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Discretisation scheme for time-inhomogeneous jump SDEs.



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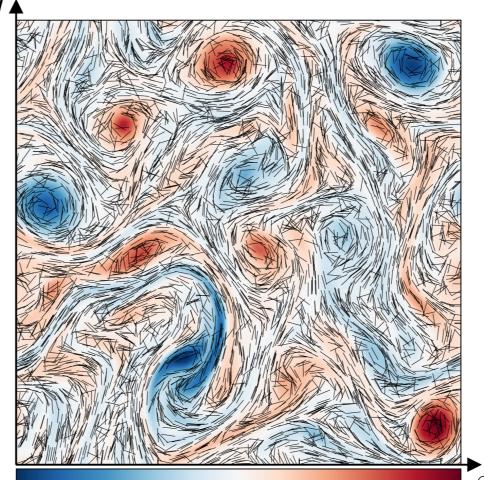
When they are driven by a Lévy process whose increments can be simulated exactly, the simulation of stochastic differential equation with jumps can be straightforward via an Euler scheme. However, this is only possible in a few restrictive cases, which arise the need for a general method that works for a larger class of SDEs. The approach taken here follows the ideas of Asmussen-Rosinski in [AR01], and consists in simulating exactly the large jumps of the process, while approximating the small jumps by Gaussian variables. We are inspired by the ideas of Fournier in [Fou12], using Wasserstein techniques to perform a strong error analysis.

L^p -wellposedness

Our generic SDE, defined for $t \in [0, T]$:

$$X_t = X_0 + \int_0^t a(s, X_s) ds + \int_0^t b(s, X_s) dW_s + \int_0^t \int_{-\infty}^\infty c(s, X_{s^-}, z) \tilde{N}(ds, dz), \tag{1}$$

Application : stochastic orientation of rods in 2D turbulence



We consider intertialless rods in a turbulent flow with position equation dX(t)/dt = v(X(t), t), coupled with a unit orientation vector p following Jeffery's equation :

$$\frac{d}{dt}p = \mathbb{A}p - (p^T \mathbb{A}p)p, \qquad (2)$$

where \mathbb{A} denotes the gradient tensor of the fluid-velocity v. The figure on the left, obtained in [CBB22], shows the vorticity and velocity fields obtained as Direct Numerical Simulation (DNS) of 2D Navier Stokes.

• N(ds, dz) is a random Poisson measure with time-dependent compensator $\nu_s(dz)ds$ • $\tilde{N}(ds, dz) = N(ds, dz) - \nu_s(dz)ds$ is the compensated Poisson measure. L^p -wellposedness framework of [BP20] with time-Hölder condition:

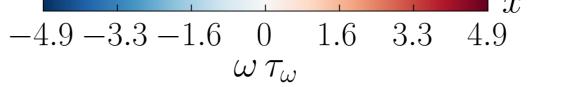
(H1) - Time-Hölder and Lipschitz: The drift and diffusion coefficients a(t, x) and b(t, x) are Hölder w.r.t. t and Lipschitz w.r.t. x. The jump coefficient c(t, x, z) verifies:

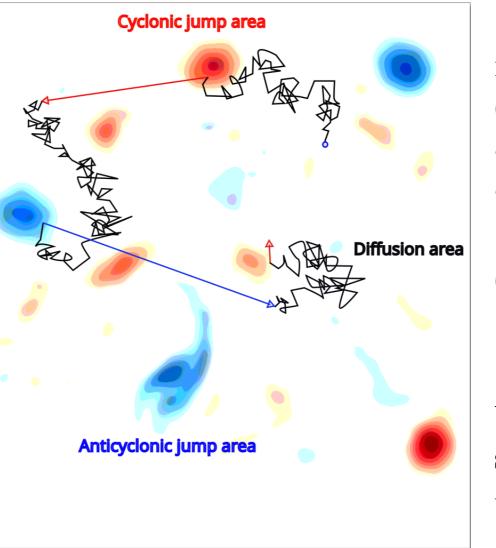
 $|c(t, x, z) - c(t, y, z)| \le L_c(t, z)|x - y|, \quad x, y \in \mathbb{R}, \ t \in [0, T].$

(H2) - Integrability: $X_0 \in L^p(\Omega)$ and

$$\int_0^T \left(\int_{-\infty}^\infty |F(t,z)|^2 \nu_t(dz) \right)^{p/2} dt < \infty, \quad \int_0^T \int_{-\infty}^\infty |F(t,z)|^p \nu_t(dz) < \infty,$$

for $F(\cdot, \cdot) = L_c(\cdot, \cdot)$ and $F(\cdot, \cdot) = c(\cdot, 0, \cdot)$.





After averaging on the gradient tensor \mathbb{A} at the equilibrium regime, a Brownian SDE followed by the unfolded angle $\theta_t = \arctan(p_2/p_1)$ has been derived in [CBB22]. To take in account the effects of the vertices as brutal variations of the angular displacement, we added a jump term leading to

$$\theta_t = \int_0^t a(\theta_s) ds + \int_0^t b(\theta_s) dW_s + \int_0^t \int_{-\infty}^\infty b(\theta_s) \tilde{N}(ds, dz),$$
(3)

where a(x) and b(x) being linear combining of $\cos(x)$ and $\sin(x)$, $\nu_s(dz) = (s^{1/2} \wedge T^*)|z|^{-1-\alpha} \mathbb{1}_{\{|z| < T^*\}}$, T^* being the average lifetime of the vertices.

