

Numerical approximation of obliquely reflected BSDEs

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Introduction

Obliquely reflected BSDEs

Application: Switching problem.

Representation result

A Discretization scheme for RBSDEs

BTZ scheme - (implicit Euler)

Approximation of the RBSDE

Previous results [CEK12]

Convergence for the obliquely RBSDE

L^2 -stability for the scheme

Convergence results

Conclusion

Outline

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Reflected BSDEs - Markovian setting

- ▶ For $(b, \sigma) : \mathbb{R}^d \rightarrow \mathbb{R}^d \times M^d$ Lipschitz (σ may be degenerate) :

$$X_t = X_0 + \int_0^t b(X_u) du + \int_0^t \sigma(X_u) dW_u$$

- ▶ 'Simply' reflected BSDEs on a boundary $l(X)$, (Y, Z, K)

$$Y_t = g(X_T) + \int_t^T f(X_t, Y_t, Z_t) dt - \int_t^T (Z_t)' dW_t + \int_t^T dK_t$$

(C1) $Y_t \geq l(X_t)$ (constrained value process)

(C2) $\int_0^T (Y_t - l(X_t)) dK_t = 0$ ("optimality" of K)

- ▶ Extension: doubly reflected BSDEs, reflected BSDEs in convex domain \hookrightarrow *normal reflection*

Geometric framework

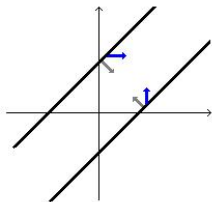
- Multidimensional process Y constrained in a domain \mathcal{C} ($d \geq 2$)

$$\mathcal{C} = \{ y \in \mathbb{R}^d \mid y^i \geq \max_j (y_j - c_{ij}) =: \mathcal{P}^i(y), 1 \leq i \leq d \}$$

with $c_{ii} = 0$, $\inf_{i \neq j} c_{ij} > 0$, $c_{ij} + c_{jk} > c_{ik}$

$\Leftrightarrow \mathcal{P}$ (oblique projection) is L -lipschitz with $L > 1$ (euclidean norm)
for this talk: c_{ij} are constant coefficients

- example $d = 2$, oblique direction of reflection



Obliquely reflected BSDEs

- ▶ System of reflected BSDEs: for $1 \leq i \leq d$,

$$Y_t^i = g^i(X_T) + \int_t^T f^i(X_u, Y_u, Z_u^i) du - \int_t^T (Z_u^i)' dW_u + \int_t^T dK_s^i$$

(C1) $Y_t \in \mathcal{C}$ (constrained by K)

(C2) $\int_0^T (Y_t^i - \mathcal{P}^i(Y_t)) dK_t^i = 0$ ('optimality' of K)

- ▶ Existence and uniqueness:

(H1): $f^i(y, z) = f^i(y^i, z^i)$, Hu & Tang 07, Hamadene & Zhang 08

(H2): $f^i(y, z) = f^i(y, z^i)$, C.-Elie-Kharroubi 11

Starting and Stopping problem (1)

Hamadene 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 (at $t = 0$, for $\alpha_0 = 1$)

$$Y_0^1 := \sup_a \mathbb{E}[J(a, 0)]$$

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

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

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

- ▶ 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)

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

From “Switching problem” to “switched BSDEs”

- ▶ “Switching” strategy $a = (\alpha_j, \theta_j)_j$ starting at (i, t)
 $N^a = \#\{k \in \mathbb{N}^* | \theta_k \leq T\}$ (number of switch)

- ▶ State process - cost process

$$a_s = \alpha_0 \mathbf{1}_{0 \leq s \leq \theta_0} + \sum_{j=1}^{N^a} \alpha_{j-1} \mathbf{1}_{\theta_{j-1} < s \leq \theta_j}, \quad A_s^a := \sum_{j=1}^{N^a} c_{\alpha_{j-1}, \alpha_j} \mathbf{1}_{\theta_j \leq s \leq T}$$

- ▶ “Switched” BSDE (following the strategy a) under **(H1)**

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

- ▶ Representation (a^* : optimal strategy) under **(H1)**

$$Y_t^i = \operatorname{esssup}_a U_t^a = U_t^{a^*}$$

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- ▶ We are given an equidistant grid
 $\pi = \{0 = t_0 < \dots < t_i < \dots < t_n = T\}$, define $h = T/n$.

