

- Rough path theory was originally introduced in terms of  $V^p$  = functions of bdd  $p$ -variations.

FACT:  $\cdot C^\alpha \subset V^{1/\alpha}$ ,  $0 < \alpha < 1$

- If  $f \in V^p_C$ , then  $\exists$  reparametrization  $\tau$  st.  $f \circ \tau \in C^{1/p}$ .

- Rough diff eqn (RDE), RPDE

Given an SDE (or SPDE) with a noise  $X$  which is rough in time ( $C^\alpha$ ,  $\alpha \leq 1/2$ ,  $V^p$ ,  $p \geq 2$ ), we lift the noise  $X$  to a rough path  $(X, \mathbb{X})$  and study the original eqn, where an integral is interpreted as a rough integral

- Digression: paracontrolled distributions  
 (Gubinelli - Imkeller - Perkowski '15)

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Consider  $dY = Y dX$

$X \in C^\alpha$ ,  $\alpha \leq 1/2$

- Paracontrolled ansatz:

$$Y = Y' \overset{\uparrow}{\circledast} X + R$$

paraproduct.

$Y' \in C^\alpha$  "Gubinelli deriv"

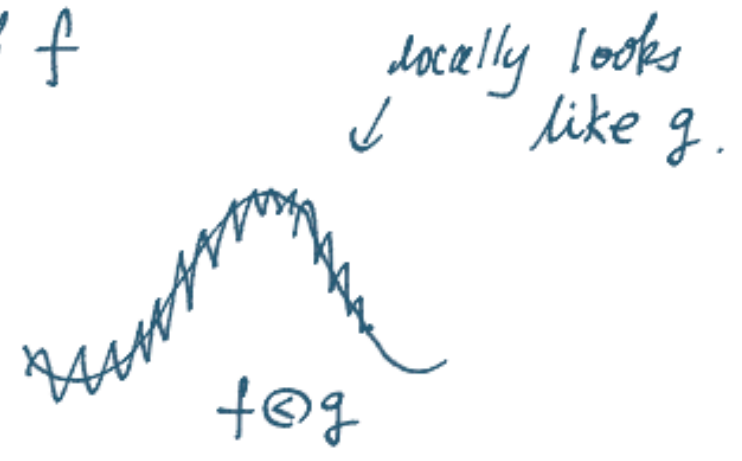
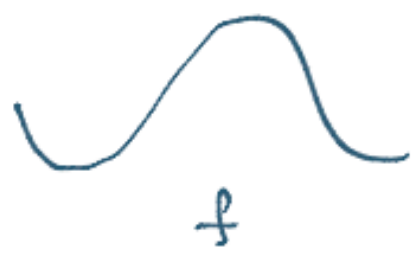
$R \in C^{2\alpha}$ .

- Paraproduct decomposition: (Bony '81)

$$fg = \underbrace{f \circledast g} + \underbrace{f \circledcircledast g}_{\text{wavy}} + f \circledast g$$

$$= \sum_{j < k-2} P_j(f) P_k(g) + \sum_{|j-k| \leq 2} P_j(f) P_k(g) + \sum_{k < j-2} P_j(f) P_k(g)$$

- $f \circledast g =$  paraproduct of  $g$  by  $f$   
 $\text{freq of } g \gg \text{freq of } f$



Note:  $f \circledast g$  is always well defined.  
 $\overset{\uparrow}{C^{\alpha_1}} \quad \overset{\uparrow}{C^{\alpha_2}} \quad f \circledast g \sim \min \{ \alpha_2, \alpha_1 + \alpha_2 \}.$

- $f \oplus g =$  resonant product  
 well defined if  $\alpha_1 + \alpha_2 > 0$ .  
 If well defined, then  $f \oplus g \sim \alpha_1 + \alpha_2$

Back to the paracontrolled ansatz:

