

Rate of convergence for the discrete-time approximation of reflected BSDEs arising in switching problems

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Introduction

Switching problem: An example.
obliquely reflected BSDEs

Discrete-time approximation of reflected BSDEs

discretely obliquely reflected BSDEs
Discrete-time approximation scheme

Outline

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Starting and Stopping problem (1)

Hamadène and Jeanblanc ('01):

- ▶ Consider e.g. a power station producing electricity whose price is given by a diffusion process X : $dX_t = b(X_t)dt + \sigma(X_t)dW_t$
- ▶ Two modes for the power station:
 mode **1**: operating, profit is then $f^1(X_t)dt$
 mode **2**: closed, profit is then $f^2(X_t)dt$
 \hookrightarrow switching from one mode to another has a cost: $c > 0$
- ▶ Management decide to produce electricity only when it is profitable enough.
- ▶ The management strategy is (θ_j, α_j) : θ_j is a sequence of stopping times representing switching times from mode α_{j-1} to α_j .
 $(a_t)_{0 \leq t \leq T}$ is the state process (the management strategy).

Starting and Stopping problem (2)

- Following a strategy a from t up to T , gives

$$J(a, t) = \int_t^T f^{a_s}(X_s) ds - \sum_{j \geq 0} c \mathbf{1}_{\{t \leq \theta_j \leq T\}}$$

- The optimization problem is then

$$Y_0^1 := \sup_{a \text{ such that } \alpha_0=1} \mathbb{E}[J(a, 0)]$$

$$Y_0^2 := \sup_{a \text{ such that } \alpha_0=2} \mathbb{E}[J(a, 0)]$$

At any date $t \in [0, T]$ in state $i \in \{1, 2\}$, the value function is Y_t^i .

Solution

- ▶ Y is solution of a coupled optimal stopping problem

$$Y_t^1 = \operatorname{ess\,sup}_{t \leq \tau \leq T} \mathbb{E} \left[\int_t^\tau f^1(X_s) ds + (Y_\tau^2 - c) \mathbf{1}_{\{\tau < T\}} \mid \mathcal{F}_t \right]$$

$$Y_t^2 = \operatorname{ess\,sup}_{t \leq \tau \leq T} \mathbb{E} \left[\int_t^\tau f^2(X_s) ds + (Y_\tau^1 - c) \mathbf{1}_{\{\tau < T\}} \mid \mathcal{F}_t \right]$$

with terminal values $Y_T^1 = g^1(X_T)$ and $Y_T^2 = g^2(X_T)$.

- ▶ The optimal strategy (θ_j^*, α_j^*) is given by

$$\theta_{j+1}^* := \inf \{ s \geq \theta_j^* \mid Y_s^{\alpha_j^*} = \max_{i \in \{1,2\}} Y_s^i - c \}$$

$$\alpha_{j+1}^* := \mathbf{1} \text{ if } \alpha_j^* = 2, \text{ or } \mathbf{2} \text{ if } \alpha_j^* = 1.$$

System of reflected BSDEs

Y is the solution of the following system of reflected BSDEs:

$$Y_t^i = \int_t^T f^i(X_s) ds - \int_t^T (Z_s^i) dW_s + \int_t^T dK_s^i, \quad i \in \{1, 2\},$$

with (the coupling...)

$$Y_t^1 \geq Y_t^2 - c \text{ and } Y_t^2 \geq Y_t^1 - c, \quad \forall t \in [0, T]$$

and ('optimality' of K)

$$\int_0^T \left(Y_s^1 - (Y_s^2 - c) \right) dK_s^1 = 0 \text{ and } \int_0^T \left(Y_s^2 - (Y_s^1 - c) \right) dK_s^2 = 0$$

System of Markovian obliquely reflected BSDEs

$$\begin{cases} Y_t = g(X_T) + \int_t^T f(X_s, Y_s, Z_s) ds - \int_t^T Z_s dW_s + K_T - K_t, \\ Y_t^\ell \geq \max_{j \in \mathcal{I}} \{Y_t^j - c^{\ell j}(X_t)\}, \\ \int_0^T \left[Y_t^\ell - \max_{j \in \mathcal{I} \setminus \{\ell\}} \{Y_t^j - c^{\ell j}(X_t)\} \right] dK_t^\ell = 0, \end{cases} \quad \begin{array}{l} \ell \in \mathcal{I}, \\ \ell \in \mathcal{I}, \end{array}$$

where $\mathcal{I} := \{1, \dots, d\}$, f , g and $(c^{ij})_{i,j \in \mathcal{I}}$ are **Lipschitz functions** and X is solution to the following forward stochastic differential equation (SDE) with **Lipschitz coefficients**

$$X_t = x + \int_0^t b(X_s) ds + \int_0^t \sigma(X_s) dW_s. \quad (1)$$

Existence Uniqueness

Theorem ((Hu-Tang '00, Hamadène-Zhang '00,
Chassagneux-Élie-Kharroubi '11))

