Uncertain Darcy's problem and the stochastic particle transport Semester Project - Master in CSE

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Problem statement

Underground flow → Uncertain Darcy's problem

$$\begin{cases} u = -A\nabla p, & \text{in } D, \\ \nabla \cdot u = f, & \text{in } D, \\ p = p_0, & \text{on } \Gamma_{in}, \\ p = 0, & \text{on } \Gamma_{out}, \\ \nabla p \cdot n = 0, & \text{on } \Gamma_N, \end{cases}$$

Stochastic particle transport \rightarrow Ornstein–Uhlenbeck process

$$\begin{cases} dX(t) = u(X(t))dt + \sigma dW(t), & 0 \le t \le T, \\ X(0) = X_0 \in D, \end{cases}$$

Outline

- Expected exit time from a domain
- ► Theoretical investigation: perturbed SDE's
- ► The uncertain Darcy's problem

Mean exit time. Setting

Given a domain $D \subset \mathbb{R}^d$, $f: D \to \mathbb{R}^d$, $g: D \to \mathbb{R}^{d \times m}$ and an m-dimensional standard Wiener process W(t), consider

$$\begin{cases} dX(t) = f(X(t))dt + g(X(t))dW(t), & 0 < t \le T, \\ X(0) = X_0, & X_0 \in D. \end{cases}$$

The equation is defined in a domain D o boundary conditions

- \blacktriangleright killing boundaries: X(t) is stopped,
- reflecting boundaries: X(t) is reflected normally inside D.

Mean exit time. Problem statement

<u>Problem</u>. Estimate the exit time of the trajectories

$$\tau = \min\{\tau_e, T\}, \text{ where } \tau_e = \min\{t \colon X(t) \notin D\}.$$

Another quantity of interest, given $F: D \to \mathbb{R}$

$$\varphi = \varphi(T, X_0, F) = \mathbb{1}_{\{T < \tau_e\}} F(X(T)).$$

If $F \equiv 1$, exit probability

$$\Phi(T,X_0) := \Pr(\tau < T | X(0) = X_0) = 1 - \mathbb{E}(\varphi(T,X_0,1)).$$

<u>Goal</u>. Estimate numerically τ and φ .

Discrete Euler-Maruyama (DEM)

Method:

$$\begin{cases} X_h^d(t_{i+1}) = f(X(t_i))h + g(X(t_i))(W(t_{i+1}) - W(t_i)), \\ X_h^d(0) = X_0. \end{cases}$$

Parameters of interest computed naively

$$\begin{split} \tau_h^d &= \min\{\tau_{h,e}^d, T\}, \text{ where } \tau_{h,e}^d = \min\{t_i \colon X_h^d(t_i) \notin D\}, \\ \varphi_h^d &= \mathbbm{1}_{\{T < \tau_{h,e}^d\}} F(X_h^d(T)). \end{split}$$

Missed exits $\Rightarrow 1/2$ loss in weak order:

$$|\mathbb{E}(\tau_h^d) - \mathbb{E}(\tau)| = O(\sqrt{h}),$$

$$|\mathbb{E}(\varphi_h^d) - \mathbb{E}(\varphi)| = O(\sqrt{h}).$$

Continuous Euler-Maruyama (CEM)

<u>Goal</u>. Restore the weak order of convergence 1 of Euler-Maruyama \Rightarrow Brownian bridge approach.

Method:

$$\begin{cases} X_h^c(t) = f(X(t_i))(t-t_i) + g(X(t_i))(W(t)-W(t_i)), & t_i < t \leq t_{i+1}, \\ X_h^c(0) = X_0. \end{cases}$$

Estimate at each time step the probability of exit. If D is an half-space

$$Pr(\exists t \in [t_{i}, t_{i+1}] \quad X_{h}^{d}(t) \notin D|X_{h}^{d}(t_{i}) = x_{i}, X_{h}^{d}(t_{i+1}) = x_{i+1})$$

$$= p(x_{i}, x_{i+1}, h)$$

$$= \exp\left(-2\frac{[n \cdot (x_{i} - z_{i})][n \cdot (x_{i+1} - z_{i})]}{hn \cdot (gg^{T}(x_{i})n)}\right).$$

Continuous Euler-Maruyama (CEM)

Parameters of interest. Given u a realization of U uniform r.v. in (0,1)

$$\begin{split} \tau_{h}^{c} &= \min\{T, \tau_{h,e}^{c}\}, \\ \text{where } \tau_{h,e}^{c} &= \min\{\tau_{h,e1}^{c}, \tau_{h,e2}^{c}\}, \\ \tau_{h,e1}^{c} &= \min\{t_{i} = hi \colon X_{h}(t_{i}) \notin D\}, \\ \tau_{h,e2}^{c} &= \min\{t_{i} = hi \colon u < p(x_{i-1}, x_{i}, h)\}, \\ \varphi_{h}^{c} &= \mathbb{1}_{\{T < \tau_{h,e}^{c}\}} F(X_{h}^{c}(T)). \end{split}$$

Weak order 1 is restored:

$$|\mathbb{E}(\tau_h^c) - \mathbb{E}(\tau)| = O(h),$$

$$|\mathbb{E}(\varphi_h^c) - \mathbb{E}(\varphi)| = O(h).$$

An adaptive procedure

<u>Goal</u>. Restore the weak order of convergence 1 of Euler-Maruyama \Rightarrow Adaptivity in space.

Setting. Consider $\sigma \in \mathbb{R}$, I the identity in $\mathbb{R}^{d \times d}$ and

$$\begin{cases} dX(t) = f(X(t))dt + \sigma IdW(t), & 0 < t \le T, \\ X(0) = X_0, & X_0 \in D. \end{cases}$$

<u>Idea</u>. Given $l \in \mathbb{N}$, $h_0 > 0$, adapt step size h in DEM as follows

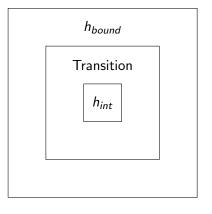
$$h = \max \left\{ h_{bound}, \min \left\{ h_{int}, \left(\frac{d}{(l+3)\sigma} \right)^2 \right\} \right\}.$$

where

$$h_{bound} = 2^{-2I}h_0,$$

 $h_{int} = 2^{-I}h_0.$

An adaptive procedure



Quel porco di mazzinga la vede.