^n, U^n)$  solution to (8) from Tang and Li (94), and Barles et al. (97).

## Convergence of the penalized BSDEs

- Convergence of  $(Y^n)$  : usually by comparison results
  - Convergence of  $(Z^n, U^n, K^n)$  : more difficult !
- Moreover, in general, we need some strong convergence to pass to the limit in the nonlinear terms  $f(x, y, z)$ ,  $c(x, y, z)$  and  $h(u, e)$ .
- Uniform boundedness
  - Weak convergence method (Peng)



## Comparison results

### Lemma

*The sequence  $(Y^n)_n$  is nondecreasing, i.e.  $\forall n \in \mathbf{N}$ ,  $Y_t^n \leq Y_t^{n+1}$ ,  $0 \leq t \leq T$ , a.s.*

**Proof.** Based on comparison theorem for BSDEs with jumps in Royer (04). We used the nonincreasing property of  $h$ .

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**Proof.** Based on comparison theorem for BSDEs with jumps in Royer (04). We used the nonincreasing property of  $h$ .

### Lemma

For any quadruple  $(\tilde{Y}, \tilde{Z}, \tilde{U}, \tilde{K}) \in \mathcal{S}^2 \times \mathbf{L}^2(\mathbf{W}) \times \mathbf{L}^2(\tilde{\mu}) \times \mathbf{A}^2$  satisfying (6)-(7), and for all  $n \in \mathbf{N}$ , we have

$$Y_t^n \leq \tilde{Y}_t, \quad 0 \leq t \leq T, \text{ a.s.}$$

**Proof.** Suitable change of probability measures.

# Uniform boundedness of the penalized BSDEs

## Asumption (H1)

There exists a triple  $(\tilde{Y}, \tilde{Z}, \tilde{U}, \tilde{K}) \in \mathcal{S}^2 \times \mathbf{L}^2(\mathbf{W}) \times \mathbf{L}^2(\mu) \times \mathbf{A}^2$  satisfying (6)-(7)

## Lemma

Under **(H1)**, there exists some constant  $C$  such that

$$\|Y^n\|_{\mathcal{S}^2} + \|Z^n\|_{\mathbf{H}^2} + \|U^n\|_{\mathbf{L}^2(\tilde{\mu})} + \|K^n\|_{\mathcal{S}^2} \leq C \quad (9)$$

for all  $n \in \mathbf{N}$ .

**Proof.** Classical arguments based on elementary inequality  $2ab \leq \frac{a^2}{\eta} + \eta b^2$ , Gronwall's lemma and Burkholder-Davis-Gundy's inequality + comparison result of previous lemma.

# Convergence of the penalized solutions

## Theorem

Under **(H1)**, there exists a unique minimal solution

$$(Y, Z, U, K) \in \mathcal{S}^2 \times \mathbf{L}^2(\mathbf{W}) \times \mathbf{L}^2(\tilde{\mu}) \times \mathbf{A}^2$$

with  $K$  predictable, to (6)-(7).  $Y$  is the increasing limit of  $(Y^n)$  and also in  $\mathbf{L}^2_{\mathbb{F}}(\mathbf{0}, \mathbf{T})$ ,  $K$  is the weak limit of  $(K^n)$  in  $\mathbf{L}^2_{\mathbb{F}}(\mathbf{0}, \mathbf{T})$ , and for any  $p \in [1, 2)$ ,

$$\|Z^n - Z\|_{\mathbf{L}^p(\mathbf{W})} + \|U^n - U\|_{\mathbf{L}^p(\tilde{\mu})} \longrightarrow 0,$$

as  $n$  goes to infinity.

**Proof.** Use the weak compactness of  $(Z^n)$ ,  $(U^n)$  and  $(f(X^n, Y^n, Z^n))$  and  $(K^n)$  to get limits  $Z$ ,  $U$ ,  $\phi$  and  $K$ . Then control jumps of the predictable process  $K$  via a random partition of the interval  $(0, T)$  and obtain a convergence in measure.