We assume:

- ▶ *Lipschitz assumptions on f, g, σ, b*
- ▶ *Natural structure condition on costs*
- ▶ *$f^j(x, y, z^j)$*

Then we have existence and uniqueness of a solution in an appropriate space.

To simplify this presentation, we assume (sometimes) that the generator does not depend on y .

Switching strategy

- Switching strategy a : nondecreasing sequence of stopping times $(\theta_j)_{j \in \mathbb{N}}$ and a sequence of random variables $(\alpha_j)_{j \in \mathbb{N}}$ valued in \mathcal{I} , such that α_j \mathcal{F}_{θ_j} -measurable.
- Current state process $(a_t)_{t \in [0, T]}$

$$a_t := \alpha_0 \mathbf{1}_{\{0 \leq t < \theta_0\}} + \sum_{j=1}^{\mathcal{N}^a} \alpha_{j-1} \mathbf{1}_{\{\theta_{j-1} \leq t < \theta_j\}}$$

- cumulative cost process $(A_t^a)_{t \in [0, T]}$

$$A_t^a := \sum_{j=1}^{\mathcal{N}^a} C_{\theta_j}^{\alpha_{j-1} \alpha_j} \mathbf{1}_{\{\theta_j \leq t \leq T\}}$$

- The set of admissible strategies starting from state i at time t :

$$\mathcal{A}_{t,i} = \{a = (\theta_j, \alpha_j)_j \mid \theta_0 = t, \alpha_0 = i, \mathbb{E}[|A_T^a|^2] < \infty\},$$

Optimal switching representation

For a strategy $a \in \mathcal{A}_{t,\ell}$, we introduce the one-dimensional *switched BSDE* whose solution (U^a, V^a) satisfies

$$U_t^a = g^{aT}(X_T) + \int_t^T f^{a_s}(X_s, V_s^a) ds - \int_t^T V_s^a dW_s - A_T^a + A_t^a \quad (2)$$

Proposition

$$(Y_t)^\ell = \operatorname{ess\,sup}_{a \in \mathcal{A}_{t,\ell}} U_t^a$$

and

$$(Y_t)^\ell = U_t^{\bar{a}}$$

with an explicit optimal strategy \bar{a} .

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Discretisation

We will adopt the same strategy than [Chassagneux-Élie-Kharroubi '12] (or [Bouchard-Chassagneux '12] for normal reflections): We introduce two grids $\pi = \{t_0 = 0, \dots, t_n\}$ and $\mathfrak{R} = \{r_0 = 0, \dots, r_\kappa\}$ with $\mathfrak{R} \subset \pi$,

- ▶ Continuous-time BSDE with discrete-time oblique reflection
 $\hookrightarrow (Y^{\mathfrak{R}}, Z^{\mathfrak{R}})$
- ▶ discrete-time BSDE with discrete-time oblique reflection
 $\hookrightarrow (Y^{\mathfrak{R}, \pi}, Z^{\mathfrak{R}, \pi})$

We want to study

$$\begin{aligned} & \text{Error}((Y, Z), (Y^{\mathfrak{R}, \pi}, Z^{\mathfrak{R}, \pi})) \\ & \leq \text{Error}((Y, Z), (Y^{\mathfrak{R}}, Z^{\mathfrak{R}})) + \text{Error}((Y^{\mathfrak{R}}, Z^{\mathfrak{R}}), (Y^{\mathfrak{R}, \pi}, Z^{\mathfrak{R}, \pi})) \end{aligned}$$

and improve [Chassagneux-Élie-Kharroubi '12] (rate and assumptions).

