

# Adaptive weak approximation of reflected diffusions

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# Outline

- 1 Introduction
  - Skorohod problem
  - Kolmogorov equation
  - Euler scheme
- 2 An adaptive algorithm for reflected diffusions
  - The minimization problem for an adaptive scheme
  - Adaptive algorithm
- 3 Numerical example
  - Problem description
  - Implementation
  - Results

# The Skorohod problem

Given a domain  $D \subset \mathbb{R}^d$ , consider an SDE reflected at the boundary  $\partial D$ .

$$\begin{cases} dX_t^x = V(X_t^x)dt + \sum_{i=1}^d V_i(X_t^x)dB_t^i + n(X_t^x)dZ_t^x, \\ X_0^x = x \in D, \quad Z_0^x = 0. \end{cases} \quad (1)$$

- $B_t = (B_t^1, \dots, B_t^d)$   $d$ -dimensional Brownian motion
- $V, V_1, \dots, V_d : \bar{D} \rightarrow \mathbb{R}^d$  vector fields
- $X_t^x \in \bar{D}$ ,  $n(X_t^x)$  inward pointing normal vector at  $X_t^x \in \partial D$
- $Z_t^x$  increasing process with  $dZ_t^x = \mathbf{1}_{\partial D}(X_t^x)dZ_t^x$  ("local time")

# Kolmogorov equation

$$\begin{cases} \frac{\partial}{\partial t} u(t, x) = -Lu(t, x), & (t, x) \in [0, T] \times D, \\ u(T, x) = f(x), & x \in D, \\ \frac{\partial}{\partial n} u(t, x) = h(x), & x \in \partial D. \end{cases} \quad (2)$$

$L$  is the infinitesimal generator of the SDE.

## Proposition

*Under suitable regularity condition, the solution of (2) has the stochastic representation*

$$u(t, x) = E \left[ f(X_T) - \int_t^T h(X_s) dZ_s \mid X_t = x \right]. \quad (3)$$

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# The Euler scheme for reflected diffusions

Given a partition  $0 = t_0 < t_1 < \dots < t_N$ , let  $\Delta t_i = t_{i+1} - t_i$ ,  
 $\Delta B_i = B_{t_{i+1}} - B_{t_i}$ .

## Algorithm

① Set  $\bar{X}_0 = x$ ,  $\bar{Z}_0 = 0$ ,  $i = 0$ .

②  $\hat{X}_{i+1} = \bar{X}_i + V(\bar{X}_i)\Delta t_i + \sum_{j=1}^d V_j(\bar{X}_i)\Delta B_i^j$

③ If  $\hat{X}_{i+1} \in \bar{D}$  set

$$\bar{X}_{i+1} = \hat{X}_{i+1}, \quad \bar{Z}_{i+1} = \bar{Z}_i,$$

else set

$$\bar{X}_{i+1} = \Pi(\hat{X}_{i+1}), \quad \bar{Z}_{i+1}$$

④ If  $i < N - 1$  increase  $i$  by 1 and return to (1), else stop.

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# The Euler scheme for reflected diffusions – 2

- Order of convergence:  $u(0, x) - \bar{u}(0, x) = \mathcal{O}(\sqrt{\Delta t_{\max}})$
- Algorithm by Gobet [2001] based on approximation of domain by half-spaces (order 1 for particular direction of reflection)
- Algorithm by Bossy, Gobet, Talay [2004] with order 1 for  $h \equiv 0$



# Turning adaptive

- Goal: given a tolerance level TOL, minimize the computational work such that the computational error remains smaller than TOL.
- Concentrate *only* on the error of the time discretization of the SDE, not the integration error
- Adaptive algorithms for SDEs (without reflection) given by Szepessy, Tempone and Zouraris [2001]; we concentrate on the error introduced by the reflection.

# The minimization problem

- The error representation has leading order term

$$E \left[ \sum_{i=0}^{N-1} \Delta \bar{Z}_i^2 \left| \frac{\partial^2}{\partial n^2} u(t_{i+1}, \bar{X}_{i+1}) \right| \right]. \quad (4)$$

- Corresponding work given by  $\tilde{N}$ , the number of hits at the boundary. Use

$$E[\tilde{N}] \approx E \left[ \sum_{i=0}^{N-1} \frac{\Delta \bar{Z}_i}{\sqrt{\Delta t_i}} \right]. \quad (5)$$

- Problem: minimize (5) given that (4) is bounded by TOL among the set of partitions of  $[0, T]$ .
- Solve by Lagrangian relaxation.

