

# Background

## (1) Motivation:

$$\text{Consider } \lambda Y_t = f(Y_t) \lambda X_t. \quad (*)$$

where  $X_t: [0, T] \rightarrow V$ ,  $Y_t: [0, T] \rightarrow U$ , and  
 $f: U \rightarrow C(V, U)$ . next take  $V = U = \mathbb{R}^n$ .

The control question is:

How to interpret (\*) and define the integral

$$Y_t = Y_0 + \int_0^t f(Y_s) \lambda X_s ?$$

① One suggests simply take  $\langle X^n \rangle$ ,  $\langle Y^n \rangle$ .

Smooth enough and satisfy  $X^n \rightarrow X$ .

$Y^n \rightarrow Y$  in suitable topology.

$$\text{Def: } \int_0^t f(Y_s) \lambda X_s \stackrel{*}{=} \lim_{n \rightarrow \infty} \int_0^t f(Y_s^n) \lambda X_s^n.$$

Rmk: It doesn't hold generally. e.g.:

$$\text{take } X_t^n = -n^{\frac{1}{3}} \cos(nt), \quad Y_t^n = n^{-\frac{1}{3}} \sin(nt).$$

then  $\| \cdot \| = 1$ .  $X_t^n \cdot Y_t^n \rightarrow 0$  ( $n \rightarrow \infty$ ).

$$\text{But } \int_0^\infty Y_s^n \lambda X_s^n = n^{\frac{1}{3}} \infty \rightarrow \infty.$$

And  $\langle X, Y \rangle \mapsto \int_0^t f(Y_s) \lambda X_s$  isn't anti.

prop.  $W = \{ X : [0, T] \rightarrow \mathbb{R}^n \mid X \text{ is cont.}\}$ .

There doesn't exist norm  $\|\cdot\|$  on

$$\bar{E}^{\wedge} := \{ x \in \tilde{C}([0, T], \mathbb{R}^n) \subseteq W \text{ s.t.}$$

i)  $(\bar{E}^{\wedge}, \|\cdot\|)$  contain all BM paths. n.s.

ii)  $J_t : \bar{E}^{\wedge} \times \bar{E}^{\wedge} \rightarrow W$  can be  
 $(x, y) \mapsto \int_0^t y_s dx_s$

extended cont. to  $\bar{E}^{\wedge} \times \bar{E}^{\wedge}$  under  $\|\cdot\|$ .

Rmk: So the topology isn't even rich enough to cover the case of BM.

① So we must consider another method. the main idea of rough theory is to enhance a path  $X$  with some additional data  $\mathcal{X}$ .

Consider  $F$  is smooth function: by Taylor,

$$\begin{aligned} \int_s^t F(X_u) \lambda X_s &= F(X_s)(X_t - X_s) + \int_s^t (F(X_u) - F(X_s)) \\ &\quad \lambda X_u \\ &= F(X_s) X_{t,s} + \int_s^t \int_s^u D F(X_v) \lambda X_v \lambda X_u \\ &= \square + D F(X_s) \int_s^t \int_s^u \lambda X_v \lambda X_u + \\ &\quad \int_s^t \int_s^u (D F(X_v) - D F(X_s)) \lambda X_v \lambda X_u. \\ &= \dots = \sum D^k F(X_s) \int_s^t \int_s^{u_k} \dots \int_s^{u_1} \lambda X_u \dots \end{aligned}$$

So it will expand on  $X = (X^1, \dots, X^n, \dots) :=$   
 $(X_{t,s}, \int_s^t \int_s^u \lambda X_u \otimes \lambda X_u, \int_s^t \int_s^u \int_s^v \lambda X_v \otimes \lambda X_u \otimes \lambda X_u, \dots)$ .  
 where  $\otimes$  is matrix tensor product.

Rmk: Depend on the roughness of path  $X_t$ .

We will refine  $\int_0^t F(X_s) dX_s$  basing on  
 the first  $N$  terms of  $X$ .  
 where  $N$  depends on regularity of  $X$ .