Discretely obliquely reflected BSDEs

$(Y^{\mathfrak{R}}, Z^{\mathfrak{R}})$ solution to the following discretely obliquely reflected BSDE:
 $Y_T^{\mathfrak{R}} = \tilde{Y}_T^{\mathfrak{R}} := g(X_T)$, and, for $j \leq \kappa - 1$ and $t \in [r_j, r_{j+1})$,

$$\begin{cases} \tilde{Y}_t^{\mathfrak{R}} = Y_{r_{j+1}}^{\mathfrak{R}} + \int_t^{r_{j+1}} f(X_u, \tilde{Y}_u^{\mathfrak{R}}, Z_u^{\mathfrak{R}}) du - \int_t^{r_{j+1}} Z_u^{\mathfrak{R}} dW_u, \\ Y_t^{\mathfrak{R}} = \tilde{Y}_t^{\mathfrak{R}} \mathbf{1}_{\{t \notin \mathfrak{R}\}} + \mathcal{P}(X_t, \tilde{Y}_t^{\mathfrak{R}}) \mathbf{1}_{\{t \in \mathfrak{R}\}}, \end{cases} \quad (3)$$

where $\mathcal{P}(x, \cdot)$ is the oblique projection operator (on a closed convex domain) defined by

$$\mathcal{P} : (x, y) \in \mathbb{R}^d \times \mathbb{R}^d \mapsto \left(\max_{j \in \mathcal{I}} \{y^j - c^{ij}(x)\} \right)_{1 \leq i \leq d}.$$

Error between discretely and continuously BSDEs

$|\mathfrak{R}|$: Mesh of the grid \mathfrak{R}

Proposition

Under same assumptions than for the existence and uniqueness theorem, we have

$$\mathbb{E} \left[\sup_{r \in \mathfrak{R}} |Y_r - Y_r^{\mathfrak{R}}|^2 + \sup_{t \in [0, T]} |Y_t - \tilde{Y}_t^{\mathfrak{R}}|^2 \right] \leq C |\mathfrak{R}| \log(2T/|\mathfrak{R}|),$$

$$\mathbb{E} \left[\int_0^T |Z_s - Z_s^{\mathfrak{R}}|^2 ds \right] \leq C \sqrt{|\mathfrak{R}| \log(2T/|\mathfrak{R}|)}.$$

Comparison with [Chassagneux-Élie-Kharroubi '12]

Almost the same speed of convergence than in [Chassagneux-Élie-Kharroubi '12] but without restrictive assumptions:

- ▶ Structural assumption $f^i(x, y, z) = f^i(x, y, z^{i\cdot})$ instead of $f^i(x, y, z) = f^i(x, y^i, z^{i\cdot})$
- ▶ No extra-regularity assumptions on costs c
- ▶ No assumption $|f(x, y, z)| \leq C(1 + |x| + |y|)$.

Idea of the proof

Key tool: optimal switching representation for the discretely reflected BSDE.



$$(Y_t^{\mathfrak{R}})^{\ell} = \operatorname{ess\,sup}_{a \in \mathcal{A}_{t,\ell}^{\mathfrak{R}}} U_t^a = U_t^{\bar{a}^{\mathfrak{R}}}$$

where $\mathcal{A}_{t,\ell}^{\mathfrak{R}}$ are admissible strategies with switching times living in \mathfrak{R} .



$$0 \leq (Y_t)^{\ell} - (Y_t^{\mathfrak{R}})^{\ell} \leq U_t^{\bar{a}} - U_t^{\tilde{a}}$$

with \tilde{a} a “projection” on \mathfrak{R} of the continuous-time strategy \bar{a} .

Thanks to dimension 1, we can use comparison results.



An other key-estimate:

$$|Z_t^{\mathfrak{R}}| \leq L(1 + |X_t|)$$

with L that does not depend on \mathfrak{R} .

discrete-time approximation

We want to discretize the following BSDE:

$$\begin{cases} \tilde{Y}_t^{\mathfrak{R}} = Y_{r_{j+1}}^{\mathfrak{R}} + \int_t^{r_{j+1}} f(X_u, \tilde{Y}_u^{\mathfrak{R}}, Z_u^{\mathfrak{R}}) du - \int_t^{r_{j+1}} Z_u^{\mathfrak{R}} dW_u, \\ Y_t^{\mathfrak{R}} = \tilde{Y}_t^{\mathfrak{R}} \mathbf{1}_{\{t \notin \mathfrak{R}\}} + \mathcal{P}(X_t, \tilde{Y}_t^{\mathfrak{R}}) \mathbf{1}_{\{t \in \mathfrak{R}\}}, \end{cases} \quad (4)$$

