

Hitting times of one dimensional diffusions and Monte Carlo approximation

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Outline of the presentation

- 1 Introduction
- 2 The forward method results
- 3 The backward method results
- 4 Some heuristics

One dimensional diffusion and exit time

In this talk, we consider a one dim. uniformly elliptic diffusion

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

and its exit time from the domain $D = (-\infty, L)$ with

$$\tau^x := \inf \{v \geq 0 : X_v^x \geq L\}, \quad \tau_t^x := \tau^x \wedge t.$$

We are interested in studying the collection of linear maps

$$P_t h(u, x) := \mathbb{E} \left[h(u + \tau_t^x, X_{\tau_t^x}^x) \right]$$

Two cases : Forward $\sigma \in C_b^2, b \in C_b^1$ and Backward $\sigma \in C_b^\alpha, b \in \mathcal{M}_b$.

• Motivations :

- Financial mathematics : pricing and hedging of barrier options, ...
- Default risk : pricing and hedging with a structural approach.

Related literature

- **Numerical computation** : $\mathbb{E}[h(X_T^x)\mathbb{I}_{\{\tau^x>T\}}]$ for smooth domain D
 - Elliptic diff. : Gobet (00), discrete & continuous Euler scheme
 - Hypo-elliptic diff. : Gobet&Menozzi (03)

$$\mathcal{E}_d(T, h, \Delta, x) := \mathbb{E}[h(\tilde{X}_T^{x,\Delta})\mathbb{I}_{\{\tau_d^{x,\Delta}>T\}}] - \mathbb{E}[h(X_T^x)\mathbb{I}_{\{\tau^x>T\}}] \asymp C\sqrt{\Delta}$$

$$\mathcal{E}_c(T, h, \Delta, x) := \mathbb{E}[h(\tilde{X}_T^{x,\Delta})\mathbb{I}_{\{\tau_c^{x,\Delta}>T\}}] - \mathbb{E}[h(X_T^x)\mathbb{I}_{\{\tau^x>T\}}] \approx C\Delta$$

- discrete Euler scheme is easy to implement in any dimension.
- continuous Euler scheme : dimension 1, more delicate in higher dimension.

Probabilistic representation & Monte Carlo simulation

One of our main results is a **probabilistic representation** of $(P_t)_{t \geq 0}$ in the same spirit as *Bally & Kohatsu-Higa (AAP, 15)*

$$\mathbb{E}[f(X_T^x)] = e^{\lambda T} \mathbb{E}[f(\bar{X}_T^\pi) \prod_{k=0}^{N_T-1} \lambda^{-1} \theta_{\zeta_{k+1}-\zeta_k}(\bar{X}_{\zeta_k}^\pi, \bar{X}_{\zeta_{k+1}}^\pi)]$$

Unbiased Monte Carlo path simulation requires :

- $\{N_t; t \geq 0\}$ Poisson process with parameter λ , Jump times : $\zeta_1 < \dots < \zeta_n$
- Euler scheme with random times :

$$\bar{X}_{\zeta_{k+1}}^\pi = \bar{X}_{\zeta_k}^\pi + b(\bar{X}_{\zeta_k}^\pi)(\zeta_{k+1} - \zeta_k) + \sigma(\bar{X}_{\zeta_k}^\pi)(W_{\zeta_{k+1}} - W_{\zeta_k}).$$

$$\theta_t(x, y) := a''(y) + 2a'(y)H_1(a(x)t, y - x - b(x)t) + (a(y) - a(x))H_2(a(x)t, y - x - b(x)t) - b'(y) - (b(y) - b(x))H_1(a(x)t, y - x - b(x)t).$$

$$H_i(c, x) := (g(c, x))^{-1} \partial_x^i g(c, x), \quad i \geq 0.$$

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Theorem : Probabilistic representation

$\{N_t; t \geq 0\}$ Poisson process with parameter λ , Jump times : $\zeta_1 < \dots < \zeta_{N(T)}$