④

⊗

$$Y = \underline{Y'} \odot X + \underline{R} \quad X \in C^\alpha, \quad \alpha < \frac{1}{2}.$$

$$=: \underline{Z} + \underline{R}$$

$$\partial_t Y = Y \partial_t X = \underbrace{Y \odot \partial_t X}_{\substack{(\frac{1}{2}-) + (-\frac{1}{2}-) < 0 \\ -\frac{1}{2}-}} + \underbrace{Y \ominus \partial_t X}_{\text{NOT well defined}} + \underbrace{Y \otimes \partial_t X}_{0-}$$

$$\Rightarrow \partial_t Z = (Z + R) \odot \partial_t X \quad \rightarrow Z \sim \frac{1}{2}-$$

$$\partial_t R = \underbrace{(Z + R) \otimes \partial_t X}_{0-} + \underbrace{Z \ominus \partial_t X}_{\text{wavy}} + \underline{R \otimes \partial_t X}$$

$$\left( \text{i.e. } \partial_t Y = \square \odot \partial_t X + \partial_t R \right)$$

By ignoring the resonant products, we expect  $R \sim 1-$

$$\Rightarrow \underline{R \otimes \partial_t X} \quad \text{well defined.}$$

$$(1-) + (-\frac{1}{2}-) > 0$$

⑤

• Issue:  $Z \in \mathcal{A}_t X$   
 $(\frac{1}{2}-) + (-\frac{1}{2}-) < 0.$

• Use the structure of  $Z$ .

$$Z(t) = Z_0 + \underbrace{\int_0^t (Z + R) \in \mathcal{A}_t X(t') dt'}_{\text{const} \Rightarrow Z_0 \in \mathcal{A}_t X \text{ is well defined.}}$$

•  $\left( \int_0^t (Z + R) \in \mathcal{A}_t X(t') dt' \right) \in \mathcal{A}_t X$

$$= \left[ (Z + R) \in \delta X_{t_0} \right] \in \mathcal{A}_t X + \underbrace{\text{Com}_1 \in \mathcal{A}_t X}_{\text{good}}$$

$$= (Z + R) \in \underbrace{(\delta X_{t_0} \in \mathcal{A}_t X)}_{\text{view this as a pad of given data.}} + \text{Com}_1 \in \mathcal{A}_t X + \text{Com}_2$$

$[\in, \in] (Z + R, \delta X_{t_0}, \mathcal{A}_t X)$

$$X_{tt} = \int_r^t \delta X_{ur} dX_u$$

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Now, all the terms make sense.

$$\partial_t Z = (Z + R) \ominus \partial_t X$$

$$\begin{aligned} \partial_t R = & (Z + R) \oplus \partial_t X + R \ominus \partial_t X + Z_0 \ominus \partial_t X \\ & + (Z + R) \ominus (\delta X_{t_0} \ominus \partial_t X) + \text{com}_1 \ominus \partial_t X + \text{com}_2 \end{aligned}$$

- Solve the system for  $Z$  and  $R$  by a standard contraction argument.  
(strictly speaking, we need to multiply by a (smooth) cutoff in time to prove LWP.)

Summary:

$$\text{paracontrolled: } X \xrightarrow{\textcircled{1}} (\partial_t X, \delta X_{t_0} \ominus \partial_t X) \xrightarrow{\textcircled{2}} (Z, R) \mapsto Y = Z + R$$

$$\text{controlled path: } X \xrightarrow{\textcircled{1}} (X, \mathbb{X}) \xrightarrow{\textcircled{2}} (Y', R) \text{ (or } (Y, Y')) \mapsto Y$$

- ① : stochastic analysis to lift  $X$  to an enhanced data set/rough path
- ② : deterministic analysis (NO probability.)