Actually.  $\int_0^t F(X_s) \lambda X_s := \lim_{\pi \rightarrow 0} \sum_{[s,t]} \sum_{k=1}^N$

$F(X_s^k) X_{s,t}^k$ . ( $N^{th}$  order).

For general  $Y_t = (Y_t^0, \dots, Y_t^{N-1})$ . we have:

$$\int_0^t Y_s \lambda X_s := \lim_{\pi \rightarrow 0} \sum_{[s,t]} (Y_s^0 X_{s,t}^0 + Y_s^1 X_{s,t}^1 + \dots + Y_s^{N-1} X_{s,t}^N)$$

$\Rightarrow$  it solves  $\lambda Y = Y \lambda X$ .

Rmk: In fact, if  $X \in C^\alpha$ .  $\alpha > \frac{1}{2}$ .

$$\text{then } \lim_{|\pi| \rightarrow 0} \sum_{[s,t]} \int_s^t \int_s^u \lambda X_u \lambda X_u = 0.$$

which is proved from Young.

Def:  $X \mapsto (X, \lambda X) \xrightarrow{\gamma} Y = y + \int_0^t f(Y_s) dX_s$ .  
 $y$  is called  $T^{1/2} - \text{Lyapunov map}$ .

## (2) Tensor product:

Def:  $V$  is a LS. Family of admissible tensor norms on  $(V^{\otimes n})$  is  $\{l \cdot l_n\}$ .

$$\text{St. i)} \|s \otimes \eta\|_{l_m} \leq \|s\|_m \|\eta\|_n. \quad \forall s \in V, \eta \in V^n$$

$$\text{ii)} \forall \sigma \in S_n. \text{ permutation. } \Pr : V^{\otimes n} \rightarrow V^{\otimes n}, \text{ defined by } \Pr(s_1 \otimes \dots \otimes s_n) = s_{\sigma(1)} \otimes \dots \otimes s_{\sigma(n)}$$

=  $V_{s(1)} \otimes \dots \otimes V_{s(n)}$ . We have:

$$\|\Pr(v)\|_n = \|v\|_n. \quad \forall v \in V^{\otimes n}.$$

Next. we fix  $V$  and  $\{l \cdot l_n\}$ . on  $(V^{\otimes n})$ .

Def: i)  $N \in \mathbb{N}$ .  $N^{\text{th}}$ -truncated tensor algebra

$$\text{on } V \text{ is } T^N(V) := \bigoplus_{n=0}^N V^{\otimes n} = \{s = (s_0, s_1, \dots, s_N) \mid s_k \in V^{\otimes k} \text{ for any } k\}$$

$$s = (s_0, s_1, \dots, s_N) \in T^N(V) \iff s_k \in V^{\otimes k} \text{ for any } k.$$

Rmk:  $T^N(V)$  is LS.

$$\text{ii) Set } (s \otimes \eta)_n = \sum_{k=0}^n s_k \otimes \eta_{n-k} \text{ for any } s, \eta \in T^N(V).$$

$$s, \eta \in T^N(V) \Rightarrow s \otimes \eta \in T^N(V).$$

Rmk: i)  $(T^N(V), +, \otimes)$  is an algebra

With unit  $I = (1, 0, \dots)$

ii)  $\forall s \in T^n(v)$ .  $s$  has an inverse under  $\otimes$ . In particular, for  $s = (1, s_1, \dots, s_n)$

$$s^{-1} = \sum_{k=0}^n (-1)^n (s - 1)^{\otimes n}$$

We denote  $T_1^n(v) \stackrel{\Delta}{=} \{s \in T^n(v) : s_0 = 1\}$ .

If: Note that  $s \otimes \eta = \sum_{\substack{i+j=n \\ i,j \geq 0}} s^i \otimes \eta^j$   
 actually ~~and therefore~~

### (3) Hölder Space:

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Def:  $x \in C^\alpha([0, T], V)$  if  $\sup_{[0, T]} \frac{\|x_{s,t}\|}{|s-t|^\alpha} =: \|x\|_\alpha < \infty$

Rmk: i)  $\|\cdot\|_\alpha$  is a seminorm. While  $\|\cdot\|_\beta$   
 $\stackrel{\sim}{=} |x_0| + \|\cdot\|_\alpha$  is a true norm.

ii)  $C^\beta \subsetneq C^\alpha$ .  $\forall \beta > \alpha$ .  $C^\alpha$  isn't separable.

iii)  $C^\alpha \subsetneq V^{**}$  (e.g. step func.)

