

Exact simulation of Brownian diffusions with drift admitting jumps

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Paul Painlevé



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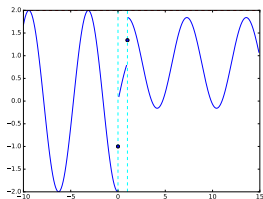
- 1 Brownian motion with discontinuous drift
- 2 The two-skew Brownian motion
- 3 Generalized and retrospective rejection sampling schemes
- 4 Some numeric examples

Real valued Brownian diffusion with discontinuous drift

Let $b \in C^1(\mathbb{R} \setminus \{0, z\})$,

$$dX_t = b(X_t) dt + dW_t, \quad t \in [0, T] \quad \longrightarrow \quad \mathbb{P}_b \in \mathcal{P}(\mathcal{C}([0, T]; \mathbb{R}))$$

Example:



Intensity of the jumps:

$$\theta_1 = \frac{b(0^+) - b(0^-)}{2}, \quad \theta_2 = \frac{b(z^+) - b(z^-)}{2}.$$

Value at the discontinuities:

$$b(0) := \frac{b(0^+) + b(0^-)}{2}, \quad b(z) := \frac{b(z^+) + b(z^-)}{2}$$

Aim (Exact simulation of \mathbb{P}_b)

One can sample directly from \mathbb{P}_b without approximating b by continuous drifts.

Exact simulation, retrospective rejection sampling scheme

For $b \in \mathcal{C}_b^1(\mathbb{R})$ [Beskos, Roberts, Papaspiliopoulos (2008)];

The exact simulation scheme consists in

- $\Omega \in \mathcal{P}(\mathcal{C}([0, T]; \mathbb{R}))$ easy to sample from,
- a functional of the path of the form $\int_0^T \varphi(X_t) dt$ with $0 \leq \varphi \leq M$,

such that

$$\mathbb{P}_b(dX) \propto e^{-\int_0^T \varphi(X_t) dt} \Omega(dX)$$

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$$\mathbb{P}_b(dX) \propto e^{-\int_0^T \varphi(X_t) dt} \Omega(dX)$$

Proposition (Retrospective rejection sampling)

- 1 Simulate a Poisson Point Process on the rectangle $[0, T] \times [0, M]$ with intensity $\lambda = 1$, and obtain $\{(t_i, y_i), i = 1, \dots, N\}$
- 2 sample the marginals $(X_{t_i})_{i \leq N}$ under the measure Ω . Obtain $(x_i)_{i \leq N}$
- 3 accept if $y_i \geq \varphi(x_i)$ for all i , otherwise reject and repeat.

Output: the finite-dimensional marginals of \mathbb{P}_b :

$$\{X_{t_1}, X_{t_2}, \dots, X_{t_N}, X_T; 0 \leq t_1 < t_2 < \dots < t_N \leq T\}.$$

Exact algorithm if the drift b has two (or more) jumps

- It is easy to find \mathfrak{Q} , but not to sample from it.

$$\mathfrak{Q}(dX) \propto e^{B(X_T) - B(X_0) - \theta_1 L_T^0(X) - \theta_2 L_T^z(X)} \mathbb{P}(dX), \quad B(x) := \int_0^x b(u) du.$$

- The function $\varphi := \frac{1}{2}(b^2 + b') + c$ a.e., c constant such that $\varphi \geq 0$.

Exact algorithm if the drift b has two (or more) jumps

- It is easy to find Ω , but not to sample from it.

$$\Omega(dX) \propto e^{B(X_T) - B(X_0) - \theta_1 L_T^0(X) - \theta_2 L_T^z(X)} \mathbb{P}(dX), \quad B(x) := \int_0^x b(u) du.$$

- The function $\varphi := \frac{1}{2}(b^2 + b') + c$ a.e., c constant such that $\varphi \geq 0$.

