# A rough-paths type approach to non-commutative stochastic integration

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 $\longrightarrow$  To show the flexibility of the rough-paths machinery.

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→ To provide another perspective on non-commutative stochastic calculus (Reference : Biane-Speicher (PTRF 98')) (Reference : Bozejko-Kümmerer-Speicher (Com. Math. Phys. 97')).

Basics on classical rough paths theory

## Outline



Basics on classical rough paths theory  $\gamma > 1/2$  ('Young case')  $\frac{1}{2} < \gamma \leq \frac{1}{2}$ 

2 Non-commutative probability theory and rough paths

Non-commutative processes Integration The free Bm case

## Outline

### Basics on classical rough paths theory $\gamma > 1/2$ ('Young case') $\frac{1}{2} < \gamma \leq \frac{1}{2}$

Non-commutative processes

## Classical rough paths theory [Lyons (98)]

Consider a non-differentiable path

$$x:[0,T]\to\mathbb{R}^n,$$

with Hölder regularity  $\gamma \in (0,1)$  (i.e.,  $||x_t - x_s|| \le c |t - s|^{\gamma}$ ).

**Question** : Given a smooth  $f : \mathbb{R}^n \to \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n)$ , how can we define

 $\int f(x_t) \, dx_t \quad ?$ 

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 $\int f(x_t) \, dx_t \quad ?$ 

We would like this definition to be sufficiently extensible to cover differential equations

$$dy_t = f(y_t) \, dx_t.$$

### **Rough paths theory** $\implies$ Give a sense to

$$\int f(x_t) dx_t$$

when  $x : [0, T] \to \mathbb{R}^n$  is a  $\gamma$ -Hölder path with  $\gamma \in (0, 1)$ .

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**Application :** Pathwise approach to stochastic calculus.

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**Application :** Pathwise approach to stochastic calculus.

• The construction of the integral depends on  $\gamma$  :

$$\gamma > \frac{1}{2}$$
 ,  $\gamma \in (\frac{1}{3}, \frac{1}{2}]$  ,  $\gamma \in (\frac{1}{4}, \frac{1}{3}]$  , ...

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• When  $\gamma \leq \frac{1}{2}$ , additional assumptions on x.

## Outline

## Basics on classical rough paths theory $\gamma > 1/2$ ('Young case')

Non-commutative processes

$$\gamma > 1/2$$
 ('Young case')

### Theorem (Young (1936))

If x is  $\gamma$ -Hölder and y is  $\kappa$ -Hölder with  $\gamma + \kappa > 1$ , then the Riemann sum  $\sum_{t_i \in \mathcal{P}_{[s,t]}} y_{t_i}(x_{t_{i+1}} - x_{t_i})$  converges as the mesh of the partition  $\mathcal{P}_{[s,t]}$  tends to 0. We define

$$\int_{s}^{t} y_{u} \, dx_{u} := \lim_{|\mathcal{P}_{[s,t]}| \to 0} \sum_{t_{i} \in \mathcal{P}_{[s,t]}} y_{t_{i}}(x_{t_{i+1}} - x_{t_{i}}).$$

Rough-paths and non-commutative probability

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**Application** :  $x : [0, T] \to \mathbb{R}^n \gamma$ -Hölder with  $\gamma > 1/2$ .  $f \text{ smooth} \Rightarrow (t \mapsto f(x_t)) \gamma$ -Hölder  $\Rightarrow \int f(x_t) dx_t$  Young integral.

Non-commutative probability theory and rough paths

## $\gamma>1/2$ ('Young case')

Another way to see Young's result :

$$\int_{s}^{t} f(x_{u}) dx_{u} = f(x_{s}) (x_{t} - x_{s}) + \int_{s}^{t} [f(x_{u}) - f(x_{s})] dx_{u}.$$

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Main term :

$$f(x_s)(x_t-x_s), \quad ext{with } |f(x_s)(x_t-x_s)| \leq c |t-s|^\gamma$$

Residual term :

$$\int_s^t [f(x_u) - f(x_s)] dx_u \quad \text{with } |\int_s^t [f(x_u) - f(x_s)] dx_u| \le c |t - s|^{2\gamma}$$

As  $2\gamma > 1$ , 'disappears in Riemann sum'.