We just have to use classical backward time discretized schemes for classical BSDEs

$$\tilde{Y}_t^{\mathfrak{R}} = Y_{r_{j+1}}^{\mathfrak{R}} + \int_t^{r_{j+1}} f(X_u, \tilde{Y}_u^{\mathfrak{R}}, Z_u^{\mathfrak{R}}) du - \int_t^{r_{j+1}} Z_u^{\mathfrak{R}} dW_u$$

Deriving the scheme

We are given our discretization grid

$\pi = \{0 = t_0 < \dots < t_i < \dots < t_n = T\}$, define $h_i = t_{i+1} - t_i$. We do not consider projections for the moment.

▶ Start with:
$$\tilde{Y}_{t_i}^{\mathfrak{R}} + \int_{t_i}^{t_{i+1}} Z_s^{\mathfrak{R}} dW_s = \tilde{Y}_{t_{i+1}}^{\mathfrak{R}} + \int_{t_i}^{t_{i+1}} f(X_s, \tilde{Y}_s^{\mathfrak{R}}, Z_s^{\mathfrak{R}}) ds \quad (1)$$

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- ▶ Start with: $\tilde{Y}_{t_i}^{\mathfrak{R}} + \int_{t_i}^{t_{i+1}} Z_s^{\mathfrak{R}} dW_s \simeq \tilde{Y}_{t_{i+1}}^{\mathfrak{R}} + h_i f(X_{t_i}^{\pi}, \tilde{Y}_{t_i}^{\mathfrak{R}}, Z_{t_i}^{\mathfrak{R}})$ (1)

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- ▶ For the $\tilde{Y}^{\mathfrak{R}}$ -part:

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Take conditional expectation, $\tilde{Y}_{t_i}^{\mathfrak{R}} \simeq \mathbb{E}_{t_i} \left[\tilde{Y}_{t_{i+1}}^{\mathfrak{R}} + h_i f(X_{t_i}^{\pi}, \tilde{Y}_{t_i}^{\mathfrak{R}}, Z_{t_i}^{\mathfrak{R}}) \right]$

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$$\Leftrightarrow \tilde{Y}_i^{\mathcal{R}, \pi} := \mathbb{E}_{t_i} \left[\tilde{Y}_{i+1}^{\mathcal{R}, \pi} \right] + h_i f(X_{t_i}^{\pi}, \tilde{Y}_i^{\mathcal{R}, \pi}, Z_i^{\mathcal{R}, \pi})$$

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$$\Leftrightarrow \tilde{Y}_i^{\mathcal{R}, \pi} := \mathbb{E}_{t_i} \left[\tilde{Y}_{i+1}^{\mathcal{R}, \pi} \right] + h_i f(X_{t_i}^{\pi}, \tilde{Y}_i^{\mathcal{R}, \pi}, Z_i^{\mathcal{R}, \pi})$$

- ▶ For the $Z^{\mathcal{R}}$ -part:

Deriving the scheme

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- ▶ For the $\tilde{Y}^{\mathcal{R}}$ -part:

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- ▶ For the $Z^{\mathcal{R}}$ -part:

Multiply (1) by $\Delta W_i := W_{t_{i+1}} - W_{t_i}$, take conditional expectation:

$$\mathbb{E}_{t_i} \left[\int_{t_i}^{t_{i+1}} Z_s^{\mathcal{R}} ds \right] \simeq \mathbb{E}_{t_i} \left[\tilde{Y}_{t_{i+1}}^{\mathcal{R}} \Delta W_i \right]$$

Deriving the scheme

We are given our discretization grid

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- ▶ For the $Z^{\mathfrak{R}}$ -part:

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$$\mathbb{E}_{t_i} \left[\int_{t_i}^{t_{i+1}} Z_s^{\mathfrak{R}} ds \right] \simeq \mathbb{E}_{t_i} \left[\tilde{Y}_{t_{i+1}}^{\mathfrak{R}} \Delta W_i \right]$$

$$\text{Say } \mathbb{E}_{t_i} \left[\int_{t_i}^{t_{i+1}} Z_s^{\mathfrak{R}} ds \right] \simeq h_i Z_{t_i}^{\mathfrak{R}}, \implies h_i Z_{t_i}^{\mathfrak{R}} \simeq \mathbb{E}_{t_i} \left[\tilde{Y}_{t_{i+1}}^{\mathfrak{R}} \Delta W_i \right]$$

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$\pi = \{0 = t_0 < \dots < t_i < \dots < t_n = T\}$, define $h_i = t_{i+1} - t_i$. We do not consider projections for the moment.

