

Approximations of fractional stochastic differential equations

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Sketch

- 1 Introduction
- 2 Some notions of rough path analysis
- 3 Approximation of the Levy area
- 4 Approximation schemes

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Equation under consideration

$$y_t = a + \int_0^t \sigma(y_s) dx_s + \int_0^t b(y_s) ds, \quad t \in [0, T] \quad (1)$$

- $a \in \mathbb{R}^n$ initial condition
- b, σ smooth coefficients
- $x = B = (B^1, \dots, B^d)$ d -dimensional fractional Brownian motion
- B^i iid fractional Brownian motions, $H > 1/3$

Motivations: Engineering, Finance

Some motivations

- Electrical engineering
Denk, Meintrup y Schäffler (2003):
Modeling, Simulation and Optimization of Integrated Circuits
- Financial time series
Bayraktar, Poor, Sircar (2004):
Estimating the fractal dimension of the S&P 500 Index using wavelet analysis
- Biophysics
Kou (2008):
Stochastic modeling in nanoscale biophysics: subdiffusion within proteins
- Biophysics 2 (with $H \approx 0.34$)
Odde, Tanaka, Hawkins, Buettner (1996):
Stochastic dynamics of the nerve growth cone and its microtubules during neurite outgrowth.

Definition of the fractional Brownian motion

- B^i centered Gaussian process
- Variance of the increments:

$$E[|(\delta B)_{st}|^2] \equiv E[|B_t - B_s|^2] = |t - s|^{2H}$$

- H^- \equiv Hölder-continuity exponent of B
- If $H = 1/2$, $B =$ Brownian motion
- If $H \neq 1/2$, pathwise methods (rough paths) in order to solve (1)
- On rough paths theory:
 - ▶ Lyons-Qian, Coutin-Qian, Lejay, Baudoin, Friz-Victoir
 - ▶ Feyel-de la Pradelle, Gubinelli

Aim of the talk

- 1 Recall how Levy areas enter into rough paths analysis:
Follows essentially Gubinelli's work
- 2 Approximations of fractional SDEs
 - ▶ **Approximation of the Levy area**
(with A. Neuenkirch, J. Unterberger)
 - ▶ **Implementable approximation of fractional SDEs**
(with A. Neuenkirch, A. Deya)
 - ▶ Weak approximations:
 - ★ **Stroock's type** (with X. Bardina, I. Nourdin, C. Rovira)
 - ★ **Donsker's type** (with X. Bardina, C. Rovira)

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Formal computations (1)

Notational simplification: x is γ -Hölder process, $y_t \in \mathbb{R}$ and

$$y_t = a + \int_0^t \sigma(y_s) dx_s$$

Hypothesis:

The solution y_t exists in a space $C^\gamma([0, T])$, with $\gamma < H$

Thus:

$$\begin{aligned}(\delta y)_{st} &\equiv y_t - y_s = \int_s^t \sigma(y_v) dx_v \\ &= \sigma(y_s)(\delta x)_{st} + \int_s^t [\sigma(y_v) - \sigma(y_s)] dx_v \\ &= \zeta_s(\delta x)_{st} + r_{st}\end{aligned}$$

Regularity of the coefficients:

$\zeta = \sigma(y)$: bounded, γ -Hölder,

r : 2γ -Hölder

Formal computations (2)

Let z be such that (controlled process):

$$(\delta z)_{st} = \zeta_s(\delta x)_{st} + r_{st}, \quad \text{with } \zeta \in C^\gamma, r \in C^{2\gamma} \quad (2)$$

Formally:

$$\begin{aligned} \int_s^t z_v dx_v &= z_s(\delta x)_{st} + \int_s^t (\delta z)_{sv} dx_v \\ &= z_s(\delta x)_{st} + \zeta_s \int_s^t (\delta x)_{sv} dx_v + \int_s^t r_{sv} dx_v \\ &= z_s(\delta x)_{st} + \zeta_s \mathbf{x}_{st}^2 + \int_s^t r_{sv} dx_v \end{aligned}$$