## $\gamma > 1/2$ ('Young case')

$$\int_{s}^{t} f(x_{u}) dx_{u} = \sum_{t_{i} \in \mathcal{P}_{[s,t]}} \int_{t_{i}}^{t_{i+1}} f(x_{u}) dx_{u}$$
  
= 
$$\sum_{t_{i} \in \mathcal{P}_{[s,t]}} f(x_{t_{i}}) (x_{t_{i+1}} - x_{t_{i}}) + \sum_{t_{i} \in \mathcal{P}_{[s,t]}} \int_{t_{i}}^{t_{i+1}} [f(x_{u}) - f(x_{t_{i}})] dx_{u}.$$

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### 2 Non-commutative probability theory and rough paths

Non-commutative processes Integration The free Bm case If  $x\in \mathcal{C}^\gamma$  with  $\gamma\leq 1/2,$  we cannot guarantee the convergence of the sum

$$\sum_{i\in\mathcal{P}_{[s,t]}}f(x_{t_i})(x_{t_{i+1}}-x_{t_i}).$$

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 $\implies$  We try to 'correct' the Riemann sum :

$$\sum_{t_i \in \mathcal{P}_{[s,t]}} \{f(x_{t_i}) (x_{t_{i+1}} - x_{t_i}) + C_{t_i,t_{i+1}}\}.$$

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To find out a proper C, a few heuristic considerations. Suppose that we can define  $\int_{s}^{t} f(x_u) dx_u$ . Then...

$$\int_{s}^{t} f(x_{u}) dx_{u}$$
  
=  $f(x_{s}) (x_{t} - x_{s}) + \int_{s}^{t} [f(x_{u}) - f(x_{s})] dx_{u}$ 

$$\int_{s}^{t} f(x_{u}) dx_{u}$$

$$= f(x_{s}) (x_{t} - x_{s}) + \int_{s}^{t} [f(x_{u}) - f(x_{s})] dx_{u}$$

$$= f(x_{s}) (x_{t} - x_{s}) + \int_{s}^{t} \nabla f(x_{s}) (x_{u} - x_{s}) dx_{u} + \int_{s}^{t} y_{s,u} dx_{u},$$
where  $|y_{s,u}| = |f(x_{u}) - f(x_{s}) - \nabla f(x_{s}) (x_{u} - x_{s})| \le c |u - s|^{2\gamma}.$ 

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 $|\int_{s}^{t} y_{s,u} dx_{u}| \leq c |t-s|^{3\gamma}$ . As  $3\gamma > 1$ , 'disappears in Riemann sum'.

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 $|\int_s^t y_{s,u} \, dx_u| \le c \ |t-s|^{3\gamma}$ . As  $3\gamma > 1$ , 'disappears in Riemann sum'.

 $\Rightarrow$  Main term :

$$f(x_s)(x_t-x_s)+\int_s^t \nabla f(x_s)(x_u-x_s)\,dx_u.$$

$$\int_{s}^{t} f(x_{u}) dx_{u} " := " \lim \sum_{t_{k}} \left\{ f(x_{t_{k}}) \left( x_{t_{k+1}} - x_{t_{k}} \right) + C_{t_{k}, t_{k+1}} \right\}$$

where

$$C_{s,t} := \int_s^t \nabla f(x_s)(x_u - x_s) \, dx_u.$$

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$$C_{s,t} := \partial_i f_j(x_s) \int_s^t (x_u^{(i)} - x_s^{(i)}) dx_u^{(j)}.$$

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 $\implies$  to define the integral when  $\gamma \in (\frac{1}{3}, \frac{1}{2}]$ , we need to assume the a priori existence of the Lévy area

$$\mathbf{x}_{s,t}^{2,(i,j)} := \int_{s}^{t} \int_{s}^{u} dx_{v}^{(i)} dx_{u}^{(j)}$$

above x. Then everything works...

