

Rough paths methods 1: Introduction

Samy Tindel

Purdue University

University of Aarhus 2016

Outline

- 1 Motivations for rough paths techniques
- 2 Summary of rough paths theory

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Equation under consideration

Equation:

Standard differential equation driven by fBm, \mathbb{R}^n -valued

$$Y_t = a + \int_0^t V_0(Y_s) ds + \sum_{j=1}^d \int_0^t V_j(Y_s) dB_s^j, \quad (1)$$

with

- $t \in [0, 1]$.
- Vector fields V_0, \dots, V_d in C_b^∞ .
- A d -dimensional fBm B with $1/3 < H < 1$.
- Note: some results will be extended to $H > 1/4$.

Fractional Brownian motion

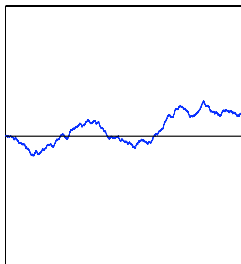
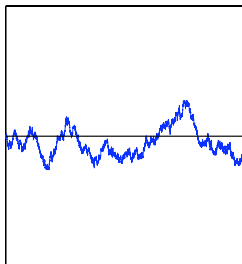
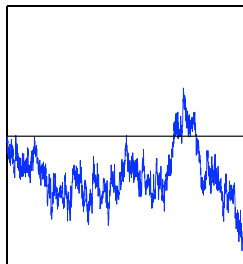
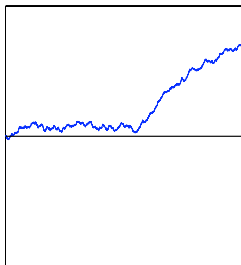
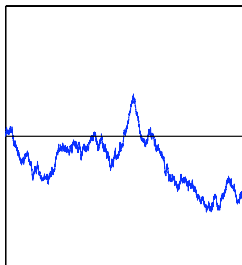
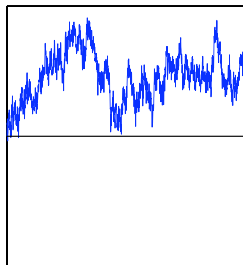
- $B = (B^1, \dots, B^d)$
- B^j centered Gaussian process, independence of coordinates
- Variance of the increments:

$$\mathbf{E}[|B_t^j - B_s^j|^2] = |t - s|^{2H}$$

- $H^- \equiv$ Hölder-continuity exponent of B
- If $H = 1/2$, $B =$ Brownian motion
- If $H \neq 1/2$ natural generalization of BM

Remark: FBm widely used in applications

Examples of fBm paths



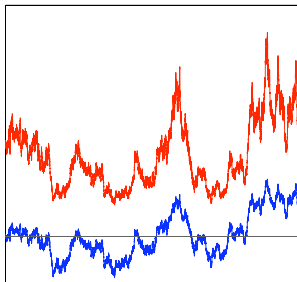
$H = 0.3$

$H = 0.5$

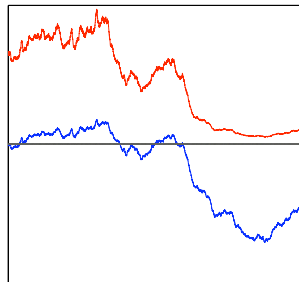
$H = 0.7$

Paths for a linear SDE driven by fBm

$$dY_t = -0.5Y_t dt + 2Y_t dB_t, \quad Y_0 = 1$$



$H = 0.5$



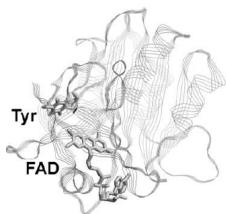
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Blue: $(B_t)_{t \in [0,1]}$ Red: $(Y_t)_{t \in [0,1]}$

Some applications of fBm driven systems

Biophysics, fluctuations of a protein:

- New experiments at molecule scale
↪ Anomalous fluctuations recorded
- Model: Volterra equation driven by fBm
↪ Samuel Kou
- Statistical estimation needed



Finance:

- Stochastic volatility driven by fBm (Sun et al. 2008)
- Captures long range dependences between transactions

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Rough paths assumptions

Context: Consider a Hölder path x and

- For $n \geq 1$, $x^n \equiv$ linearization of x with mesh $1/n$
 $\hookrightarrow x^n$ piecewise linear.
- For $0 \leq s < t \leq 1$, set

$$\mathbf{x}_{st}^{2,n,i,j} \equiv \int_{s < u < v < t} dx_u^{n,i} dx_v^{n,j}$$

Rough paths assumption 1:

- x is a \mathcal{C}^γ function with $\gamma > 1/3$.
- The process $\mathbf{x}^{2,n}$ converges to a process \mathbf{x}^2 as $n \rightarrow \infty$
 \hookrightarrow in a $\mathcal{C}^{2\gamma}$ space.

