

Lec 8 : 15 / 03 / 22 (Tue)

①

Stochastic nonlinear heat eqn

with multiplicative space-time white noise

$$(SNLH) \quad \begin{cases} \partial_t u - \Delta u = N(u) + \sigma(u) \underline{\xi} & \text{on } \mathbb{T}^d \\ u|_{t=0} = u_0 \end{cases}$$

ξ = space-time white noise
i.e. $\Phi = \text{Id}$.

• Duhamel formulation (= mild formulation)

$$u(t) = P(t) u_0 + \int_0^t P(t-\tau) N(u)(\tau) d\tau + \underbrace{\int_0^t P(t-\tau) \sigma(u)(\tau) dW(\tau)}_{=: \Psi[u](t)}$$

$P(\tau) = e^{+\Delta \tau}$

• Schauder estimate: $1 \leq p \leq q \leq \infty$, $s \geq 0$ (2)

$$(*) \quad \|P(t)f\|_{W^{s,q}} \lesssim \underline{t^{-\frac{d}{2}(\frac{1}{p}-\frac{1}{q})-\frac{s}{2}}} \|f\|_{L^p}$$

for any $t > 0$ on \mathbb{R}^d

$0 < t \leq 1$ on \mathbb{T}^d

(For $\tilde{P}(t) = e^{t(\Delta-1)}$, then

(*) holds for any $t > 0$ even on \mathbb{T}^d .

• Besov spaces & Hölder-Besov spaces

$$\text{Hölder norm: } \|u\|_{C^s} = \sup_{x \neq y} \frac{|u(x) - u(y)|}{|x - y|^s}, \quad 0 < s < 1$$

↑
a semi-norm

• Littlewood - Paley freq projector:

• Grafakos

• Bahouri - Chemin - Danchin (3)

• P_j = "smooth" freq projector onto freq.

$$\{|n| \sim 2^j\}, j \geq 1$$

$$\{|n| \leq 1\}, j = 0$$

• $\varphi \in C_c^\infty(\mathbb{R}; [0, 1])$

$$\varphi(\xi) \equiv 1 \text{ for } |\xi| \leq 1$$

$$\varphi(\xi) \equiv 0 \text{ for } |\xi| \geq 2$$



$$\text{Now, set } \varphi_j(\xi) = \begin{cases} \varphi(|\xi|), & j=0 \\ \varphi\left(\frac{|\xi|}{2^j}\right) - \varphi\left(\frac{|\xi|}{2^{j-1}}\right), & j \geq 1. \end{cases}$$

$$\equiv 1 \text{ for } |\xi| \sim 2^j$$

$$\equiv 0 \text{ for } |\xi| \ll 2^j$$

$$\text{or } |\xi| \gg 2^j.$$

$$\Rightarrow \widehat{P_j u}(n) = \varphi_j(n) \widehat{u}(n)$$

(4)

- Littlewood - Paley thm: $1 < p < \infty$.

$$\left\| \underbrace{\left(\sum_{j=0}^{\infty} |P_j(u)|^2 \right)^{1/2}}_{\text{square function}} \right\|_{L^p} \sim \|f\|_{L^p}$$

- LP characterization of H^s :

$$\|u\|_{H^s} \sim \left\| \left\| 2^{js} \|P_j(u)\|_{L_x^2} \right\|_{\ell_{j \geq 0}^2} \right\|, \quad s \in \mathbb{R}$$

- Besov space $B_{p,q}^s$ (or $B_{p,q}^{s,p}$)

$$\|u\|_{B_{p,q}^s} = \left\| \left\| 2^{js} \|P_j(u)\|_{L_x^p} \right\|_{\ell_{j \geq 0}^q} \right\|$$

$$1 \leq p, q \leq \infty, \quad s \in \mathbb{R}$$

On \mathbb{R}^d , by setting $\Psi_j(\xi) = \varphi\left(\frac{|\xi|}{2^j}\right) - \varphi\left(\frac{|\xi|}{2^{j-1}}\right)$, $j \in \mathbb{Z}$ (5)
 and $\widehat{Q_j u}(\xi) = \Psi_j(\xi) \widehat{u}(\xi)$, we can define

the homogeneous Besov space $\dot{B}_{p,q}^s$ by

$$\|u\|_{\dot{B}_{p,q}^s(\mathbb{R}^d)} = \left\| \left\{ 2^{js} \|Q_j(u)\|_{L^p_x} \right\}_{j \in \mathbb{Z}} \right\|_q$$

Then, we have

$$\|u\|_{C^s} = \|u\|_{\dot{B}_{\infty,\infty}^s}, \quad 0 < s < 1$$

• If we set $\Lambda^s = C^s \cap L^\infty$, then we have

on $\mathbb{T}^d, \mathbb{R}^d$.

$$\|u\|_{\Lambda^s} = \|u\|_{\dot{B}_{\infty,\infty}^s}, \quad 0 < s < 1$$

\uparrow Lipschitz space ($C^s = \Lambda^s$)
 make sense for any $s \in \mathbb{R}$.