## Nonmarkovian case

### Remark

Existence and uniqueness results for the minimal solution hold true in a nonmarkovian framework :

$\mathbb{F}$  = filtration generated by  $W$  and  $\mu$

$$g(X_T) = \zeta$$

$$f(x, y, z) = f(\omega, y, z)$$

$$c(x, y, z) = c(\omega, y, z)$$

## Related semilinear QVIs and viscosity property

- Markov property of  $X \rightarrow Y_t = v(t, X_t)$  for some deterministic function  $v$
- Consider the **semilinear QVI** :

$$\min \left[ -\partial_t w - \mathcal{L}w - f(\cdot, w, \sigma' D_x w), \inf_{e \in E} h(\mathcal{H}^e w - w, e) \right] = 0 \quad (10)$$

where  $\mathcal{L}$  is the second order local operator as before, and  $\mathcal{H}^e$ ,  $e \in E$ , are the **nonlocal operators**

$$\mathcal{H}^e w(t, x) = w(t, x + \gamma(x, e)) + c(x, w(t, x), \sigma'(x) D_x w(t, x), e).$$

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$$\mathcal{H}^e w(t, x) = w(t, x + \gamma(x, e)) + c(x, w(t, x), \sigma'(x) D_x w(t, x), e).$$

### Assumption (H2)

The function  $v$  has linear growth :  $\sup_{[0, T] \times \mathbb{R}^d} \frac{v(t, x)}{1 + |x|} < \infty$ .

### Proposition

Under (H2), the function  $v$  is a viscosity solution to (10).

## Elements of proof : arguments by penalized IPDE

- Markov property of  $X \rightarrow Y_t^n = v_n(t, X_t)$  for some deterministic function  $v_n$ .
- From Barles et al (98), we know that  $v_n$  is a viscosity solution to the Integral PDE :

$$\begin{aligned}
 & -\partial_t w - \mathcal{L}w - f(\cdot, w, \sigma' D_x w) \\
 & -n \int_E h^-(\mathcal{H}^e w(t, x) - w(t, x), e) \lambda(de) = 0
 \end{aligned} \tag{11}$$

- We then pass to the limit by adapting [stability arguments](#) for viscosity solutions.



## Terminal condition for $v$

- Need a **terminal condition** to complete the PDE characterization of the function  $v$ .
- Condition  $v(T, \cdot) = g$  is irrelevant : discontinuity in  $T^-$  due to constraints

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- Condition  $v(T, \cdot) = g$  is irrelevant : discontinuity in  $T^-$  due to constraints
- ▶ **Face-lifting terminal data** :

$$\min \left[ v(T^-, \cdot) - g, \inf_{e \in E} h(\mathcal{H}^e v(T^-, \cdot) - v(T^-, \cdot), e) \right] = 0 \quad (12)$$

### Proposition

Under **(H2)**, the function  $v$  is a viscosity solution to (12)

# Comparison results for semilinear QVIs

## Asumption (H3)

There exists a nonnegative function  $\Lambda \in C^2(\mathbb{R}^d)$  satisfying

- (i)  $b \cdot D\Lambda + \frac{1}{2} \text{tr}(\sigma \sigma' D^2 \Lambda) + f(\cdot, \Lambda, \sigma' D\Lambda) \leq \rho \Lambda$  for some  $\rho > 0$
- (ii)  $\inf_{e \in E} h(\Lambda(x + \gamma(x, e) + c(x, \Lambda(x), \sigma(x)') D\Lambda(x) - \Lambda(x), e) \geq q(x)$   
for all  $x \in \mathbb{R}^d$  for some continuous function  $q > 0$  on  $\mathbb{R}^d$ .
- (iii)  $\Lambda \geq g$  on  $\mathbf{R}^d$
- (iv)  $\lim_{|x| \rightarrow \infty} \frac{\Lambda(x)}{1+|x|} = +\infty$

Assumption **(H3)** essentially ensures the existence of strict supersolution which allows to control the nonlocal term in QVI (10)-(12) via some convex small perturbation.  $\Rightarrow$  require some convexity conditions to deal with the dependence of  $f$  and  $c$  on  $y, z$ .