Rough paths assumption 2:

- Vector fields V_0, \dots, V_j in \mathcal{C}_b^∞ .

Brief summary of rough paths theory

Main rough paths theorem (Lyons): Under previous assumptions

↪ Consider y^n solution to equation

$$y_t^n = a + \int_0^t V_0(y_u^n) du + \sum_{j=1}^d \int_0^t V_j(y_u^n) dx_u^{n,j}.$$

Then

- y^n converges to a function Y in \mathcal{C}^γ .
- Y can be seen as solution to

$$\Leftrightarrow Y_t = a + \int_0^t V_0(Y_u) du + \sum_{j=1}^d \int_0^t V_j(Y_u) dx_u^j.$$

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Rough paths theory

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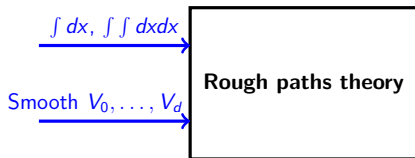
↪ Consider y^n solution to equation

$$y_t^n = a + \int_0^t V_0(y_u^n) du + \sum_{j=1}^d \int_0^t V_j(y_u^n) dx_u^{n,j}.$$

Then

- y^n converges to a function Y in \mathcal{C}^γ .
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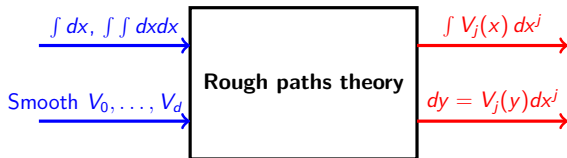
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Then

- y^n converges to a function Y in \mathcal{C}^γ .
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$$\hookrightarrow Y_t = a + \int_0^t V_0(Y_u) du + \sum_{j=1}^d \int_0^t V_j(Y_u) dx_u^j.$$



Iterated integrals and fBm

Nice situation: $H > 1/4$

↪ 2 possible constructions for **geometric** iterated integrals of B .

- Malliavin calculus tools
↪ Ferreiro-Utzet
- Regularization or linearization of the fBm path
↪ Coutin-Qian, Friz-Gess-Gulisashvili-Riedel

Conclusion: for $H > 1/4$, one can solve equation

$$dY_t = V_0(Y_t) dt + V_j(Y_t) dB_t^j,$$

in the rough paths sense.

Remark: Extensions to $H \leq 1/4$ (Unterberger, Nualart-T).

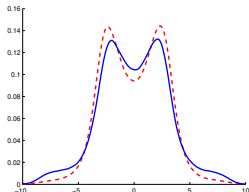
Study of equations driven by fBm

Basic properties:

- 1 Moments of the solution
- 2 Continuity w.r.t initial condition, noise

More advanced natural problems:

- 1 Density estimates
↪ Hu-Nualart + Lots of people
- 2 Numerical schemes
↪ Neuenkirch-T, Friz-Riedel
- 3 Invariant measures, ergodicity
↪ Hairer-Pillai, Deya-Panloup-T
- 4 Statistical estimation (H , coeff. V_j)
↪ Berzin-León, Hu-Nualart, Neuenkirch-T



Extensions of the rough paths formalism

Stochastic PDEs:

- Equation: $\partial_t Y_t(\xi) = \Delta Y_t(\xi) + \sigma(Y_t(\xi)) \dot{x}_t(\xi)$
- $(t, \xi) \in [0, 1] \times \mathbb{R}^d$
- Easiest case: x finite-dimensional noise
- Methods:
 - ↪ viscosity solutions or adaptation of rough paths methods

KPZ equation:

- Equation: $\partial_t Y_t(\xi) = \Delta Y_t(\xi) + (\partial_\xi Y_t(\xi))^2 + \dot{x}_t(\xi) - \infty$
- $(t, \xi) \in [0, 1] \times \mathbb{R}$
- $\dot{x} \equiv$ space-time white noise
- Methods:
 - ▶ Extension of rough paths to define $(\partial_x Y_t(\xi))^2$
 - ▶ Renormalization techniques to remove ∞

Aim

- 1 Definition and properties of fractional Brownian motion
- 2 Some estimates for Young's integral, case $H > 1/2$
- 3 Extension to $1/3 < H \leq 1/2$

General strategy

- 1 In order to solve our equation, we shall go through the following steps:
 - ▶ Young integral for $H > 1/2$
 - ▶ Case $1/3 < H < 1/2$, with a semi-pathwise method
- 2 For each case, 2 main steps:
 - ▶ Definition of a stochastic integral $\int u_s dB_s$ for a reasonable class of processes u
 - ▶ Resolution of the equation by means of a fixed point method