• Pathwise local well-posedness of SNLS with a multip. noise ⑦  
on  $\mathbb{T}^d$ .

$$\begin{cases} i\partial_t u = \Delta u + N(u) + \underbrace{u \Phi}_{\uparrow} \sum & \text{on } \mathbb{T}^d \\ u|_{t=0} = u_0 \end{cases}$$

linear in  $u$ . (but nonlinear in noise)

- Interaction representation:  $v(t) = S(-t) u(t) = e^{it\Delta} u(t)$
- For simplicity, set  $N(u) = 0$ .

$$v(t) = u_0 - i \int_0^t S(t-\tau) \underbrace{(S(\tau) v(\tau) \Phi d\mathbb{Z}(\tau))}_{= dX_\tau}$$

Set  $Y_t = v(t)$  and write

$$Y_t = Y_r - i \int_r^t \underbrace{S_{-t} dX_\tau}_{\uparrow} \underbrace{S_\tau Y_\tau}_{\leftarrow}$$

make things harder: need to lose spatial reg. to compute  $C_t^\alpha$ -norms.

$$D_t^\alpha e^{it|m|^2} \sim \underline{|m|}^{2\alpha} e^{it|m|^2}$$

Young case:  $X \in C_t^\alpha H^s$ ,  $\alpha > 1/2$   $\Phi \in HS(L^2; H^s)$  (8)

$$\int_r^t S_{-t} dX_{t'} S_{t'} \underline{Y_{t'}} = \int_r^t S_{-t} dX_{t'} S_{t'} \underline{Y_r}$$

$$+ \left[ -i \int_r^t S_{-t'} dX_{t'} S_{t'} \int_r^{t'} S_{-t''} dX_{t''} S_{t''} Y_{t''} \right]$$

$$\Rightarrow \delta \square = - \delta \left( \int_r^t S_{-t} dX_{t'} S_{t'} Y_r \right)$$

$=: X'_{tr}$  operator

$$= X'_{t_1 t_2} \delta Y_{t_2 t_3} = X' \delta Y$$

**(\*\*)** If  $X'$  maps  $H^s$  to  $H^s$  with  $\|X'_{tr}\|_{L(H^s; H^s)} \lesssim |t-r|^\alpha$ ,

then

$$\delta \square = X' \delta Y \in C_3^{2\alpha} > 1$$

$|t_1 - t_2|^\alpha \rightarrow |t_2 - t_3|^\alpha$

denote the class by  $C_2^\alpha L(H^s; H^s)$



By the sewing lemma, we obtain

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$$\int_r^t S_{-t} dX_t S_{-t} Y_t = (\text{Id} - \Lambda f) \underline{X^1 Y}$$

operator.

Q: How to verify ~~(\*)~~?

$$\widehat{X_{tr}^1 f}(m) = \sum_{n=n_1+n_2} \int_r^t e^{-it(|m|^2 - |n_2|^2)} \underbrace{\phi(m, \cdot)}_{\in dx} d\beta_{n_1}^H(\cdot) \underbrace{\widehat{f}(n_2)}_{\text{wavy}}$$

Assume:

•  $\widehat{\Phi f}(n) = \phi(m) \widehat{f}(n)$

and  $\Phi \in \text{HS}(L^2; H^s)$ ,  $\underline{s > d/2}$   
algebra

• when  $X$  is a fBM in time with a Hurst parameter  $H > 1/2$

(slightly smoother in time than a BM)

Main tool: random matrix / tensor estimate

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from Deng-Nahmod-Yue. Invent. Math. '21  
Oh-Wang-Zine SPDE 122.

$$\Rightarrow X' \in C_2^\alpha L(H^s; H^s), \quad \alpha > \frac{1}{2} \quad (\alpha = H^-)$$

$\rightarrow$  perform a contraction argument as in Lec 13.