Non-commutative probability theory and rough paths

Assume that we can define the Lévy area 
$$x^2 = \iint dx dx$$
. Then :

Rough-paths and non-commutative probability

Assume that we can define the Lévy area 
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• We can indeed **define the integral**  $\int f(x_u) dx_u$  as

$$\int f(x_u) \, dx_u := \lim \sum_{t_k} \Big\{ f(x_{t_k}) \, (x_{t_{k+1}} - x_{t_k}) + \nabla f(x_{t_k}) \cdot \mathbf{x}_{t_k, t_{k+1}}^2 \Big\}.$$

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• We can extend this definition to  $\int f(y) dx$  for a large class of paths y, and then solve the equation

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• We can extend this definition to  $\int f(y) dx$  for a large class of paths y, and then solve the equation

$$dy_t = f(y_t) \, dx_t.$$

• We can show that the solution y is a continuous function of the pair  $(x, \mathbf{x}^2)$ , i.e.,  $y = \Phi(x, \mathbf{x}^2)$  with  $\Phi$  continuous w.r.t Hölder topology.

The procedure can be extended to any Hölder coefficient  $\gamma \in (0, \frac{1}{3}]$  provided one can define the iterated integrals of x :

$$\int dx, \quad \iint dx dx, \quad \iiint dx dx dx, \dots$$

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It applies to stochastic processes in a pathwise way.

$$Y(\omega) = \Phi(B(\omega), \mathbf{B}^{2}(\omega)),$$

where  $\Phi$  is continuous and deterministic.

Rough-paths and non-commutative probability

Basics on classical rough paths theory

(Non-commutative probability theory and rough paths)

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#### 2 Non-commutative probability theory and rough paths

Non-commutative processes Integration The free Bm case

 $\mathbf{Question}:$  In the rough-paths machinery, what happens if we replace the  $\gamma\textsc{-H\"older}$  path

$$x:[0,T]\to\mathbb{R}^n$$

with a  $\gamma$ -Hölder path

$$X: [0, T] \rightarrow \mathcal{A},$$

where A is a non-commutative probability space?

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#### Non-commutative processes

Integration The free Bm case

## Motivation : large random matrices [Voiculescu]

Consider the sequence of random matrix-valued processes

$$M_t^n := \left(B_t^{n,(i,j)}\right)_{1 \le i,j \le n},$$

where the  $(B^{n,(i,j)})_{n\geq 1,1\leq i,j\leq n}$  are independent standard Bm.

Denote  $S_t^n := \frac{1}{\sqrt{2}} (M_t^n + (M_t^n)^*).$ 

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Denote  $S_t^n := \frac{1}{\sqrt{2}} (M_t^n + (M_t^n)^*)$ . Then, almost surely, for all  $t_1, \ldots, t_k$ ,

$$\frac{1}{n} \operatorname{Tr} \left( S_{t_1}^n \cdots S_{t_k}^n \right) \xrightarrow{n \to \infty} \varphi \left( X_{t_1} \cdots X_{t_k} \right),$$

for a certain path  $X : \mathbb{R}_+ \to \mathcal{A}$ , where  $(\mathcal{A}, \varphi)$  is a particular noncommutative probability space. This process is called the free Brownian motion. **Definition** : A non-commutative probability space  $\mathcal{A}$  is an algebra of bounded operators (acting on some Hilbert space) endowed with a trace  $\varphi$ , i.e., a linear functional  $\varphi : \mathcal{A} \to \mathbb{C}$  such that

$$arphi(1)=1 \quad, \quad arphi(XY)=arphi(YX) \quad, \quad arphi(XX^*)\geq 0.$$

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"To retain for the sequel" : If  $X, Y \in A$ , then XY may be different from YX.

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"To retain for the sequel" : If  $X, Y \in A$ , then XY may be different from YX.

It is the natural framework to study the asymptotic behaviour of large random matrices with size tending to infinity.