## Comparison results for semilinear QVIs (II)

### Asumption (H4)

- (i) *The function  $f(x, \cdot, \cdot)$  is convex in  $(y, z) \in \mathbf{R} \times \mathbf{R}^d \forall x \in \mathbf{R}^d$ .*
- (ii) *The function  $h(\cdot, e)$  is concave in  $u \in \mathbf{R} \forall e \in E$ .*
- (iii) *The function  $c(x, \cdot, \cdot, e)$  is convex in  $(y, z) \in \mathbf{R} \times \mathbf{R}^d$ ,  $\forall (x, e) \in \mathbf{R}^d \times E$ .*
- (iv) *The function  $c(x, \cdot, z, e)$  is decreasing in  $y \in \mathbf{R}$ ,  $\forall (x, z, e) \in \mathbf{R}^d \times \mathbf{R}^d \times E$ .*

### Proposition

Assume that **(H3)** and **(H4)** hold. Let  $U$  (resp.  $V$ ) be LSC (resp. USC) viscosity supersolution (resp. subsolution) of (10)-(12) satisfying the linear growth condition

$$\sup_{[0, T] \times \mathbf{R}^d} \frac{|U(t, x)| + |V(t, x)|}{1 + |x|} < \infty$$

Then,  $U \geq V$  on  $[0, T] \times \mathbf{R}^d$ .

## PDE characterization of the function $v$

### Theorem

Under **(H2)**, **(H3)** and **(H4)**, the function  $v$  is the unique viscosity solution to (10)-(12) satisfying the linear growth condition.

$$\sup_{(t,x) \in [0, T] \times \mathbb{R}^d} \frac{|v(t, x)|}{1 + |x|} < \infty.$$

Moreover  $v$  is continuous on  $[0, T) \times \mathbb{R}^d$ .

→ Probabilistic representation of semilinear QVIs, and in particular of impulse control problems by means of BSDEs with constrained jumps.

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## Approximation by the penalized BSDE

- We set  $V_t^n(e) = U_t^n(e) - c(X_t, Y_{t-}^n, Z_s^n, e)$ , and we rewrite the penalized BSDE for  $(Y^n, Z^n, V^n)$  as :

$$\begin{aligned}
 Y_t^n &= g(X_T) + \int_t^T \int_E f_n(X_s, Y_s^n, Z_s^n, V_s^n(e), e) \lambda(de) ds \\
 &\quad - \int_t^T Z_s^n dW_s - \int_t^T \int_E V_s^n(e) \tilde{\mu}(de, ds)
 \end{aligned}$$

where  $\tilde{\mu}(dt, de) = \mu(dt, de) - \lambda(de)dt$ , and

$$f_n(x, y, z, v, e) := \frac{1}{\lambda(E)} f(x, y, z) - v + nh^-(v + c(x, y, z, e), e).$$

- We assume for simplicity that the state space of jump size  $E$  is finite :  $E = \{1, \dots, m\}$  (otherwise discretize  $E$ ).

## Time discretization of the penalized BSDE

- Time grid  $\pi = (t_i)$  on  $[0, T]$  :  $t_i = i\Delta t$ ,  $i = 0, \dots, N$ ,  $\Delta t = T/N$
- Forward Euler scheme  $X^\pi$  for  $X$

$$X_{t_0}^\pi = x$$

$$X_{t_{i+1}}^\pi := X_{t_i}^\pi + b(X_{t_i}^\pi)\Delta t + \sigma(X_{t_i}^\pi)(W_{t_{i+1}} - W_{t_i}) + \sum_{e=1}^m \gamma(X_{t_i}^\pi, e)\mu((t_i, t_{i+1}] \times \{e\}).$$