Euler scheme : $\bar{X}_{\zeta_{j+1}}^\pi = \bar{X}_{\zeta_j}^\pi + \sigma(\bar{X}_{\zeta_j}^\pi)(W_{\zeta_{j+1}} - W_{\zeta_j})$

$$\Gamma_N(x) = e^{\lambda T} \prod_{j=0}^{N-1} \lambda^{-1} \bar{\theta}_{\zeta_{j+1}-\zeta_j}(\bar{X}_{\zeta_j}^\pi, \bar{X}_{\zeta_{j+1}}^\pi) \in L^1(\Omega)$$

$$\bar{\Gamma}_N(x) = e^{\lambda T} \frac{(a(L) - a(\bar{X}_{\zeta_{N-1}}^\pi))}{a(\bar{X}_{\zeta_{N-1}}^\pi)} \prod_{j=0}^{N-2} \lambda^{-1} \bar{\theta}_{\zeta_{j+1}-\zeta_j}(\bar{X}_{\zeta_j}^\pi, \bar{X}_{\zeta_{j+1}}^\pi) \in L^1(\Omega)$$

Then, for all $h \in \mathcal{M}_b$,

$$\begin{aligned} \mathbb{E}[h(\tau_T^x, X_{\tau_T^x}^x)] &= e^{\lambda T} \mathbb{E}[h((\zeta_{N(T)} + \bar{\tau}^{\zeta_{N(T)}, \bar{X}_{\zeta_{N(T)}}^\pi}) \wedge T, \bar{X}_{(\zeta_{N(T)} + \bar{\tau}^{\zeta_{N(T)}, \bar{X}_{\zeta_{N(T)}}^\pi)} \wedge T}^\pi) \Gamma_N(T)(x)] \\ &+ e^{\lambda T} \mathbb{E}[h(\zeta_{N(T)-1} + \bar{\tau}^{\zeta_{N(T)-1}, \bar{X}_{\zeta_{N(T)-1}}^\pi}, L) \mathbb{I}_{\{\bar{\tau}^{\zeta_{N(T)-1}, \bar{X}_{\zeta_{N(T)-1}}^\pi} \leq T - \zeta_{N(T)-1}\}} \bar{\Gamma}_N(T)(x)]. \end{aligned}$$

Weight definitions

Here, we define

$$\theta_t(x, z) = \mathbb{I}_{\{x < L\}} \left\{ \left(\frac{1}{2} a''(z) - b'(z) \right) + (a'(z) - b(z)) \mu_t^1(x, z) + \frac{1}{2} (a(z) - a(x)) \mu_t^2(x, z) \right\},$$

$$\bar{\theta}_t(x, z) = \theta_t(x, z) \Lambda_t(x, z)$$

$$\Lambda_t(x, z) := \mathbb{P} \left(\sup_{0 \leq v \leq t} W_v \leq \frac{L-x}{\sigma(x)} \mid W_t = \frac{z-x}{\sigma(x)} \right) = \left\{ 1 - e^{-2 \frac{(L-x)(L-x-(z-x))}{a(x)t}} \right\} \mathbb{I}_{\{x \leq L\}} \mathbb{I}_{\{z \leq L\}}$$

$$\mu_t^1(x, z) = H_1(a(x)t, z-x) - \frac{1}{a(x)t} \frac{2(L-x)}{\left(\exp\left(-\frac{2(z-L)(L-x)}{a(x)t}\right) - 1 \right)},$$

$$\mu_t^2(x, z) = H_2(a(x)t, z-x) + \frac{1}{a^2(x)t^2} \frac{4(z-L)(L-x)}{\left(\exp\left(-\frac{2(z-L)(L-x)}{a(x)t}\right) - 1 \right)}.$$

$$\bar{\tau}^x \sim f_{\bar{\tau}}(x, s) = \partial_s \mathbb{P}(\bar{\tau}^x \leq s) = \frac{L-x}{\sqrt{2\pi a(x)s^{3/2}}} \exp\left(-\frac{(L-x)^2}{2a(x)s}\right) \mathbb{I}_{\{x \leq L\}}.$$

Corollary : Integration by parts for killed process

Let $T > 0$. Then, for all $x \in (-\infty, L)$, one has

$$\begin{aligned} & \mathbb{E}[h'(X_T^x) \mathbb{I}_{\{\tau^x > T\}}] \\ &= -\mathbb{E} \left[h(\bar{X}_T^\pi) \Lambda_{T-\zeta_{N(T)}}(\bar{X}_{\zeta_{N(T)}}^\pi, \bar{X}_T^\pi) \mu_{T-\zeta_{N(T)}}^1(\bar{X}_{\zeta_{N(T)}}^\pi, \bar{X}_T^\pi) \Gamma_{N(T)}(x) \right]. \end{aligned}$$

For all $x \in (-\infty, L]$, the joint law of $(\tau_T^x, X_{\tau_T^x}^x)$ is given by the measure

$$p_T(x, dt, dz) := p^K(x, t) \delta_L(dz) dt + p_T^D(x, z) \delta_T(dt) dz$$

and, for some positive $C, c > 1$, for all $(t, z) \in (0, T] \times (-\infty, L]$:

$$p^K(x, t) \leq Ct^{-1/2} g(ct, L-x) \quad \text{and} \quad p_T^D(x, z) \leq Cg(cT, z-x),$$

$z \mapsto p_T^D(x, z) \in \mathcal{C}^{1+\alpha}((-\infty, L])$, $t \mapsto p^K(x, t) \in \mathcal{C}^{\alpha/2}((0, T])$, for all $\alpha < 1$.