Basics on classical rough paths theory

(Non-commutative probability theory and rough paths)

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Non-commutative processes

#### Integration

The free Bm case

 $\mathbf{Question}:$  In the rough-paths machinery, what happens if we replace the  $\gamma\textsc{-H\"older}$  path

$$x: [0, T] \rightarrow \mathbb{R}^n \quad (\|x_t - x_s\| \le |t - s|^{\gamma})$$

with a  $\gamma$ -Hölder path

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For instance, how can we define

$$\int P(X_t) \cdot dX_t$$

when P is a polynomial and  $\cdot$  refers to the product in  $\mathcal{A}$ ?

$$\gamma > rac{1}{2}$$
 : Young theorem  $\Longrightarrow$   
 $\int P(X_t) \cdot dX_t := \lim \sum_{t_i} P(X_{t_i}) \cdot (X_{t_{i+1}} - X_{t_i}).$ 

Rough-paths and non-commutative probability

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$$\gamma \in \left(\frac{1}{3}, \frac{1}{2}\right]$$
 : corrected Riemann sums  
$$\int P(X_t) \cdot dX_t := \lim \sum_{t_i} \left\{ P(X_{t_i}) \cdot (X_{t_{i+1}} - X_{t_i}) + C_{t_i, t_{i+1}} \right\}.$$

Let us find out C in this context...

As in the finite-dimensional case, one has (morally)

$$\begin{split} \int_{s}^{t} P(X_{u}) \cdot dX_{u} \\ &= P(X_{s}) \cdot (X_{t} - X_{s}) + \int_{s}^{t} \nabla P(X_{s})(X_{u} - X_{s}) \cdot dX_{u} \\ &+ \int_{s}^{t} Y_{s,u} \cdot dX_{u}, \end{split}$$
with  $\|\int_{s}^{t} Y_{s,u} \cdot dX_{u}\| \leq c |t - s|^{3\gamma}.$ 

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with  $\|\int_{s}^{t} Y_{s,u} \cdot dX_{u}\| \leq c |t - s|^{3\gamma}.$ 

Since  $3\gamma > 1$ , main term :

$$P(X_s) \cdot (X_t - X_s) + \int_s^t \nabla P(X_s)(X_u - X_s) \cdot dX_u.$$

#### A natural definition :

$$\int P(X_t) \cdot dX_t \ " := "$$
$$\lim \sum_{t_k} \Big\{ P(X_{t_k}) \cdot (X_{t_{k+1}} - X_{t_k}) + \int_{t_k}^{t_{k+1}} \nabla P(X_{t_k}) (X_u - X_{t_k}) \cdot dX_u \Big\}.$$

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But : Remember that in finite dimension,

$$\int_{s}^{t} \nabla f(x_s)(x_u - x_s) \cdot dx_u = \partial_i f_j(x_s) \int_{s}^{t} \int_{s}^{u} dx_v^{(i)} dx_u^{(j)}.$$

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$$\lim \sum_{t_k} \Big\{ P(X_{t_k}) \cdot (X_{t_{k+1}} - X_{t_k}) + \int_{t_k}^{t_{k+1}} \nabla P(X_{t_k}) (X_u - X_{t_k}) \cdot dX_u \Big\}.$$

But : Remember that in finite dimension,

$$\int_{s}^{t} \nabla f(x_{s})(x_{u}-x_{s}) \cdot dx_{u} = \partial_{i}f_{j}(x_{s}) \int_{s}^{t} \int_{s}^{u} dx_{v}^{(i)} dx_{u}^{(j)}$$

No longer possible for  $\int_{s}^{t} \nabla P(X_{s})(X_{u} - X_{s}) \cdot dX_{u} \dots$  $\implies$  What can play the role of the Lévy area?