Lemma: i) (low-semi conti.)  $X^n \in C^\alpha \rightarrow X$  pointwise.

Then:  $\|X\|_\alpha \leq \liminf_{n \rightarrow \infty} \|X^n\|_\alpha$ .

ii) (Interpolation-)  $\forall 0 < \alpha < \beta \leq 1$ . We have:

$$\|X\|_\alpha \leq \|X\|_\beta^\frac{\alpha}{\beta} \left( \sup_{[0,T]} \|x_{s,t}\| \right)^{1-\frac{\alpha}{\beta}}$$

Cor. If  $0 < \alpha < \beta \leq 1$ .  $X \in C$ . If  $(X^n) \subset C^\beta$

$\rightarrow X$  and  $\sup_n \|X^n\|_\beta < \infty$ . Then:

$X \in C^\beta$  and  $\|X^n - X\|_\alpha \rightarrow 0$ .

Pf:  $\|X\|_\beta \leq \liminf \|X^n\|_\beta \leq \sup_n \|X^n\|_\beta$ .

Besides. by interpolation,

we have  $X^n \rightarrow X$  under  $\|\cdot\|_\alpha$ .

Cor. If  $0 < \alpha < \beta \leq 1$ . If  $\sup_n (\|X^n\|_1 + \|X^n\|_\beta) < \infty$

Then.  $\exists (x_k)$  and  $X \in C^\beta$ . st.

$X^{n_k} \rightarrow X$  under  $\|\cdot\|_\alpha$ .

Pf: By condition, it satisfies  
Ascoli Lemma. So with the  
cor. above. we have the result.

Rmk: It means:  $C^\beta \subset C^\alpha$ .

Dif:  $C^{0,\alpha}([0,T], V) =$  closure of  $C^\alpha$  under  
the norm  $\|\cdot\|_\alpha$ .

Thm. If  $0 < \alpha < 1$ .  $X \in C^{0,\alpha} \Leftrightarrow \lim_{s \rightarrow 0} \sup_{|s-t| < s} \frac{\|X_{s,t}\|}{|s-t|^\alpha} = 0$ .

Pf: ( $\Rightarrow$ ). Let  $X^n \rightarrow X$ .  $X^n \in C^\alpha$ .

$$\text{Not} \quad \frac{\|X_{s,t}\|}{|s-t|^\alpha} \leq \frac{\|X_{s,t} - X_{s,t}^n\|}{|s-t|^\alpha} + \frac{\|X_{s,t}^n\|}{|s-t|^\alpha}$$

Rmk:  $C^{0,\alpha} \not\subseteq C^\beta$ . since  $x_t = t^\beta \in C^\beta / C^{0,\alpha}$

Cor. If  $0 < \tau < \beta \leq 1$ .  $C^\beta \subset C^{0,\tau}$

Cor.  $C(C^{0,\alpha}, \| \cdot \|_1)$  = span of conti. diff. paths

## ② Two-parameter Func.

Def: Set  $A_{(0,T)} = \{(s,t) \in [0,T]^2 \mid s \neq t\}$ .

$$C_2 \stackrel{\Delta}{=} C_2(A_{(0,T)}, V)$$

$$C_2^\alpha \stackrel{\Delta}{=} \{A_{s,t} \in C_2 \mid \sup_{|s-t|^\alpha} \frac{\|A_{s,t}\|}{\|A\|_\alpha} =: \|A\|_\alpha < \infty\}.$$

Rmk: For  $\alpha > 1$ .  $C^\alpha = \text{const. func.}$  But

for  $C_2^\alpha$ . it contains non-trivial.

func. satisfying:

$$\sum_{\pi} \|A_{s,t}\| \leq \|A\|_\alpha |\Sigma_{(0,T)}| / |\pi|^{1-\alpha} \rightarrow 0$$