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$$\vartheta_t(z, x) := \frac{1}{2}(a(x) - a(z))\hat{\mu}_t^2(z, x) + b(x)\hat{\mu}_t^1(z, x)$$

$$\hat{\mu}_t^1(x, z) := H_1(a(z)t, z - x) - \frac{1}{a(z)t} \frac{2(L - z)}{\left(\exp\left(\frac{2(L-z)(L-x)}{a(z)t}\right) - 1\right)}$$

$$\hat{\mu}_t^2(x, z) := H_2(a(z)t, z - x) - \frac{1}{a^2(z)t^2} \frac{4(L - z)(L - x)}{\left(\exp\left(\frac{2(L-z)(L-x)}{a(x)t}\right) - 1\right)}.$$

$$\bar{\vartheta}_t(z, x) := \vartheta_t(z, x)\Lambda_t(z, x)$$

$$\hat{\vartheta}_t^i(z, x) := \begin{cases} \frac{\bar{q}_t^z(x, z)}{\bar{q}_t^x(x, z)} & i = 0, \\ \bar{\vartheta}_t(z, x) \frac{\bar{q}_t^z(x, z)}{\bar{q}_t^x(x, z)} & i = 1, \dots, n - 1. \end{cases}$$

Probabilistic representation in the Backward method

Let $T > 0$. Assume that $a \in C^\alpha$ and $b \in \mathcal{M}_b$. Define

$$\bar{\Gamma}_N^D(z) = e^{\lambda T} \lambda^{-N} \prod_{i=0}^{N-1} \bar{\vartheta}_{\zeta_{i+1}-\zeta_i}(\bar{X}_{\zeta_i}^{\pi,z}, \bar{X}_{\zeta_{i+1}}^{\pi,z})$$

$$\bar{\Gamma}_N^K(x) := e^{\lambda T} \lambda^{-N} \mathbb{I}_{\{\zeta_N < t < T\}} \hat{\vartheta}_L(\bar{X}_{\zeta_N}^{\pi,x}, t - \zeta_N) \prod_{i=0}^{N-1} \hat{\vartheta}_{\zeta_{i+1}-\zeta_i}^i(\bar{X}_{\zeta_{i+1}}^{\pi,x}, \bar{X}_{\zeta_i}^{\pi,x})$$

Then, for all $(t, x) \in (0, T] \times (-\infty, L)$, the following probabilistic representation holds

$$p_T^D(x, z) = \mathbb{E} \left[\bar{q}_{T-\zeta_N}^{\bar{X}_{\zeta_N}^{\pi,z}}(x, \bar{X}_{\zeta_N}^{\pi,z}) \bar{\Gamma}_N^D(z) \right],$$

$$p^K(x, t) = \mathbb{E} \left[f_{\bar{T}}^L(\bar{X}_{\zeta_N}^{\pi,x}, t - \zeta_N) \bar{\Gamma}_N^K(x) \right].$$

Corollary : Integration by parts formula

Let $T > 0$. Assume that $a \in C^\alpha$ and $b \in \mathcal{M}_b$.

$$\partial_x p_T^D(x, z) = \mathbb{E} \left[\bar{q}_{T-\zeta_N}^{\bar{X}_{\zeta_N}^{\pi, z}}(x, \bar{X}_{\zeta_N}^{\pi, z}) \hat{\mu}_{T-\zeta_N}^1(x, \bar{X}_{\zeta_N}^{\pi, z}) \bar{\Gamma}_N^D(z) \right].$$

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Markov maps

$$\begin{aligned}
 P_t h(u, x) &= \mathbb{E}[h(u + \tau_t^x, X_{\tau_t^x}^x)] \\
 &= h(u, x) \mathbb{I}_{\{x \geq L\}} \\
 &\quad + \mathbb{I}_{\{x < L\}} \left\{ \mathbb{E}[h(u + t, X_t^x) \mathbb{I}_{\{\tau^x > t\}}] + \mathbb{E}[h(u + \tau^x, L) \mathbb{I}_{\{\tau^x \leq t\}}] \right\}.
 \end{aligned}$$