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$$\int_{s}^{t} \nabla P(X_{s})(X_{u} - X_{s}) \cdot dX_{u}$$
  
=  $X_{s} \cdot \int_{s}^{t} (X_{u} - X_{s}) \cdot dX_{u} + \int_{s}^{t} (X_{u} - X_{s}) \cdot X_{s} \cdot dX_{u}$ 

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=  $X_{s} \cdot \int_{s}^{t} \int_{s}^{u} dX_{v} \cdot dX_{u} + \int_{s}^{t} \int_{s}^{u} dX_{v} \cdot X_{s} \cdot dX_{u}.$ 

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 $\implies$  For s < t, we 'define' the Lévy area  $X_{s,t}^2$  as the operator on  $\mathcal{A}$ 

$$\mathbf{X}_{s,t}^{2}[Y] = \int_{s}^{t} \int_{s}^{u} dX_{v} \cdot Y \cdot dX_{u} \quad \text{for } Y \in \mathcal{A}.$$

$$\int_{s}^{t} \nabla P(X_{s})(X_{u} - X_{s}) \cdot dX_{u}$$

$$= X_{s} \cdot \int_{s}^{t} (X_{u} - X_{s}) \cdot dX_{u} + \int_{s}^{t} (X_{u} - X_{s}) \cdot X_{s} \cdot dX_{u}$$

$$= X_{s} \cdot \mathbf{X}_{s,t}^{2}[1] + \mathbf{X}_{s,t}^{2}[X_{s}].$$

 $\implies$  For s < t, we 'define' the Lévy area  $\mathsf{X}^2_{s,t}$  as the operator on  $\mathcal A$ 

$$\mathbf{X}_{s,t}^{2}[Y] = \int_{s}^{t} \int_{s}^{u} dX_{v} \cdot Y \cdot dX_{u} \text{ for } Y \in \mathcal{A}.$$

### Assume that we can define the Lévy area $X^2$ in the previous sense.

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Then, for every polynomial P, we can indeed define the integral

$$\int P(X_t) \cdot dX_t$$

as the limit of corrected Riemann sums involving X and  $X^2$ .

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#### With this definition in hand, we can solve the equation

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 $\Rightarrow$  approximation results.

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 $\Rightarrow$  approximation results.

Application : X free Bm  $(||X_t - X_s||_{\mathcal{A}} \le c |t - s|^{1/2})$ 

Rough-paths and non-commutative probability

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Basics on classical rough paths theory

(Non-commutative probability theory and rough paths)

## Outline

# 1 Basics on classical rough paths theory $\gamma > 1/2$ ('Young case') $\frac{1}{3} < \gamma \leq \frac{1}{2}$

#### 2 Non-commutative probability theory and rough paths

Non-commutative processes Integration The free Bm case

$$\mathbf{X}_{s,t}^{\mathbf{2}}[Y] = \int_{s}^{t} \int_{s}^{u} dX_{v} \cdot Y \cdot dX_{u} \quad \text{for } Y \in \mathcal{A}.$$
 (1)

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(1)

We can use the results of Biane-Speicher (PTRF 98') to define (1) as an 'ltô' integral, denoted by  $X^{2,lt\delta} \implies$  the 'rough-paths' calculus based on  $X^{2,lt\delta}$  then coincides with ltô's stochastic calculus w.r.t X.

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Another way to define (1) : consider the linear interpolation  $X^n$  of X along  $t_i^n = \frac{i}{n}$  (i = 1, ..., n), and

$$\mathbf{X}^{2,n}_{s,t}[Y] = \int_s^t \int_s^u dX_v^n \cdot Y \cdot dX_u^n \quad \text{(Lebesgue integral)}.$$

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Then  $X^n \to X$  and  $\mathbf{X}^{2,n} \to \mathbf{X}^{2,\mathsf{Strato}}$ .

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Then  $X^n \to X$  and  $\mathbf{X}^{2,n} \to \mathbf{X}^{2,\text{Strato}}$ . Moreover,

$$\mathbf{X}_{s,t}^{\mathbf{2},\mathsf{Strato}}[Y] = \mathbf{X}_{s,t}^{\mathbf{2},\mathsf{lt\hat{o}}}[Y] + rac{1}{2}\varphi(Y)(t-s).$$

## **Open questions**

- Lévy area for other non-commutative  $\gamma$ -Hölder processes (with  $\gamma \in (\frac{1}{3}, \frac{1}{2}])$  : q-Brownian motion,  $q \in (-1, 1)$ .
- extension to non-martingale processes : *q*-Gaussian processes ([Bożejko-Kümmerer-Speicher]), ...
- smaller  $\gamma$ .