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- ▶ The scheme: given the terminal condition $Y_n = g(X_T)$, the transition from step $i + 1$ to i is

$$Y_i := \mathbb{E}_{t_i}[Y_{i+1} + hf(Y_i, Z_i)]$$

$$Z_i := \mathbb{E}_{t_i}\left[\frac{\Delta W_i}{h}(Y_{i+1})'\right]$$

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\Leftrightarrow in practice: approximation of the forward process/ estimation of the conditional expectation.

Deriving the scheme (1/2) - Y part

On the equidistant grid $\pi = \{0 = t_0 < \dots < t_i < \dots < t_n = T\}$, with $h = T/n$.

- ▶ Start with:

$$Y_{t_i} + \int_{t_i}^{t_{i+1}} Z_s dW_s = Y_{t_{i+1}} + \int_{t_i}^{t_{i+1}} f(Y_s, Z_s) ds \quad (1)$$

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$$Y_{t_i} \simeq \mathbb{E}_{t_i}[Y_{t_{i+1}} + hf(Y_{t_i}, Z_{t_i})]$$

$$\Leftrightarrow Y_i := \mathbb{E}_{t_i}[Y_{i+1} + hf(Y_i, Z_i)]$$

Deriving the scheme (2/2) - Z part

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- ▶ For the Z-part:

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

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

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$$\hookrightarrow Z_i := \mathbb{E}_{t_i} [H_i Y_{i+1}] \quad \text{with} \quad H_i := h^{-1} \Delta W_i .$$

A scheme for the RBSDE: an example with $f = 0$

- ▶ RBSDE, Snell envelop of $I(X_t)$:

$$Y_t = I(X_T) - \int_t^T (Z_u)' dW_u + \int_t^T dK_s, \quad Y_t \geq I(X_t)$$

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- ▶ Given a grid $\mathfrak{R} = \{0 = r_0 < \dots < r_k < \dots < r_\kappa = T\}$, \bar{Y} discrete Snell envelop of $(I(X_{r_k}))_k$:

$$\tilde{Y}_{r_k} := \mathbb{E} [\bar{Y}_{r_{k+1}} \mid \mathcal{F}_{r_k}]$$

$$\bar{Y}_{r_k} := \tilde{Y}_{r_k} \vee I(X_{r_k})$$

\Leftrightarrow terminal condition $\bar{Y}_n = \tilde{Y}_n := I(X_T)$.

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\Leftrightarrow terminal condition $\bar{Y}_n = \tilde{Y}_n := I(X_T)$.

- ▶ More general domain/reflection:

$$\bar{Y}_{r_k} := \tilde{Y}_{r_k} \vee I(X_{r_k}) \rightarrow \bar{Y}_{r_k} = \mathcal{P}(\tilde{Y}_{r_k})$$

Moonwalk scheme for the RBSDE

- ▶ Implicit Euler scheme for the “BSDE part” :

$$\tilde{Y}_i := \mathbb{E} [Y_{i+1} \mid \mathcal{F}_{t_i}] + hf(\tilde{Y}_i, Z_i)$$

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- ▶ Taking into account the reflection

$$Y_i := \tilde{Y}_i \mathbf{1}_{t_i \notin \mathfrak{R}} + \mathcal{P}(\tilde{Y}_i) \mathbf{1}_{t_i \in \mathfrak{R}}$$

$\mathfrak{R} \subset \pi$ is the reflection grid with κ dates.