Suppose $\theta_2 = 0$. Ω is BM conditioned on $(X_T, L_T^0(X)) \sim h(y, \ell) dy d\ell$

$$h(y, \ell) dy d\ell \propto \exp(B(y) - B(x_0) - \theta_1 \ell) \mathbb{P}(X_T \in dy, L_T^0 \in d\ell).$$

Remark (Two approaches)

- Obtain the rejection sampling as the limit of rejection sampling schemes for one-skew BM with drift b . [Étoré and Martinez (2013)]
- Sample jointly Brownian motion and its local times [Papaspiliopoulos, Roberts and Taylor (2015)]

Heuristics on the (β_1, β_2) -skew BM

Let $\beta_1, \beta_2 \in [-1, 1]$

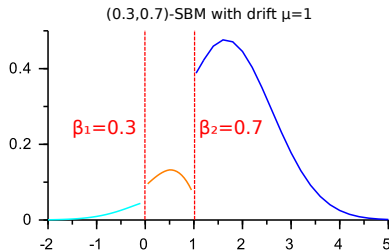


$$\begin{cases} dX_t = dW_t + \mu dt + \beta_1 dL_t^0(X) + \beta_2 dL_t^z(X); \\ L_t^0(X) = \int_0^t \mathbb{I}_{\{X_s=0\}} dL_s^0(X), \quad L_t^z(X) = \int_0^t \mathbb{I}_{\{X_s=z\}} dL_s^z(X) \end{cases}$$

Example:

$$y \mapsto p_1^{(0.3, 0.7)}(1, 0.5, y)$$

is the density of X_1 where $(X_t)_t$ is the $(0.3, 0.7)$ -skew BM with drift $\mu = 1$ starting at $X_0 = 0.5$.



The transition density of the (β_1, β_2) -skew BM

Proposition (without drift)

$$p^{(\beta_1, \beta_2)}(t, x, y) = p(t, x, y) \cdot v^{(\beta_1, \beta_2)}(t, x, y)$$

where z is the distance between the barriers,

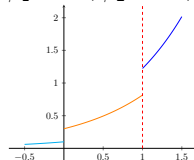
$$v^{(\beta_1, \beta_2)}(t, x, y) = \sum_{k=0}^{\infty} (-\beta_1 \beta_2)^k \sum_{j=1}^4 c_j(y) e^{-\frac{(a_j(x, y) + 2z k)^2}{2t}} e^{-|x-y| \frac{a_j(x, y) + 2z k}{t}}$$

$$\begin{cases} c_1(y) \equiv 1 \\ c_2(y) = (2\mathbb{1}_{[0, +\infty)}(y) - 1) \beta_1 \\ c_3(y) = (2\mathbb{1}_{[z, +\infty)}(y) - 1) \beta_2 \\ c_4(y) = (1 - 2\mathbb{1}_{[0, z)}(y)) \beta_1 \beta_2 \end{cases} \quad \begin{cases} a_1(x, y) \equiv 0 \\ a_2(x, y) = |y| + |x| - |y - x| \\ a_3(x, y) = |y - z| + |y - z| - |y - x| \\ a_4(x, y) = 2(z - \max(x, y, 0))^+ + 2(\min(x, y, z))^+ \end{cases}$$

Representation of the transition density

$$\text{Let } k(x) = \begin{cases} \frac{1}{4}(1 - \beta_1)(1 - \beta_2)e^{2\mu x} & x < 0, \\ \frac{1}{4}(1 + \beta_1)(1 - \beta_2)e^{2\mu x} & 0 \leq x < z, \\ \frac{1}{4}(1 + \beta_1)(1 + \beta_2)e^{2\mu x} & x \geq z. \end{cases}$$

$$\beta_1 = 0.2, \beta_2 = 0.5, \mu = 0.5$$



Lemma

The infinitesimal generator \mathcal{L} is self adjoint in $L^2(k(x)dx)$, and

$$p_{\mu}^{(\beta_1, \beta_2)}(t, x, y) = \frac{1}{2\pi i} \int_{\Gamma} e^{\lambda t} G_{\mu}^{(\beta_1, \beta_2)}(x, y; \lambda) d\lambda,$$

where Γ is a complex contour of $\sigma(\mathcal{L}) \subseteq (-\infty, 0]$,

and $G_{\mu}^{(\beta_1, \beta_2)}$ is the Green's function of the resolvent $(\lambda - \mathcal{L})^{-1}$.

If $\mu = 0$, this is known as the Titchmarsh-Kodaira-Yoshida method.