Rough case:  $X = \Phi W$ ,  $\Phi \in HS(L^2; H^s)$ ,  $s > d/2$ .  
 $\uparrow$   $L^2$ -cylindrical Wiener process

$$Y_t = Y_r - i \int_r^t S_{-t} dX + S_t \underbrace{(D^{-\varepsilon} Y_t)}_W$$

Controlled path:  $\delta Y_{tr} = X'_{tr} Y'_r + R_{tr}^Y$   
 $= -i \underbrace{X'_{tr}}_W \underbrace{D^{-\varepsilon} Y_r}_W + R_{tr}^Y$

~~xxx~~

$$\int_r^t S_{-t} dX_t S_{-t} D^{-\varepsilon} Y_t$$

$$\stackrel{***}{=} \underbrace{\int_r^t S_{-t'} dX_{t'} S_{-t'} D^{-\varepsilon} Y_r}_{X'_{tr}}$$

$$- i \underbrace{\int_r^t S_{-t_1} dX_{t_1} S_{t_1} X'_{t'r} D^{-2\varepsilon} Y_r}_{= \cancel{X}^2_{tr}}$$

$$+ \boxed{X'_{tr} R_{\cdot r}}$$

Remark:  $(X', \cancel{X}^2)$  is an operator-valued rough path adapted to the Schrödinger flow, satisfying

$$\cancel{X}^2_{t_1 t_3} - \cancel{X}^2_{t_1 t_2} - \cancel{X}^2_{t_2 t_3} = X_{t_1 t_2} \circ X_{t_2 t_3}$$

↑  
composition of operators.

$$\delta \square = -\delta(X^1 D^{-\varepsilon} Y) + i\delta(X^2 D^{-2\varepsilon} Y)$$

$$\cdot -\delta(X^1 D^{-\varepsilon} Y)_{t_1 t_2 t_3} = X^1_{t_1 t_2} \delta D^{-\varepsilon} Y_{t_2 t_3}$$

$$\stackrel{\text{Chen}}{=} \underbrace{-i X^1_{t_1 t_2} \circ X^1_{t_2 t_3} D^{-2\varepsilon} Y_{t_3}}_{\alpha + 2\alpha = 3\alpha > 1} + \underline{X^1_{t_1 t_2} D^{-\varepsilon} R_{t_2 t_3}}$$

$$\cdot i\delta(X^2 D^{-2\varepsilon} Y) = i X^2_{t_1 t_3} D^{-2\varepsilon} Y_{t_3} - i X^2_{t_1 t_2} D^{-2\varepsilon} Y_{t_2} - i X^2_{t_2 t_3} D^{-2\varepsilon} Y_{t_3}$$

$$\stackrel{\text{Chen}}{=} \underline{-i X^2_{t_1 t_2} \delta D^{-2\varepsilon} Y_{t_2 t_3}} + \underbrace{i X^2_{t_1 t_2} \circ X^2_{t_2 t_3} D^{-2\varepsilon} Y_{t_3}}_{\text{crossed out}}$$

$$2\alpha + \alpha = 3\alpha > 1 \text{ if } \alpha > 1/3 \text{ (see next page)}$$

⇒ Apply the sewing lemma;

$$\int_r^t S_{-t} dX_t S_t Y_t = (\text{Id} - \Omega \delta)(X^1 Y - i X^2 Y)$$

• For this computation, we need

$$X^1 \in C_2^\alpha \mathcal{L}(H^s; H^{s-\varepsilon})$$

$$X^2 \in C_2^\alpha \mathcal{L}(H^s; H^{s-2\varepsilon})$$

← fails when  $\varepsilon=0$

Thm: (Oh-Zheng '22). Let  $\varepsilon \geq 0$ ,  $\Phi \in HS(L^2; H^s)$ ,  $s > d/2$ .

$$\text{Then, SNLS: } \begin{cases} i\partial_t u = \Delta u + \mathcal{N}(u) + u \Phi \Xi \\ u|_{t=0} = u_0 \end{cases}$$

↑  $\mathcal{N}(u) = |u|^{p-1}u$ ,  $p \in 2\mathbb{N}+1$   
↑ space-time white noise

• When  $\varepsilon=0$ , this argument fails and we need another idea.

• heat case: Gubinelli-Tindel AOP '10.

Also, see Hairer-Pardoux JMSJ '15.  
for the regularity structure approach.