Proxy semigroup with frozen diffusion coefficient : For $y \in \mathbb{R}$, consider

$$\bar{X}_t^x = x + \sigma(y) W_t, \quad \bar{\tau}^x = \inf \{t \geq 0 : \bar{X}_t^x \geq L\}, \quad y \in \mathbb{R}.$$

and

$$\bar{P}_t^y h(u, x) = \mathbb{E}[h(u + \bar{\tau}_t^x, \bar{X}_{\bar{\tau}_t^x}^x)].$$

Infinitesimal generators

From Itô's rule, one obtains :

Proposition : Infinitesimal generators

For $h \in C_b^{1,2}(\mathbb{R}_+ \times]-\infty, L]) \cap \mathcal{M}_b(\mathbb{R}_+ \times \mathbb{R})$, one has

$$\begin{aligned} \frac{(P_t h - h)(u, x)}{t} &\xrightarrow[t \downarrow 0]{} \mathcal{L}h(u, x) \\ &:= \mathbb{I}_{\{x < L\}} \left(b(x) \partial_2 h(u, x) + \frac{1}{2} a(x) \partial_2^2 h(u, x) + \partial_1 h(u, x) \right), \\ \frac{(\bar{P}_t^y h - h)(u, x)}{t} &\xrightarrow[t \downarrow 0]{} \bar{\mathcal{L}}^y h(u, x) \\ &:= \mathbb{I}_{\{x < L\}} \left(\frac{1}{2} a(y) \partial_2^2 h(u, x) + \partial_1 h(u, x) \right). \end{aligned}$$

The proxy semigroup

$$\begin{aligned}
 \bar{P}_t^y h(u, x) &= \mathbb{I}_{\{x \geq L\}} h(u, x) + \mathbb{I}_{\{x < L\}} \mathbb{E}[h(u + t, \bar{X}_t^x) \mathbb{I}_{\{\bar{\tau}^x > t\}}] + \mathbb{I}_{\{x < L\}} \mathbb{E}[h(u + \bar{\tau}^x, L) \mathbb{I}_{\{\bar{\tau}^x \leq t\}}] \\
 &= \mathbb{I}_{\{x \geq L\}} h(u, x) + \mathbb{I}_{\{x < L\}} \int_{-\infty}^L h(u + t, z) \bar{q}_t^y(x, z) dz + \mathbb{I}_{\{x < L\}} \int_0^t h(u + s, L) f_{\bar{\tau}}^y(x, s) ds \\
 &= \mathbb{I}_{\{x \geq L\}} h(u, x) + \mathcal{S}_t^y h(u, x) + \mathcal{K}_t^y h(u, x) \\
 &= \mathbb{I}_{\{x \geq L\}} h(u, x) + \mathbb{I}_{\{x < L\}} \int_u^{u+t} \int_{-\infty}^L h(s, z) [\delta_{u+t}(ds) \bar{q}_t^y(x, z) dz + \delta_L(dz) f_{\bar{\tau}}^y(x, s - u) ds].
 \end{aligned}$$

with (from the reflection principle)

$$\bar{q}_t^y(x, z) = \frac{\mathbb{P}(\bar{X}_t^x \in dz, \sup_{v \in [0, t]} \bar{X}_v^x < L)}{dz} = (g(a(y)t, z - x) - g(a(y)t, z + x - 2L)) \mathbb{I}_{\{x, z \leq L\}},$$

$$f_{\bar{\tau}}^y(x, s) = \partial_s \mathbb{P}(\bar{\tau}^x \leq s) = \frac{L - x}{\sqrt{2\pi a(y) s^3/2}} \exp\left(-\frac{(L - x)^2}{2a(y)s}\right) \mathbb{I}_{\{x \leq L\}}.$$

Forward parametrix expansion

From now on, we freeze the coefficient to the **starting point** : $y = x$.