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- ▶ For the $Z^{\mathcal{R}}$ -part:

Multiply (1) by $\Delta W_i := W_{t_{i+1}} - W_{t_i}$, take conditional expectation:

$$\mathbb{E}_{t_i} \left[\int_{t_i}^{t_{i+1}} Z_s^{\mathcal{R}} ds \right] \simeq \mathbb{E}_{t_i} \left[\tilde{Y}_{t_{i+1}}^{\mathcal{R}} \Delta W_i \right]$$

$$\text{Say } \mathbb{E}_{t_i} \left[\int_{t_i}^{t_{i+1}} Z_s^{\mathcal{R}} ds \right] \simeq h_i Z_{t_i}^{\mathcal{R}}, \implies h_i Z_{t_i}^{\mathcal{R}} \simeq \mathbb{E}_{t_i} \left[\tilde{Y}_{t_{i+1}}^{\mathcal{R}} \Delta W_i \right]$$

$$\hookrightarrow Z_i^{\mathcal{R}, \pi} := \mathbb{E}_{t_i} \left[\tilde{Y}_{i+1}^{\mathcal{R}, \pi} H_i \right] \quad \text{with } \langle H_i := h_i^{-1} \Delta W_i \rangle$$

discrete-time approximation scheme

$$\begin{cases} \tilde{Y}_n^{\mathfrak{R},\pi} = g(X_T^\pi) \\ Z_i^{\mathfrak{R},\pi} := \mathbb{E}[Y_{i+1}^{\mathfrak{R},\pi} H_i \mid \mathcal{F}_{t_i}], \\ \tilde{Y}_i^{\mathfrak{R},\pi} := \mathbb{E}[Y_{i+1}^{\mathfrak{R},\pi} \mid \mathcal{F}_{t_i}] + h_i f(X_{t_i}^\pi, \tilde{Y}_i^{\mathfrak{R},\pi}, Z_i^{\mathfrak{R},\pi}), \\ Y_i^{\mathfrak{R},\pi} := \tilde{Y}_i^{\mathfrak{R},\pi} \mathbf{1}_{\{t_i \notin \mathfrak{R}\}} + \mathcal{P}(X_{t_i}^\pi, \tilde{Y}_i^{\mathfrak{R},\pi}) \mathbf{1}_{\{t_i \in \mathfrak{R}\}}, \end{cases}$$

with $H_i = \frac{W_{t_{i+1}} - W_{t_i}}{t_{i+1} - t_i}$.

discrete-time approximation scheme

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with $H_i = \frac{W_{t_{i+1}} - W_{t_i}}{t_{i+1} - t_i}$.

↔ need a numerical approximation of conditional expectations !

Study of the approximation

- ▶ We can write the discretely reflected BSDE as a perturbed backward discrete-time scheme:

$$\begin{cases} \tilde{Y}_T^{\mathfrak{R}} = g(X_T^\pi) + \zeta^g \\ \tilde{Y}_{t_i}^{\mathfrak{R}} := \mathbb{E}[Y_{t_{i+1}}^{\mathfrak{R}} | \mathcal{F}_{t_i}] + h_i f(X_{t_i}^\pi, \tilde{Y}_{t_i}^{\mathfrak{R}}, \mathbb{E}[Y_{t_{i+1}}^{\mathfrak{R}} H_i | \mathcal{F}_{t_i}]) + \zeta_i^f, \\ Y_{t_i}^{\mathfrak{R}} := \tilde{Y}_{t_i}^{\mathfrak{R}} \mathbf{1}_{\{t_i \notin \mathfrak{R}\}} + (\mathcal{P}(X_{t_i}^\pi, \tilde{Y}_{t_i}^{\mathfrak{R}}) + \zeta_i^c) \mathbf{1}_{\{t_i \in \mathfrak{R}\}}, \end{cases}$$

Study of the approximation

- ▶ We can write the discretely reflected BSDE as a perturbed backward discrete-time scheme:

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- ▶ We introduce an optimal switching problem representation for backward discrete-time schemes
- ▶ We benefit from dimension 1 and comparison results for schemes
- ▶ Problem: no comparison result for the previous backward discrete-time scheme !