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- ▶ and terminal condition $Y_n = \tilde{Y}_n := g(X_{t_n})$.

Previous results and known difficulties

- ▶ In C.-Elie-Kharroubi 12, the convergence of the above method with a polynomial rate is proved under the following assumption on f

$$f^\ell(y, z) = f^\ell(y^\ell) .$$

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- ▶ In CEK12, the absence of z in f comes really from the final approximation i.e. the scheme: **stability issue**.

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Stability - perturbation approach

Recall the scheme,

$$\begin{aligned}\tilde{Y}_i &:= \mathbb{E} \left[Y_{i+1} + hf(\tilde{Y}_i, Z_i) \mid \mathcal{F}_{t_i} \right] \\ Z_i &:= \mathbb{E} \left[Y_{i+1} H'_i \mid \mathcal{F}_{t_i} \right] \\ Y_i &:= \tilde{Y}_i \mathbf{1}_{t_i \notin \mathfrak{R}} + \mathcal{P}(\tilde{Y}_i) \mathbf{1}_{t_i \in \mathfrak{R}}\end{aligned}$$

We consider a perturbed (for the Y -part) scheme:

$$\begin{aligned}\tilde{\mathcal{Y}}_i &:= \mathbb{E} \left[\mathcal{Y}_{i+1} + hf(\tilde{\mathcal{Y}}_i, \mathcal{Z}_i) + \zeta_i \mid \mathcal{F}_{t_i} \right] \\ \mathcal{Z}_i &:= \mathbb{E} \left[\mathcal{Y}_{i+1} H'_i \mid \mathcal{F}_{t_i} \right] \\ \mathcal{Y}_i &:= \tilde{\mathcal{Y}}_i \mathbf{1}_{t_i \notin \mathfrak{R}} + \mathcal{P}(\tilde{\mathcal{Y}}_i) \mathbf{1}_{t_i \in \mathfrak{R}}\end{aligned}$$

set $\delta Y := \mathcal{Y} - Y$, $\delta \tilde{Y} := \tilde{\mathcal{Y}} - \tilde{Y}$ and $\delta Z := \mathcal{Z} - Z$.

Stability - Definition

The scheme is said to be L^2 -stable if

$$\max_i \mathbb{E} \left[|\delta Y_i|^2 + |\delta \tilde{Y}_i|^2 \right] + \frac{1}{\alpha(\kappa)} \sum_i h \mathbb{E} [|\delta Z_i|^2] \leq C \sum_{i=0}^{n-1} h \mathbb{E} \left[\frac{1}{h^2} |\zeta_i|^2 \right]$$

i.e. the overall error is the sum of the error done at each step. α positive non decreasing function.

- (Sufficient Condition for L^2 -Stability) If f is Lipschitz-continuous and $f^\ell(y, z) = f^\ell(y, z^\ell)$, then the scheme is L^2 -stable with $\alpha(\kappa) = \kappa$.

Key points of the proof

- No reflection, stability OK: Difficulty from the reflection: $\delta \tilde{Y}_i \leftrightarrow \delta Y_i$?

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- Usual proof by induction on π using Euclidian norm & identity:

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\hookrightarrow But $|\delta Y_i|_2 := |\mathcal{P}(\tilde{Y}_i) - \mathcal{P}(\tilde{Y}_i)|_2 \leq L|\tilde{Y}_i - \tilde{Y}_i|_2 =: L|\delta \tilde{Y}_i|_2$, $L > 1$

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- Nevertheless, we observe that \mathcal{P} is 1-Lipschitz for $|\cdot|_\infty$ and use:

$$|\delta \tilde{Y}_{i-1}|_\infty^2 \leq |\delta Y_i|_\infty^2 + \sum_{j=1}^d (2\delta \tilde{Y}_{i-1}^j(\delta \tilde{Y}_{i-1}^j - \delta Y_i^j) - |\delta Y_i^j - \delta \tilde{Y}_{i-1}^j|^2) \mathbf{1}_{\{j=j^*\}}$$

where $j^* = \min\{j \mid |\delta \tilde{Y}_{i-1}^j| = \max_k |\delta \tilde{Y}_{i-1}^k|\}$.