- **Forward parametrix expansion** consists in writing

$$\begin{aligned}
 P_T h(u, x) - \bar{P}_T h(u, x) &= \int_0^T \partial_s (\bar{P}_{T-s} P_s) h(u, x) ds = \int_0^T \bar{P}_{T-s} (\mathcal{L} - \bar{\mathcal{L}}) P_s h(u, x) ds \\
 &\stackrel{\text{IBP}}{=} \int_0^T ds \left(\bar{K}_{T-s} P_s h(u, x) + \bar{S}_{T-s} P_s h(u, x) \right)
 \end{aligned}$$

with

$$\bar{K}_t h(u, x) = \bar{K}_t(x, L) h(u+t, L) \quad \text{and} \quad \bar{S}_t h(u, x) = \int_{-\infty}^L h(u+t, z) \bar{S}_t(x, z) dz$$

with

$$\bar{K}_t(x, L) = \mathbb{I}_{\{x < L\}} \frac{(a(L) - a(x))}{a(x)} f_{\bar{T}}^x(x, t)$$

$$\bar{S}_t(x, z) = \mathbb{I}_{\{x < L\}} \left\{ \partial_z^2 \left[\frac{1}{2} (a(z) - a(x)) \bar{q}_t^x(x, z) \right] - \partial_z [b(z) \bar{q}_t^x(x, z)] \right\}$$

Expansion of the semigroup

Iterating the one step expansion, we get

$$\begin{aligned}
 P_T h(u, x) &= \bar{P}_T h(u, x) + \int_0^T ds_1 \{ \bar{K}_{T-s_1} h(u, x) + \bar{S}_{T-s_1} P_{s_1} h(u, x) \} \\
 &= \bar{P}_T h(u, x) + \int_0^T ds_1 \{ \bar{K}_{T-s_1} h(u, x) + \bar{S}_{T-s_1} \bar{P}_{s_1} h(u, x) \} \\
 &+ \int_0^T ds_1 \int_0^{s_1} ds_2 \{ \bar{K}_{T-s_1} \bar{K}_{s_1-s_2} P_{s_2} h(u, x) + \bar{K}_{T-s_1} \bar{S}_{s_1-s_2} P_{s_2} h(u, x) \} \\
 &+ \int_0^T ds_1 \int_0^{s_1} ds_2 \{ \bar{S}_{T-s_1} \bar{K}_{s_1-s_2} P_{s_2} f(x, u) + \bar{S}_{T-s_1} \bar{S}_{s_1-s_2} P_{s_2} h(u, x) \} \\
 &= \bar{P}_T h(u, x) + \int_0^T ds_1 \{ \bar{K}_{T-s_1} h(u, x) + \bar{S}_{T-s_1} \bar{P}_{s_1} h(u, x) \} \\
 &+ \int_0^T ds_1 \int_0^{s_1} ds_2 \{ \bar{S}_{T-s_1} \bar{K}_{s_1-s_2} P_{s_2} f(x, u) + \bar{S}_{T-s_1} \bar{S}_{s_1-s_2} P_{s_2} h(u, x) \} \\
 &= \bar{P}_T h(u, x) + \sum_{n=1}^N \int_{\Delta_n(T)} ds_n \bar{S}_{T-s_1} \bar{S}_{s_1-s_2} \cdots \bar{S}_{s_{n-2}-s_{n-1}} \bar{K}_{s_{n-1}-s_n} \bar{P}_{s_n} h(u, x) \\
 &+ \sum_{n=1}^N \int_{\Delta_n(T)} ds_n \bar{S}_{T-s_1} \bar{S}_{s_1-s_2} \cdots \bar{S}_{s_{n-1}-s_n} \bar{P}_{s_n} h(u, x) + \mathcal{R}_T^N h(u, x)
 \end{aligned}$$

Defining

$$I_{s_0}^n h(u, x) = \begin{cases} \int_{\Delta_n(s_0)} d\mathbf{s}_n \left\{ \left(\prod_{i=0}^{n-1} \bar{S}_{s_i - s_{i+1}} \right) \bar{P}_{s_n} h(u, x) + \left(\prod_{i=0}^{n-2} \bar{S}_{s_i - s_{i+1}} \right) \bar{K}_{s_{n-1} - s_n} h(u, x) \right\}, & n \geq 1, \\ \bar{P}_{s_0} h(u, x), & n = 0, \end{cases}$$

one concludes

Theorem : Semigroup expansion

Let $T > 0$. For every $h \in C_b^{1,2}(\mathbb{R} + \times] - \infty, L]) \cap \mathcal{B}_b(\mathbb{R}_+ \times \mathbb{R})$ and for every $(t, u, x) \in [0, T] \times \mathbb{R}_+ \times \mathbb{R}$, one has

$$P_t h(u, x) = \sum_{n \geq 0} I_t^n h(u, x).$$

Then, the probabilistic interpretation follows...