# The refinement algorithm

Given a partition  $t_0 = 0 < t_1 < \dots < t_N = T$  and corresponding values for  $\Delta B_i$ ,  $\Delta \bar{Z}_i$ ,  $\bar{X}_i$ ,  $i = 0, \dots, N - 1$ .

## Algorithm

**For**  $i = 0, \dots, N$  **do**:

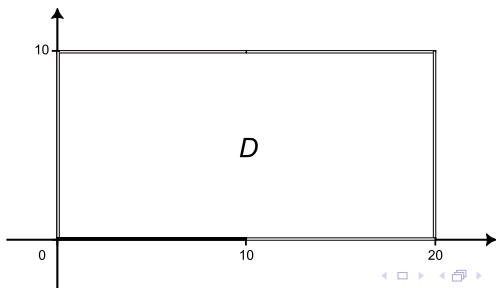
- *Compute*

$$\ell_i = \frac{1}{2\sqrt{2\pi}} E(\tilde{N}) \left| \frac{\partial^2}{\partial n^2} v(t_i, \bar{X}_i) \right| \exp\left(-\frac{d(\bar{X}_i)^2}{2\Delta t_i}\right) \Delta t_i.$$

- *If  $\ell_i > \text{TOL}$  refine the mesh by inserting  $t = \frac{t_{i+1} + t_i}{2}$  in the partition. Sample the corresponding value of the Brownian motion  $B_t$  using the Brownian bridge.*

# A mixed boundary condition

$$\left\{ \begin{array}{l} \frac{\partial}{\partial t} u(t, x) = -\frac{1}{2} \Delta u(t, x), \quad (t, x) \in [0, 2] \times D, \\ u(2, x) = 10 \exp\left(-\sqrt{(10 - x_1)^2 + x_2^2}\right), \quad x \in D, \\ \frac{\partial}{\partial n} u(t, x) = x_1, \quad (t, x) \in [0, 2] \times D_N, \\ u(t, x) = 10 \exp\left(-\sqrt{(10 - x_1)^2 + x_2^2}\right), \quad x \in \times(\partial D \setminus D_N). \end{array} \right.$$



# Implementation

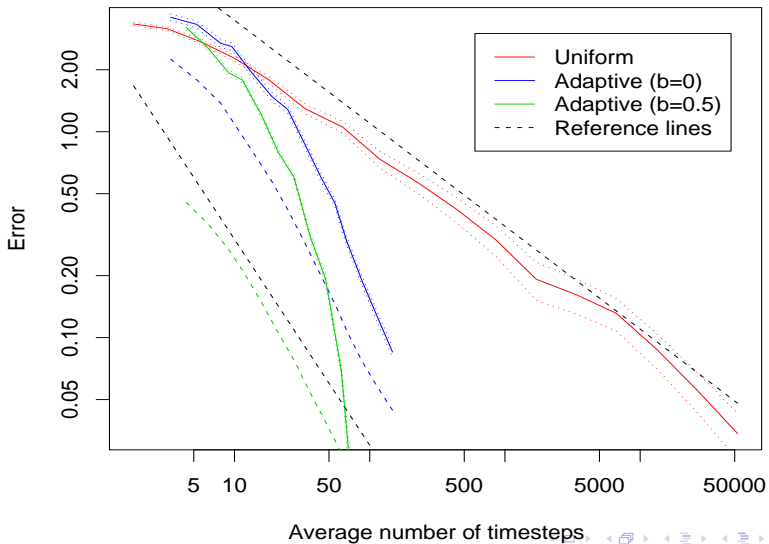
- Combine the adaptive Euler scheme for reflected diffusions with the adaptive Euler scheme for stopped diffusions by Dzougoutov et al. [2005].
- Approximate

$$\left| \frac{\partial^2}{\partial n^2} v(t_i, \bar{X}_i) \right| \approx \frac{1}{\|\bar{X}_i - x_{\text{sing}}\|^\beta + \text{TOL}^\alpha},$$





where  $x_{\text{sing}} = (10, 0)$ .

- Used  $\alpha = 2$  and  $\beta = 0, 1/2$ .

## Result



# References

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-  E. Gobet. *Euler schemes and half-space approximation for the simulation of diffusion in a domain.*, ESAIM Prob. Statist. 5, 2001.