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- Forward Euler scheme  $X^\pi$  for  $X$

$$X_{t_0}^\pi = x$$

$$X_{t_{i+1}}^\pi := X_{t_i}^\pi + b(X_{t_i}^\pi)\Delta t + \sigma(X_{t_i}^\pi)(W_{t_{i+1}} - W_{t_i}) + \sum_{e=1}^m \gamma(X_{t_i}^\pi, e)\mu((t_i, t_{i+1}] \times \{e\}).$$

- Backward Euler scheme  $(Y^{n,\pi}, Z^{n,\pi}, V^{n,\pi})$  for  $(Y^n, Z^n, V^n)$

$$Y_{t_N}^{n,\pi} = g(X_{t_N}^\pi)$$

$$Y_{t_i}^{n,\pi} = Y_{t_{i+1}}^{n,\pi} + \Delta t \sum_{e=1}^m \lambda(e) f_n(X_{t_i}^\pi, Y_{t_i}^{n,\pi}, Z_{t_i}^{n,\pi}, V_{t_i}^{n,\pi}, e) \\ - Z_{t_i}^{n,\pi} \cdot (W_{t_{i+1}} - W_{t_i}) - \sum_{e=1}^m V_{t_i}^{n,\pi}(e) \tilde{\mu}((t_i, t_{i+1}] \times \{e\})$$

## Time discretization of the penalized BSDE (II)

- Backward Euler scheme  $(Y^{n,\pi}, Z^{n,\pi}, V^{n,\pi})$  for  $(Y^n, Z^n, V^n)$

$$Y_{t_i}^{n,\pi} = Y_{t_{i+1}}^{n,\pi} + \Delta t \sum_{e=1}^m \lambda(e) f_n(X_{t_i}^\pi, Y_{t_i}^{n,\pi}, Z_{t_i}^{n,\pi}, V_{t_i}^{n,\pi}, e) - Z_{t_i}^{n,\pi} \cdot [W_{t_{i+1}} - W_{t_i}] - \sum_{e=1}^m V_{t_i}^{n,\pi}(e) \tilde{\mu}((t_i, t_{i+1}] \times \{e\}) \quad (13)$$

- By taking conditional expectation in (13) :

$$Y_{t_i}^{n,\pi} = \mathbf{E} \left[ Y_{t_{i+1}}^{n,\pi} \middle| \mathcal{F}_{t_i} \right] + \Delta t \sum_{e=1}^m \lambda(e) f_n(X_{t_i}^\pi, Y_{t_i}^{n,\pi}, Z_{t_i}^{n,\pi}, V_{t_i}^{n,\pi}, e)$$

- By multiplying by  $W_{t_{i+1}} - W_{t_i}$  and taking expectation :

$$Z_{t_i}^{n,\pi} = \frac{1}{\Delta t} \mathbf{E} \left[ Y_{t_{i+1}}^{n,\pi} (W_{t_{i+1}} - W_{t_i}) \middle| \mathcal{F}_{t_i} \right]$$

- By multiplying by  $\tilde{\mu}((t_i, t_{i+1}] \times \{e\})$  and taking expectation :

$$V_{t_i}^{n,\pi}(e) = \frac{1}{\lambda(e)\Delta t} \mathbf{E} \left[ Y_{t_{i+1}}^{n,\pi} \tilde{\mu}((t_i, t_{i+1}] \times \{e\}) \middle| \mathcal{F}_{t_i} \right], \quad e = 1, \dots, m.$$

# Simulation of the penalized BSDE

- For fixed penalization coefficient  $n$ , the rate of convergence of the time discretization procedure was analyzed in Bouchard and Elie (06).
  - ▶ Here, we need to refine the estimation for  $n$  large
- Computation of the conditional expectations : Longstaff-Schwarz algorithm, Monte-Carlo method, quantization method, random walk method ...

# Conclusion

- New insight into impulse control problems, and more generally into semilinear QVIs by means of BSDEs with constrained jumps
  - This provides direct (without iteration) probabilistic numerical procedure
- Current investigation and further questions
  - Analysis of the convergence of these approximation schemes
  - Numerical implementation.