The Green's function for the (β_1, β_2) -SBM with drift

Let $y \in \mathbb{R}$ and $\lambda \in \mathbb{C} \setminus (-\infty, 0]$, the Green's function solves

$$\begin{cases} (\lambda - \mathcal{L})G(x) = \delta_{\{y\}}(x), & G \in \mathcal{C}^2(\mathbb{R} \setminus \{0, z\}) \cap \mathcal{C}(\mathbb{R}) \\ k(0^+)u'(0^+) = k(0^-)u'(0^-), & k(z^+)u'(z^+) = k(z^-)u'(z^-). \end{cases}$$

Let the change of variable: $\lambda \xrightarrow{\phi} \sqrt{2\lambda + \mu^2} =: w \in \{v \in \mathbb{C}; \Re(v) > 0\}$

Lemma (The Green's function)

$$G_{\mu}^{(\beta_1, \beta_2)}(x, y; w) = \frac{e^{\mu(y-x)}}{w} \frac{\sum_{j=1}^4 c_j(\mu, y; w) e^{-w(a_j(x,y) + |x-y|)}}{\beta_1 \beta_2 e^{-2wz} (w^2 - \mu^2) + (w + \beta_1 \mu)(w + \beta_2 \mu)}.$$

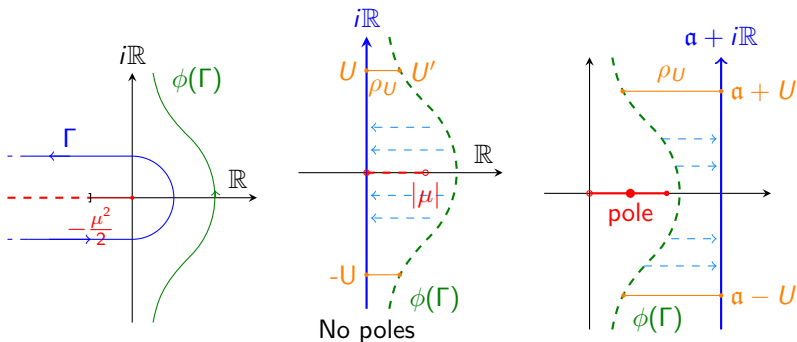
$c_j(\mu, y; w) = w^2 c_{j,0}(y) + w \mu c_{j,1}(y) + \mu^2 c_{j,2}(y)$, where

$$\begin{cases} c_{1,0} = 1, \\ c_{2,0} = (2\mathbb{1}_{[0,+\infty)}(y) - 1) \beta_1 \\ c_{3,0} = (2\mathbb{1}_{[z,+\infty)}(y) - 1) \beta_2 \\ c_{4,0} = (1 - 2\mathbb{1}_{[0,z)}(y)) \beta_1 \beta_2 \end{cases} \quad \begin{cases} c_{1,1} = \beta_1 + \beta_2 \\ c_{2,1} = -\beta_1 - c_{4,0} \\ c_{3,1} = -\beta_2 + c_{4,0} \\ c_{4,1} = 0 \end{cases} \quad \begin{cases} c_{1,2} = \beta_1 \beta_2 \\ c_{2,2} = \beta_1 c_{3,0} \\ c_{3,2} = -\beta_2 c_{2,0} \\ c_{4,2} = -c_{4,0}. \end{cases}$$

Ideas of the proof

$$p_{\mu}^{(\beta_1, \beta_2)}(t, x, y) = \frac{1}{2\pi i} \int_{\Gamma} e^{\frac{1}{2}\phi(\lambda)^2 t} e^{-\frac{1}{2}\mu^2 t} G_{\mu}^{(\beta_1, \beta_2)}(x, y; \phi(\lambda)) d\lambda,$$

and we use techniques of complex analysis:



The transition density for the (β_1, β_2) -SBM with drift

Proposition

$$p_{\mu}^{(\beta_1, \beta_2)}(t, x, y) = p_{\mu}(t, x, y) v_{\mu}^{(\beta_1, \beta_2)}(t, x, y)$$

where the function $v_{\mu}^{(\beta_1, \beta_2)}$ is given by a series of Fourier transforms. They involve two functions

$$\begin{cases} \mathcal{J}_0(\omega, \beta_i \mu \sqrt{t}) := \sqrt{2\pi} e^{-\frac{\omega^2}{2}} e^{\frac{1}{2}(\omega + \beta_i \mu \sqrt{t})^2} \Phi^c(\omega + \beta_i \mu \sqrt{t}) \text{ with } i \in \{1, 2\} \\ \mathcal{J}_1(\omega) := -e^{-\frac{\omega^2}{2}}, \end{cases}$$

evaluated in $\omega \in \left\{ \omega_{j,k} := \frac{a_j(x,y) + 2zk + |y-x|}{\sqrt{t}}, j = 1, 2, 3, 4, k \in \mathbb{N} \right\}$.