Key object - Discretely reflected BSDEs (dRBSDEs)

Given a grid $\mathfrak{R} = \{0 = r_0 < \dots < r_k < \dots < r_\kappa = T\}$,
 a triplet (Y^d, \tilde{Y}^d, Z^d) satisfying

$$Y_T^d = \tilde{Y}_T^d := g(X_T)$$

and, for $j \leq \kappa - 1$ and $t \in [r_j, r_{j+1})$,

$$\begin{cases} \tilde{Y}_t^d &= Y_{r_{j+1}}^d + \int_t^{r_{j+1}} f(X, \tilde{Y}^d, Z^d) du - \int_t^{r_{j+1}} (Z_u^d)' dW_u, \\ Y_t^d &= \tilde{Y}_t^d \mathbf{1}_{\{t \notin \mathfrak{R}\}} + \mathcal{P}(\tilde{Y}_t^d) \mathbf{1}_{\{t \in \mathfrak{R}\}}. \end{cases}$$

Example: simply reflected on $l(X)$, set $f = 0$, $g = l \dots$ dRBSDE is the discrete Snell envelop of $l(X_r)_{r \in \mathfrak{R}}$

Strategy: first step

I. Approximate the discretely reflected BSDE:

1. *truncation error*: regularity of the discretely RBSDE (minimal Lipschitz condition, σ can be degenerate)
 \hookrightarrow Use a representation result for $Z^{\mathfrak{R}}$ based on Malliavin Calculus.
2. *global error*: Control via stability result.
 \hookrightarrow See discretely RBSDE as a pertubed scheme.

We obtain

$$\text{Err}(Y^{\mathfrak{R}}, Y^{\pi}) + \text{Err}(Z^{\mathfrak{R}}, Z^{\pi}) \leq C(\kappa^{\frac{1}{2}} h^{\frac{1}{2}} + h^{\frac{1}{2}}).$$

Strategy: second step

II. Extension to the continuously reflected BSDE:

1. Discretely RBSDE is a good proxy for continuously BSDEs:

$$\mathcal{E}rr(Y, Y^{\mathfrak{R}}) + \mathcal{E}rr(Z, Z^{\mathfrak{R}}) \leq C\kappa^{-\frac{1}{2}}$$

2. Use the scheme for the discretely RBSDE setting $\kappa = O(h^{-\frac{1}{2}})$ to obtain :

$$\mathcal{E}rr(Y, Y^{\pi}) + \mathcal{E}rr(Z, Z^{\pi}) \leq Ch^{\frac{1}{4}} .$$

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$$0 \leq (Y_t)^i - (Y_t^{\mathfrak{R}})^i \leq U_t^a - U_t^{\bar{a}}, \quad \bar{a} \text{ is the projection of strategy } a \text{ on } \mathfrak{R}$$

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2. Under **(H2)**: no comparison...

\hookrightarrow force comparison! Introduce (\check{Y}, \check{Z}) "max" of (Y, Z) and $(Y^{\mathfrak{R}}, Z^{\mathfrak{R}})$ and then

$$|(Y_t)^i - (Y_t^{\mathfrak{R}})^i| \leq |(\check{Y}_t)^i - (Y_t)^i| + |(\check{Y}_t)^i - (Y_t^{\mathfrak{R}})^i|$$

Conclude by using 1.

Concluding Remarks

- ▶ New convergence result when f depends on z
- ▶ Based on a careful study of the scheme L^2 stability
- ▶ Empirical Scheme:
 - "Low" regularity \leftrightarrow Use 1) quantization or 2) regression method.
 - Full convergence for 1) follows "easily" from key stability result for the scheme.