Formal computations (3)

If we are allowed to write:

$$\int_s^t z_v dx_v = z_s(\delta x)_{st} + \zeta_s \mathbf{x}_{st}^2 + \int_s^t r_{sv} dx_v$$

Then we have:

- $z_s(\delta x)_{st}$ trivially defined
- $\zeta_s \mathbf{x}_{st}^2$ easily defined, provided a Levy area \mathbf{x}^2 exists
- $\int_s^t r_{sv} dx_v$ defined in a generalized Young sense if $r \in C^{2\gamma}$, $x \in C^\gamma$ and $3\gamma > 1$

Remark

- *These claims can be rigorously proved, and allow to define $\int_s^t z_v dx_v$ for a controlled process z*
- *The SDE can be solved thanks to a fixed point argument, in the class of processes admitting the decomposition (2)*

Levy area: important facts

Hypothesis: $x \in \mathcal{C}_1^\gamma(\mathbb{R}^d)$ allows to define $\mathbf{x}^2 = "$ $\int dx \int dx$ " such that:

- \mathbf{x}^2 is an element of $\mathcal{C}_2^{2\gamma}(\mathbb{R}^{d \times d})$, where

$$\|f\|_\mu = \sup_{s,t \in [a,b]} \frac{|f_{st}|}{|t-s|^\mu}, \quad \text{and} \quad \mathcal{C}_2^\mu = \{f \in \mathcal{C}_2; \|f\|_\mu < \infty\}.$$

- Algebraic property of \mathbf{x}^2 : for $0 \leq s < u < t \leq T$

$$\mathbf{x}_{st}^2(i,j) - \mathbf{x}_{su}^2(i,j) - \mathbf{x}_{ut}^2(i,j) = (\delta x^i)_{su} (\delta x^j)_{ut}$$

The rough paths black box

Hypothesis:

$x \in \mathcal{C}_1^\gamma(\mathbb{R}^d)$ with $\gamma > 1/3$ and Levy area $\mathbf{x}^2 \in \mathcal{C}_2^{2\gamma}(\mathbb{R}^{d \times d})$

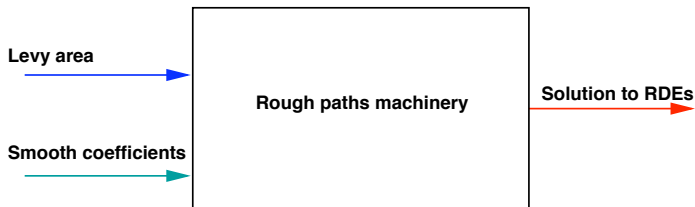
Coefficient $\sigma \in \mathcal{C}_b^3$

Main rough paths theorem: Let y be the solution to $dy = \sigma(y) dx$.

Then

$$F : \mathbb{R}^n \times \mathcal{C}_1^\gamma(\mathbb{R}^d) \times \mathcal{C}_2^{2\gamma}(\mathbb{R}^{d \times d}) \longrightarrow \mathcal{C}_1^\gamma(\mathbb{R}^n), \quad (a, x, \mathbf{x}^2) \mapsto y$$

is a continuous map



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Definition of the fractional Levy area

Let B be a 2-dim fBm with $H > 1/4$

The Stratonovich-Levy area \mathbf{B}^2 is defined by

$$\mathbf{B}_{st}^2(1, 1) = \frac{[B_t^{(1)} - B_s^{(1)}]^2}{2}, \quad \mathbf{B}_{st}^2(2, 2) = \frac{[B_t^{(2)} - B_s^{(2)}]^2}{2}$$

The term $\mathbf{B}_{st}^2(1, 2) = \int_s^t dB_{u_1}^{(1)} \int_s^{u_1} dB_{u_2}^{(2)}$ can be defined either by:

- Malliavin calculus tools
- Regularization or linearization of the fBm path (Coutin-Qian)
- Analytic approximation (Unterberger)
 - ▶ Based on a complex-valued Gaussian process Γ
 - ▶ Γ defined on the complex upper half plane $\bar{\Pi}^+$, analytic on Π^+
 - ▶ The restriction of Γ on \mathbb{R} is a fBm
 - ▶ Allows contour deformations and other complex analysis tools

Approximations of the Levy area

Consider $X_T = \mathbf{B}_{0T}^2(2, 1) = \int_0^T B_s^{(1)} dB_s^{(2)}$.