The explicit representation of transition density...

Let $\mathbf{a} \geq 0$, $-2\beta_1\mu$, $-2\beta_2\mu$. The function $v_\mu^{(\beta_1, \beta_2)}$, which does not depend on \mathbf{a} , admits the following series representation

$$v_\mu^{(\beta_1, \beta_2)}(t, x, y) = \sum_{k=0}^{\infty} (-\beta_1\beta_2\mu^2 t)^k \sum_{j=1}^4 F_{j,k}(\omega_{j,k}, \mathbf{a}),$$

$$F_{j,k} := \begin{cases} \sum_{n=0}^k \binom{2k-n}{k} \frac{\mathfrak{C}_{j,k}(\mathbf{a})}{n!(\beta_1\mu\sqrt{t} - \beta_2\mu\sqrt{t})^{2k+1-n}} \mathcal{F}_{k+h-s, m, n}(\omega_{j,k}, \mathbf{a}), & \text{if } \beta_1 \neq \beta_2; \\ (-1)^{k+1} \frac{\mathfrak{C}_{j,k}(\mathbf{a})}{(2k+1)!} \mathcal{G}_{k+h-s, m, 2k+1}(\omega_{j,k}, \mathbf{a}\sqrt{t}, \beta_1\mu\sqrt{t}), & \text{if } \beta_1 = \beta_2; \end{cases}$$

where

$$\mathfrak{C}_{j,k}(\mathbf{a}) := e^{\frac{1}{2}\omega_{1,0}^2} \sum_{m=0}^k \sum_{s=0}^{k-m} \sum_{h=0}^2 \binom{k-m}{s} \binom{k}{m} \frac{(-2\mathbf{a}\sqrt{t})^{k-m-s} (\mu^2 - \mathbf{a}^2)^s C_{j,2-h}(y)}{\mu^{2k} t^{k-s}}$$

$$C_{j,0} := c_{j,0}, \quad C_{j,1} := \mu c_{j,1} + 2c_{j,0}\mathbf{a}, \quad C_{j,2} := c_{j,2}\mu^2 + c_{j,1}\mu\mathbf{a} + c_{j,0}\mathbf{a}^2.$$

The transition density... completion

$$\mathcal{F}_{K,m,n}(\omega, \mathbf{a}) := \mathcal{G}_{K,m,n}(\omega, \mathbf{a}\sqrt{t}, \beta_2\mu\sqrt{t}) - (-1)^n \mathcal{G}_{K,m,n}(\omega, \mathbf{a}\sqrt{t}, \beta_1\mu\sqrt{t}).$$

$$\mathcal{G}_{K,m,n}(\omega, \mathbf{a}, \tau) := \sum_{\ell=0}^{\lfloor \frac{K+m}{2} \rfloor} \frac{(-1)^{\ell+K}}{2^\ell} \frac{(K+m)!}{\ell!(K+m-2\ell)!} \mathcal{S}_{K+m-2\ell,n}(\omega, \mathbf{a}, \tau)$$

$$\mathcal{S}_{L,n}(\omega, \mathbf{a}, \tau) = \sum_{n'=0}^n \sum_{L'=0}^L \binom{n}{n'} \binom{L}{L'} (\omega + \tau)^{n-n'} (\mathbf{a} + \tau)^{L-L'} \mathcal{J}_{n'+L'}(\omega, \tau),$$

$$\mathcal{J}_q := \begin{cases} \mathcal{J}_0(\omega, \tau) = \sqrt{2\pi} e^{-\frac{1}{2}\omega^2} e^{\frac{1}{2}(\omega+\tau)^2} \Phi^c(\omega + \tau) & q = 0, \\ \mathcal{J}_1(\omega) = -e^{-\frac{1}{2}\omega^2} & q = 1, \\ \mathcal{J}_0(\omega, \tau)(q-1)!! - \mathcal{J}_1(\omega) \sum_{k=0}^{\frac{q}{2}-1} (\omega + \tau)^{q-2k-1} \frac{(q-1)!!}{(q-2k-1)!!} & q \geq 2 \text{ even}, \\ \mathcal{J}_1(\omega) \sum_{k=0}^{\frac{q-1}{2}} (\omega + \tau)^{(q-1-2k)} 2^k \frac{(\frac{q-1}{2}!) }{(\frac{q-1}{2}-k)!} & q \geq 3 \text{ odd}, \end{cases}$$

where $(2n+1)!! = (2n+1) \cdot (2n-1) \cdot \dots \cdot 3 \cdot 1$, $n \in \mathbb{N}$.