A modified discrete-time scheme

$$\begin{cases} Z_i^{\mathfrak{R},\pi} := \mathbb{E}[Y_{i+1}^{\mathfrak{R},\pi} H_i^R \mid \mathcal{F}_{t_i}], \\ \tilde{Y}_i^{\mathfrak{R},\pi} := \mathbb{E}[Y_{i+1}^{\mathfrak{R},\pi} \mid \mathcal{F}_{t_i}] + h_i f(X_{t_i}^\pi, \tilde{Y}_i^{\mathfrak{R},\pi}, Z_i^{\mathfrak{R},\pi}), \\ Y_i^{\mathfrak{R},\pi} := \tilde{Y}_i^{\mathfrak{R},\pi} \mathbf{1}_{\{t_i \notin \mathfrak{R}\}} + \mathcal{P}(X_{t_i}^\pi, \tilde{Y}_i^{\mathfrak{R},\pi}) \mathbf{1}_{\{t_i \in \mathfrak{R}\}}, \end{cases}$$

where

$$(H_i^R)^\ell = \frac{-R}{h_i} \vee \frac{W_{t_{i+1}}^\ell - W_{t_i}^\ell}{h_i} \wedge \frac{R}{h_i}, \quad 1 \leq \ell \leq d,$$

with R a positive parameter small enough.

Error between discrete-time BSDE and discretely reflected BSDE

$|\mathfrak{R}|$: Mesh of the grid \mathfrak{R}

$|\pi|$: Mesh of the grid π

Proposition

Under same assumptions than for the existence and uniqueness theorem, we have

$$\begin{aligned} & \sup_{0 \leq i \leq n} \mathbb{E} \left[|\tilde{Y}_{t_i}^{\mathfrak{R}} - \tilde{Y}_i^{\mathfrak{R}, \pi}|^2 \right] + |\mathfrak{R}| \mathbb{E} \left[\sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} |Z_s^{\mathfrak{R}} - Z_i^{\mathfrak{R}, \pi}|^2 ds \right] \\ & \leq C_R \left(|\pi|^{1/2} + |\mathfrak{R}|^{-1} |\pi| \right). \end{aligned}$$

Comparison with [Chassagneux-Élie-Kharroubi '12]

- ▶ Almost the same speed of convergence than in [Chassagneux-Élie-Kharroubi '12] but without the very restrictive assumption: $f^i(x, y, z) = f^i(x, y^i)$.
- ▶ When the generator depends on z , [Chassagneux-Élie-Kharroubi '12] obtain a bad speed of convergence by a direct geometric approach

$$\begin{aligned} & \sup_{0 \leq i \leq n} \mathbb{E} \left[|\tilde{Y}_{t_i}^{\mathfrak{R}} - \tilde{Y}_i^{\mathfrak{R}, \pi}|^2 \right] + \mathbb{E} \left[\sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} |Z_s^{\mathfrak{R}} - Z_i^{\mathfrak{R}, \pi}|^2 ds \right] \\ & \leq CL^{2\kappa} \left(|\pi|^{1/2} + |\mathfrak{R}|^{-1} |\pi| \right) \end{aligned}$$

with $L > 1$ the Lipschitz constant of the projection operator $\mathcal{P}(x, \cdot)$. This is the main difference with normal projection where $L \leq 1$.

Main result

Theorem

(i) Taking $|\mathfrak{R}| \sim |\pi|^{1/2}$, we have

$$\sup_{0 \leq i \leq n} \mathbb{E} \left[|Y_{t_i} - \tilde{Y}_i^{\mathfrak{R}, \pi}|^2 + |Y_{t_i} - Y_i^{\mathfrak{R}, \pi}|^2 \right] \leq C |\pi|^{1/2} \log(2T/|\pi|).$$

(ii) Taking $|\mathfrak{R}| \sim |\pi|^{1/3}$, we have

$$\sup_{0 \leq i \leq n} \mathbb{E} \left[|Y_{t_i} - \tilde{Y}_i^{\mathfrak{R}, \pi}|^2 + |Y_{t_i} - Y_i^{\mathfrak{R}, \pi}|^2 \right] \leq C |\pi|^{1/3} \log(2T/|\pi|),$$

and

$$\mathbb{E} \left[\sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} |Z_s - Z_i^{\mathfrak{R}, \pi}|^2 ds \right] \leq C |\pi|^{1/6} \sqrt{\log(2T/|\pi|)}.$$