Natural approximation: partition $\{iT/n; i \leq n\}$, and

$$X_T^n = \sum_{i=0}^{n-1} B_{iT/n}^{(1)} \left(B_{(i+1)T/n}^{(2)} - B_{iT/n}^{(2)} \right)$$

Another natural (trapezoidal) approximation:

$$\widehat{X}_T^n = \frac{1}{2} \sum_{i=0}^{n-1} \left(B_{iT/n}^{(1)} + B_{(i+1)T/n}^{(1)} \right) \left(B_{(i+1)T/n}^{(2)} - B_{iT/n}^{(2)} \right)$$

L^2 approximation

Theorem

$\mathbf{E}|X_T - X_T^n|^2$ can be expanded as:

$$\begin{cases} \alpha_1(H) \cdot T^{4H} \cdot n^{-4H+1} + o(n^{-4H+1}) & \text{for } H \in (1/4, 3/4), \\ \frac{9}{128} \cdot T^{4H} \cdot \log(n) n^{-2} + o(\log(n) n^{-2}) & \text{for } H = 3/4, \\ \alpha_2(H) \cdot T^{4H} \cdot n^{-2} + o(n^{-2}) & \text{for } H \in (3/4, 1). \end{cases}$$

$\mathbf{E}|X_T - \widehat{X}_T^n|^2$ can be expanded, for $H > 1/4$, as:

$$\mathbf{E}|X_T - \widehat{X}_T^n|^2 = \alpha_3(H) \cdot T^{4H} \cdot n^{-4H+1} + o(n^{-4H+1}),$$

Sketch of the proof (1)

Set $\mathcal{A}_{st} = \mathbf{B}_{st}^2(2, 1)$.

The error for the Euler scheme can be decomposed as:

$$X_T - X_T^n = \sum_{i=1}^n J_i^n$$

where the random variables J_i^n are defined by

$$J_i^n = \int_{iT/n}^{(i+1)T/n} (B_s^{(1)} - B_{i/n}^{(1)}) dB_s^{(2)} = \mathcal{A}_{(iT/n), (i+1)T/n}$$

Hence

$$\mathbf{E}[|X_T - X_T^n|^2] = \sum_{i,j} \mathbf{E}[J_i^n J_j^n] = \frac{T^{4H}}{n^{4H}} \sum_{i,j} \mathbf{E}[\mathcal{A}_{i,i+1} \mathcal{A}_{j,j+1}]$$

Sketch of the proof (2)

Example of computation:

$$\mathbf{E}[\mathcal{A}_{01} \mathcal{A}_{12}] \asymp \int_1^2 \int_0^1 |s_1 - s_2|^{2H-2} (s_1^{2H} - 1 - |s_1 - s_2|^{2H} + |s_2 - 1|^{2H}) ds_2 ds_1,$$

which can be computed explicitly

Differences in the rates of convergence:

Due to the influence of the off diagonal terms, which grows with H
More specifically, one has to compare

$$\sum_{|i-j|>1, 1 \leq i, j \leq n} |i-j|^{4H-4} \quad \text{and} \quad n$$

Central and non central limit theorems

Theorem

Let Z be a standard Gaussian random variable.

- ① Case $1/4 < H < 3/4$: the following central limit theorems hold:

$$\lim_{n \rightarrow \infty} n^{2H-1/2} (X_T - X_T^n) \stackrel{(d)}{=} \sqrt{\alpha_1(H)} T^{2H} \cdot Z$$

- ② $\lim_{n \rightarrow \infty} n(\log(n))^{-1/2} (X_T - X_T^n) \stackrel{(d)}{=} \frac{3}{4\sqrt{8}} T^{2H} \cdot Z$ for $H = 3/4$.