Sample the finite-dimensional distributions of Ω .

Propagation of the retrospective rejection sampling

Let $\mathbb{P}_b^{(\beta_1, \beta_2)}$ the distribution of the (β_1, β_2) -skew BM with drift b .

Lemma

Choosing $\beta_1 = \frac{1}{n}$, $\beta_2(n) = \frac{\theta_2}{\theta_1 n + b(0) - b(z)}$ and $\mu(n) = \theta_1 n + b(0)$

$$\mathbb{P}_b^{(\beta_1, \beta_2)}(dX) \propto e^{-\int_0^T \varphi(X_t) dt} \Omega^{(\beta_1, \beta_2)}(dX)$$

$$\Omega^{(\beta_1, \beta_2)}(dX) \propto \exp(B(X_T) - B(X_0) - \mu(n)(X_T - X_0)) \mathbb{P}_{\mu(n)}^{(\beta_1, \beta_2)}(dX)$$

$$\mathbb{P}_b^{(\beta_1, \beta_2)} \quad \beta_i(n) \rightarrow 0 \quad \mathbb{P}_b$$

$$\updownarrow \quad \quad \quad \updownarrow$$

$$\mathbb{P}_{\mu(n)}^{(\beta_1, \beta_2)} \quad \quad \quad \mathbb{P}$$

$$\Omega^{(\beta_1, \beta_2)} \quad \beta_i(n) \rightarrow 0 \quad \Omega$$

$$\updownarrow \quad \quad \quad \updownarrow$$

Sampling from $\Omega^{(\beta_1, \beta_2)}$

The finite-dimensional distributions of $\Omega^{(\beta_1, \beta_2)}$ converge to the finite-dimensional distributions of Ω , and are given by the product

$$h^{(\beta_1, \beta_2)}(\mathbf{y}) \prod_{i=0}^{N-1} q_{\mu}^{(\beta_1, \beta_2)}(t_i, t_{i+1}, T, y_i, y_{i+1}, y) dy_1 \dots dy_N dy,$$

where $q_{\mu}^{(\beta_1, \beta_2)}$ are the bridges of the skew BM with drift and

$$h^{(\beta_1, \beta_2)}(\mathbf{y}) \propto e^{B(y) - B(x_0)} p(T, x_0, y) v_{\mu}^{(\beta_1, \beta_2)}(T, x_0, y)$$

$$\frac{q_{\mu}^{(\beta_1, \beta_2)}(t', t, T, x', y, x)}{q(t', t, T, x', y, x)} = \frac{v_{\mu}^{(\beta_1, \beta_2)}(T - t, y, x) v_{\mu}^{(\beta_1, \beta_2)}(t - t', x', y)}{v_{\mu}^{(\beta_1, \beta_2)}(T - t', x', x)}$$

Lemma

There exists a pointwise limit $v^{(\theta_1, \theta_2)}(t, x, y)$ of $v_{\mu}^{(\beta_1, \beta_2)}(t, x, y)$ for $\beta_i \rightarrow 0$, $\beta_i \mu \rightarrow \theta_i$, $i = 1, 2$ where θ_i are the half jumps.

Sampling from Ω

The finite-dimensional distributions of $\Omega^{(\beta_1, \beta_2)}$ converge to the finite-dimensional distributions of Ω given by the product

$$h^{(\theta_1, \theta_2)}(\mathbf{y}) \prod_{i=0}^{N-1} q^{(\theta_1, \theta_2)}(t_i, t_{i+1}, T, y_i, y_{i+1}, y) dy_1 \dots dy_N dy,$$

where $q^{(\theta_1, \theta_2)}$ are the limit of the bridges of the skew BM with drift and

$$h^{(\theta_1, \theta_2)}(\mathbf{y}) \propto e^{B(y) - B(x_0)} p(T, x_0, y) v^{(\theta_1, \theta_2)}(T, x_0, y)$$

$$\frac{q^{(\theta_1, \theta_2)}(t', t, T, x', y, x)}{q(t', t, T, x', y, x)} = \frac{v^{(\theta_1, \theta_2)}(T - t, y, x) v^{(\theta_1, \theta_2)}(t - t', x', y)}{v^{(\theta_1, \theta_2)}(T - t', x', x)}$$

Lemma

There exists a pointwise limit $v^{(\theta_1, \theta_2)}(t, x, y)$ of $v_{\mu}^{(\beta_1, \beta_2)}(t, x, y)$ for $\beta_i \rightarrow 0$, $\beta_i \mu \rightarrow \theta_i$, $i = 1, 2$ where θ_i are the half jumps.