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(4)

General framework for the simulation of a Poisson integral

The stochastic Poisson integral can be separated into *large jumps* and *small jumps*:

$$\int_{-\infty}^{+\infty} F(s,z)\tilde{N}(ds,dz) = \int_{0}^{T} \int_{\mathbb{R}\setminus B(\varepsilon)} F(s,z)\tilde{N}(ds,dz) + \int_{0}^{T} \int_{B(\varepsilon)} F(s,z) + \int_{0}^{T} \int_{B(\varepsilon)} F(s,z)\tilde{N}(ds,dz) + \int_{0}^{T} F(s,z)\tilde{N}(ds,dz) + \int_{0}^{T} F(s,z)\tilde{N}(ds,dz) + \int_{0}^{T} F(s,z)\tilde{N}(ds,dz) + \int_{0}^{T} F(s,z)$$

A. Numerically tractable representation of large jumps

$$\int_{0}^{T} \int_{\mathbb{R}\setminus B(\varepsilon)} F(s,z)N(ds,dz) = \sum_{j=1}^{N^{\varepsilon}(T)} F(T^{\varepsilon}(j), Z^{\varepsilon}(j)) \mathbb{1}_{\{\mathbb{R}\setminus B(\varepsilon)\}}(Z^{\varepsilon}(j)), \quad \text{where} \quad \begin{cases} N^{\varepsilon}(T) \text{ is a time-inomogeneous Poisson process with intensity } \lambda^{\varepsilon}(t) = \int_{\mathbb{R}\setminus B(\varepsilon)} \nu_{t}(dz), \\ T^{\varepsilon}(j) = \inf\{t \in [0,T], N^{\varepsilon}(t) = j\} \text{ are the jump times of } N^{\varepsilon}, \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) - \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) - \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) - \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) - \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) + \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) + \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) + \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) + \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz) \text{ are the jump sizes of } N^{\varepsilon} dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) + \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j-1),dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) + \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz) + \int_{\mathbb{R}\setminus B(\varepsilon)} zN(T^{\varepsilon}(j),dz \\ Z^{\varepsilon}(j) = \int_{\mathbb{R}\setminus B(\varepsilon)}$$

The jump sizes $Z^{\varepsilon}(j)$ has conditional distribution $\left| \mathbb{P}\left(Z^{\varepsilon}(j) \in B \mid T^{\varepsilon}(j) = t \right) = \frac{\nu_t(B \cap \mathbb{R} \setminus B(\varepsilon))}{\nu_t(\mathbb{R} \setminus B(\varepsilon))} \right|$ and can be simulated by inversion or rejection method for simple cases.

 J_0

B. Gaussian approximation of the small jumps

We approximate the *small jumps* by a Gaussian r.v, making use of the following bound in L^p -Wasserstein distance. The proof rely on a Berry-Essen type bound for the CLT obtained by Bobkov in [Bob18].

Proposition: L^p -Wasserstein approximation bound Assume that F satisfies (H2) for $p \ge 2$ and that for any $s \in [0, T]$, $F(s, \cdot)$ does not vanish on $B(\varepsilon)$ and has at most polynomial growth. Then, there exists a constant \mathcal{A}_p only depending on p such that the following inequality holds for any $r, t \in [0, T]$:

$$\mathcal{W}_p\left(\mathscr{L}aw\left(\int_r^t \int_{B(\varepsilon)} F(s,z)\tilde{N}(ds,dz)\right), \ \mathcal{N}\left(0,\int_r^t \int_{B(\varepsilon)} |F(s,z)|^2 \nu_s(dz)ds\right)\right)^p \leq \mathcal{A}_p \frac{\int_r^t \int_{B(\varepsilon)} |F(s,z)|^{p+2} \nu_s(dz)ds}{\int_r^t \int_{B(\varepsilon)} |F(s,z)|^2 \nu_s(dz)ds} \quad \frac{1}{\varepsilon \to 0}$$

Numerical scheme

Let $t_i = i\frac{T}{n}$. Define the numerical scheme \overline{X} by $\overline{X}_{t_0} = X_0$ and for each $i \in \{1, \ldots, n\}$:

$$\overline{X}_{t_i} = \overline{X}_{t_{i-1}} + a(t_{i-1}, \overline{X}_{t_{i-1}}) \frac{T}{n} + b(t_{i-1}, \overline{X}_{t_{i-1}}) (W_{t_i} - W_{t_{i-1}}) \\ + B(\overline{X}_{t_{i-1}}) + \int_{t_{i-1}}^{t_i} \int_{\mathbb{R} \setminus B(\varepsilon)} c(s, \overline{X}_{t_{i-1}}, z) \tilde{N}(ds, dz),$$

where the Gaussian r.v. $B(\overline{X}_{t_{i-1}})$ is the 2nd component of the L^p -Wasserstein optimal coupling between $\mathscr{L}aw\left(\int_{t_{i-1}}^{t_i}\int_{B(\varepsilon)}c(s,x,z)\tilde{N}(ds,dz)\right)$ and $\mathcal{N}\left(0,\int_{t_{i-1}}^{t_i}\int_{B(\varepsilon)}c(s,x,z)\nu_s(dz)ds\right)$, taken at $x=\overline{X}_{t_{i-1}}$.

Main theorem : Strong error convergence –

Assume (H1), (H2), and that $c(s, x, \cdot)$ does not vanish on $B(\varepsilon)$. Let $\eta(t) = t_i$ for $t \in [t_i, t_i + 1)$. Then:

1. There exists
$$m(p,T) > 0$$
 such that $\sup_{t \in [0,T]} \mathbb{E}[|\overline{X}_{\eta(t)}|^p] \leq m(p,T) < \infty$,