- ③ Case $H > 3/4$: let R_1 and R_2 be two independent Rosenblatt processes. Then it holds

$$\lim_{n \rightarrow \infty} n (X_T - X_T^n) \stackrel{(d)}{=} \sqrt{2\alpha_4(H)} T^{2H} \cdot (R_1 - R_2).$$

Sketch of the proof

Reduction of the problem: Random variables in the second chaos of B
 \hookrightarrow one can use Nualart-Peccati's criterion

Criterion: Let $\{Z_n; n \geq 1\}$ in the p^{th} chaos of B .

Assume that $\lim_{n \rightarrow \infty} \mathbf{E}[Z_n^2] = 1$.

Then $\mathcal{L} - \lim_{n \rightarrow \infty} Z_n = \mathcal{N}(0, 1)$ if and only if $\lim_{n \rightarrow \infty} \mathbf{E}[Z_n^4] = 3$

Computation of the 4th moments: Complex analysis tools and graphical representations

Case $H > 3/4$:

Study of the difference between Euler and trapezoidal scheme

\hookrightarrow Products of increments $(B_{(i+1)/n}^{(1)} - B_{i/n}^{(1)})(B_{(i+1)/n}^{(2)} - B_{i/n}^{(2)})$ appear

Yields the Rosenblatt type behavior

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Euler scheme

Driving signal: m -dimensional fBm B , with $1/3 < H < 1/2$

Equation: $dY_t = \sum_{i=1}^m \sigma^{(i)}(Y_t) dB_t^{(i)}, \quad Y_0 \in \mathbb{R}^d$

Grid: $t_k = \frac{kT}{n}$

Euler scheme: $Y_{t_{k+1}}^n = Y_{t_k}^n + \sum_{i=1}^m \sigma^{(i)}(Y_{t_k}^n)(B_{t_{k+1}}^{(i)} - B_{t_k}^{(i)})$

Problem: when $1/3 < H < 1/2$, the Euler scheme is divergent!

A 1-d example: $dY_t = Y_t dB_t, \quad Y_0 = 1$

Exact solution: $Y_t = \exp(B_t)$

Approximation for the example

Euler approximation $t = 1$: $Y_1^n = \prod_{k=0}^{n-1} (1 + (B_{(k+1)/n} - B_{k/n}))$

Asymptotics of the Euler approximation: for $n \in \mathbb{N}$ sufficiently large

$$\begin{aligned} Y_1^n &= \exp \left(\sum_{k=0}^{n-1} \log(1 + (B_{(k+1)/n} - B_{k/n})) \right) \\ &= \exp \left(B_1 - \frac{1}{2} \sum_{k=0}^{n-1} |B_{(k+1)/n} - B_{k/n}|^2 + \rho_n \right), \end{aligned}$$

where $\rho_n \xrightarrow{\text{Prob.}} 0$ for $n \rightarrow \infty$ for $H > 1/3$.

But for $H < 1/2$: $\sum_{k=0}^{n-1} |B_{(k+1)/n} - B_{k/n}|^2 \xrightarrow{\text{a.s.}} \infty$
 $\implies Y_1^n \xrightarrow{\text{Prob.}} 0$

Milstein-Davie scheme

Equation: $dY_t = \sum_{i=1}^m \sigma^{(i)}(Y_t) dB_t^{(i)}$, $Y_0 \in \mathbb{R}^d$

Milstein-Davie scheme: set $\mathcal{D}^{(i)} = \sum_{l=1}^d \sigma_l^{(i)} \partial_{x_l}$ and

$$\tilde{Y}_{t_{k+1}}^n = \tilde{Y}_{t_k}^n + \sum_{i=1}^m \sigma^{(i)}(\tilde{Y}_{t_k}^n) (B_{t_{k+1}}^{(i)} - B_{t_k}^{(i)}) + \frac{1}{2} \sum_{i,j=1}^m \mathcal{D}^{(i)} \sigma^{(j)}(\tilde{Y}_{t_k}^n) \mathbf{B}_{t_k t_{k+1}}^2(i, j)$$