Generalized rejection sampling method

Rejection sampling method

Instrumental density: $g(x)dx$ of a random variable Y

Goal density: $h(x)dx$ of a random variable X

assume $\exists M > 0$ such that $f(x) := \frac{1}{M} \frac{h(x)}{g(x)} \leq 1$ for all $x \in \mathbb{R}$;

then $X \stackrel{(d)}{=} (Y | U < f(Y))$ is an exact simulation if f can be evaluated.

Generalization

If there exists a sequence of functions $(f_n)_n$

converging pointwise to f at a decreasing rate $(\delta_n)_n$.

Moreover for each $x \in \mathbb{R}$ $f_n(x), \delta_n(x)$ can be evaluated,

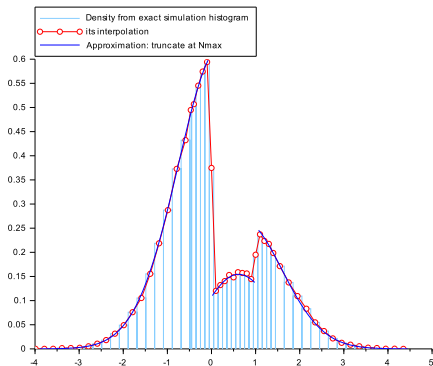
then $X \stackrel{(d)}{=} (Y | \exists n : U < f_n(Y) - \delta_n(Y))$.

Application of the generalized RS method at the 2-SBM

Lemma (Bound, uniform in x and y)

For each term of the series $v_\mu^{(\beta_1, \beta_2)}(t, x, y)$,

$$\sup_{x, y \in \mathbb{R}} \left| \left(v_\mu^{(\beta_1, \beta_2)} \right)_k(t, x, y) \right| \leq C_{\beta_1, \beta_2} e^{-\frac{2z^2}{t} k}$$



$y \mapsto p^{(-0.7, 0.3)}(1, 0.5, y)$ with drift $\mu = 0$:

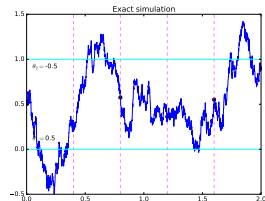
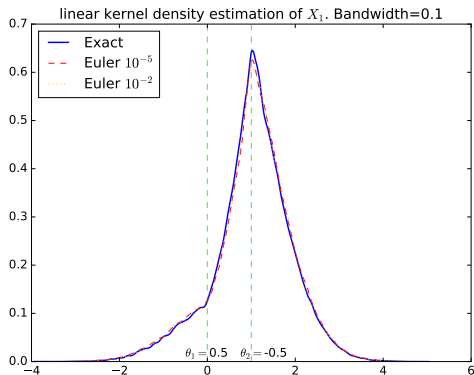
comparison between

-the truncation at the 10th term

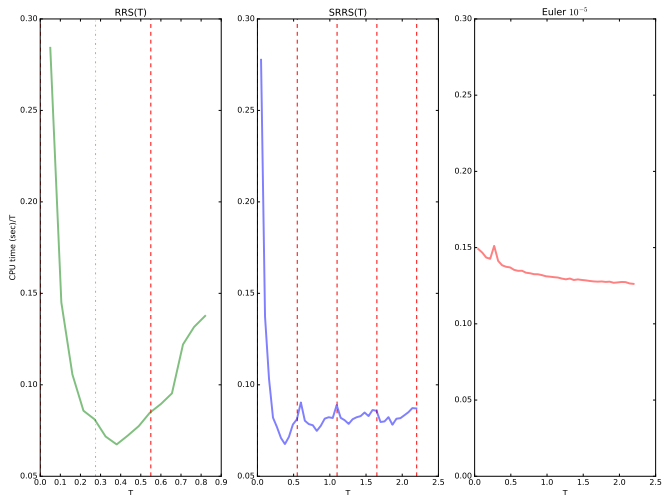
-the histogram for 50000 exact simulations through GRS.