Theorem (Davie; Friz-Victoir): $|\tilde{Y}_T^n - Y_T| \rightarrow 0$, with an a.s. rate of convergence $n^{-(3H-1-\varepsilon)}$

Problem: $\mathbf{B}_{t_k t_{k+1}}^2$ cannot be simulated exactly
 \hookrightarrow Davie's scheme cannot be implemented directly

An implementable scheme

Approximation:

$$\begin{aligned}\widehat{Y}_{t_{k+1}}^n &= \widehat{Y}_{t_k}^n + \sum_{i=1}^m \sigma^{(i)}(\widehat{Y}_{t_k}^n) \left(B_{t_{k+1}}^{(i)} - B_{t_k}^{(i)} \right) \\ &\quad + \frac{1}{2} \sum_{i,j=1}^m \mathcal{D}^{(i)} \sigma^{(j)}(\widehat{Y}_{t_k}^n) \left(B_{t_{k+1}}^{(i)} - B_{t_k}^{(i)} \right) \left(B_{t_{k+1}}^{(j)} - B_{t_k}^{(j)} \right)\end{aligned}$$

Theorem

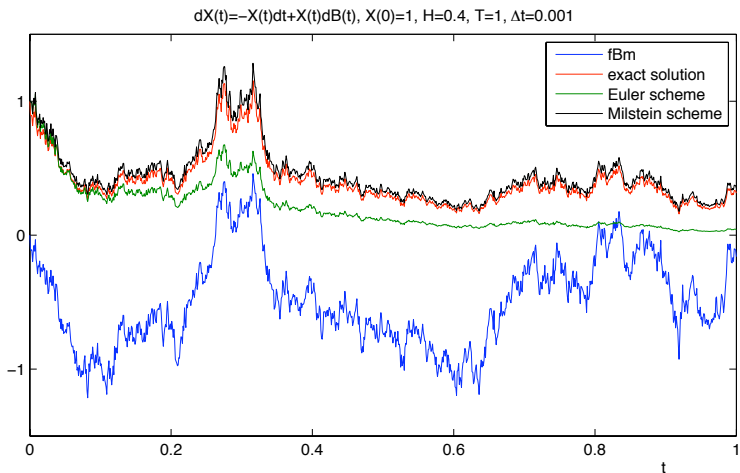
Let $\gamma \in (1/4, H)$ and $\varepsilon \in (0, H - \gamma)$.

Then, there exists a finite r.v. $\eta_{\gamma, \varepsilon, T}$ such that

$$\|Y - \widehat{Y}^n\|_{\gamma} \leq \eta_{\gamma, \varepsilon, T} \cdot n^{-(H-\gamma-\varepsilon)}.$$

Convergence illustration

Equation: $dY_t = -Y_t dt + Y_t dB_t$, hence $Y_t = e^{-t+B_t}$



Sketch of the proof

Observation:

\bar{B}^n := linear approximation of B based on the **same** grid kT/n

\bar{Y} := the solution to $d\bar{Y}_t = \sum_{i=1}^m \sigma^{(i)}(\bar{Y}_t) d\bar{B}_t^{(i),n}$

Then \hat{Y}^n = Davie-Milstein scheme of \bar{Y}

Splitting the proof:

(1) Convergence of \bar{Y} towards Y , which amounts to the convergence of $\bar{\mathbf{B}}^{2,n}$ towards \mathbf{B}^2

↔ elaboration of Coutin-Qian, and of Section 3

(2) Convergence of \hat{Y}^n towards \bar{Y}

↔ elaboration of Davie's and Friz-Victoir results

Ongoing work: Analysis of the error process

↔ a.s optimal rate of convergence $n^{-(2H-1/2)}$?

Open problem: moment estimates (and linear equations)!