Exact simulation of the BM with drift $b = \mathbb{1}_{(0,1)}$

The solution of $dX_t = dW_t + \mathbb{1}_{(0,1)}(X_t)dt$, $t \in [0, T]$,
with initial condition $X_0 = 0.5$,



CPU times



Sampling from the finite-dimensional distributions of Ω

$$\frac{h^{(\theta_1, \theta_2)}(y)}{p_0\left(\frac{T}{1-\delta}, x_0, y\right)} = \underbrace{\frac{C_{\theta, x_0, T}}{\sqrt{1-\delta}} e^{M_B} \bar{v}(T)}_{C^{\mathcal{H}}} \underbrace{\frac{e^{-\frac{(y-x_0)^2}{2T}\delta} e^{B(y)-B(x_0)}}{e^{M_B}} \frac{v^{(\theta_1, \theta_2)}(T, x_0, y)}{\bar{v}(T)}}_{f_\delta^{\mathcal{H}}(y)},$$

where $C_{\theta, x_0, T}$ is the normalizing constant for the density $h^{(\theta_1, \theta_2)}$

and $M_B \leq \frac{\|b\|_\infty^2 T}{2\delta}$ is an upper bound for $B(y) - B(x) - \frac{(y-x)^2 \delta}{2T}$.

An appropriate choice of δ , for each specific case, can make the bound sharper.

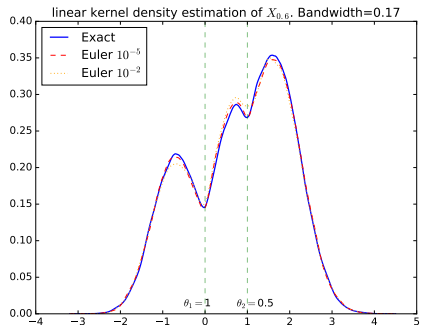
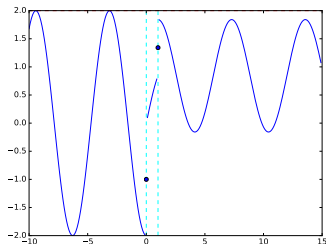
$$\frac{q^{(\theta_1, \theta_2)}(t, T, x_1, x_2, x)}{q(t, T, x_1, x_2, x)} = \underbrace{\frac{\bar{v}(t) \bar{v}(T-t)}{v^{(\theta_1, \theta_2)}(T, x_1, x_2)}}_{C^{\mathcal{B}}} \underbrace{\frac{v^{(\theta_1, \theta_2)}(t, x_1, x)}{\bar{v}(t)} \frac{v^{(\theta_1, \theta_2)}(T-t, x, x_2)}{\bar{v}(T-t)}}_{f_{x_1, x_2}^{\mathcal{B}}(x)}.$$

Another example

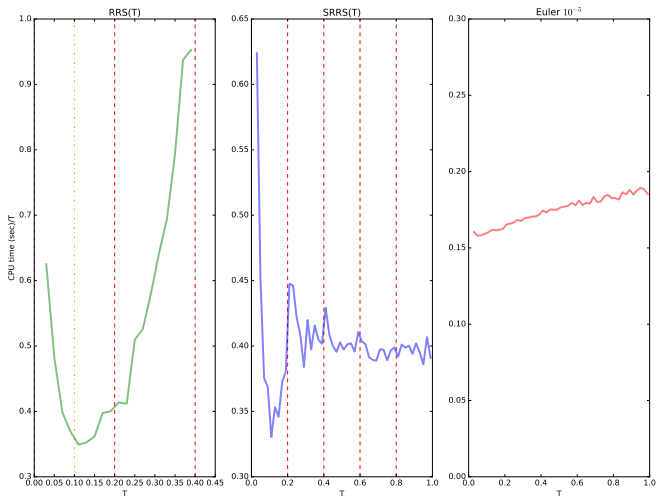
The solution of $dX_t = dW_t + b(X_t)dt$, $t \in [0, T]$; with initial condition $X_0 = 0.5$ for a piecewise continuous drift

$$b(x) = \begin{cases} -2 \cos(x) & x < 0 \\ \sin(x) & 0 < x < 1, \\ \cos(x - 1) + \sin(1) & x > 1 \end{cases}$$

The drift b



CPU times



Acknowledgements

Thank you all for your
attention!

Bibliography

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