• Hölder - Besov space: $\mathcal{C}^s = B_{\infty, \infty}^s$, $s \in \mathbb{R}$ (6)

• $\|u\|_{\mathcal{C}^s} = \|u\|_{B_{\infty, \infty}^s} = \sup_{j \geq 0} 2^{js} \|P_j(u)\|_{L_x^\infty}$.

• $\mathcal{C}^s \supset W^{s, \infty}$

• $s > 0$: \mathcal{C}^s is an algebra.

• Schauder: $\|P(t)f\|_{\mathcal{C}^{s_2}} \lesssim t^{-\frac{s_2-s_1}{2}} \|f\|_{\mathcal{C}^{s_1}}$
 $s_2 \geq s_1$

• We want to study the stochastic convolution

$$\Psi[u](t) = \int_0^t P(t-t') \sigma(u)(t') dW(t')$$

via BDG ineq. (in $C_T \mathcal{C}_x^s$).

• Ideal property of γ -radonifying operator:

(7)

$$B_1 \xrightarrow{S} B_2 \xrightarrow{\Phi} B_3 \xrightarrow{T} B_4$$

$$\Phi \in \gamma(B_2, B_3)$$

$$S \in L(B_1, B_2), \quad T \in L(B_3, B_4)$$

\uparrow lin, bdd

Then,

$$T \circ \Phi \circ S \in \gamma(B_1, B_4) \text{ with}$$

$$\|T \circ \Phi \circ S\|_{\gamma(B_1, B_4)} \lesssim \|T\|_{L(B_3, B_4)} \|\Phi\|_{\gamma(B_2, B_3)} \|S\|_{L(B_1, B_2)}$$

(*)

$$C^s(\mathbb{T}^d) \supset W_{\text{Young}}^{s, \infty}(\mathbb{T}^d) \supset W_{\text{Sobolev}}^{s+\varepsilon, r}(\mathbb{T}^d), \quad \varepsilon r > d.$$

$r \gg 1 \text{ s.t. } \varepsilon = 0+$

For $t \geq 0$. Then, by BDG ineq,

$$\mathbb{E} \left[\|\Psi[u](t)\|_{\ell_x^s}^p \right] \leq \mathbb{E} \left[\sup_{0 \leq t_0 \leq t} \left\| \int_0^{t_0} P(t-t') \sigma(u)(t') dW(t') \right\|_{W_x^{s+\varepsilon, r}}^p \right] \quad (8)$$

$$\stackrel{\text{BDG}}{\lesssim} \mathbb{E} \left[\left(\int_0^t \|P(t-t') \circ M_{\sigma(u)(t')}\|_{\gamma(L^2; W^{s+\varepsilon, r})}^2 dt' \right)^{p/2} \right]$$

When is this finite?

Note: $\text{Id} \in \gamma(L^2; W^{d, r})$ iff $\underline{\alpha < -d/2}$
 $1 \leq r \leq \infty$

① $\sigma(u) \equiv 1$ (i.e. additive case)

$$\|P(t-t') \circ \text{Id}\|_{\gamma(L^2; W^{s+\varepsilon, r})}$$

$$\lesssim \|P(t-t')\|_{L(W^{-\frac{d}{2}-\varepsilon, r}, W^{s+\varepsilon, r})} \| \underbrace{\text{Id} \|_{\gamma(L^2; W^{-\frac{d}{2}-\varepsilon, r})} \lesssim 1$$

Schauder

$$\lesssim (t-t')^{-\frac{(s+\frac{d}{2}+2\varepsilon)}{2}}$$

⑨

$$\Rightarrow \int_0^t \|P(t-t') \circ \text{Id}\|_{\mathcal{Y}(L^2; W^{s+\varepsilon, r})}^2 dt'$$

$$\lesssim \int_0^t (t-t')^{-\left(s+\frac{d}{2}+2\varepsilon\right)} dt'$$

$$< \infty \iff s + \frac{d}{2} + 2\varepsilon < 1$$

$$\iff \underline{s < 1 - \frac{d}{2} - 2\varepsilon}$$

On the other hand, we need $s > 0$ (s.t. \mathcal{C}^s is an algebra) to handle the nonlinearity $N(u) = u^k$

$$\Rightarrow \text{Hence, } 0 < s < 1 - \frac{d}{2} - 2\varepsilon$$

$$\Rightarrow \boxed{d=1}$$

Rmk: • In the additive case ($\sigma(u) \equiv 1$), we do not need to work in $L^2(\Omega; C_T C_x^s)$ with the BDG ineq and the truncation method. Instead, we can directly prove pathwise local well-posedness. (10)

- We imposed the condition $s > 0$ s.t. u is a function (in x). In the additive case, the solution theory can be built for higher dimensions ($d = 2, 3$) even when $u(t)$ is only a distribution (in x). In this case, we need to introduce a renormalization to give a proper meaning to the nonlinearity $N(u)$. See my course note from Spring 2021.

② general case $\sigma(u) = u^\gamma$

⑪

Correction In this case, we can NOT close the argument in $C_T C_x^s$, simply using the BDG ineq. We instead work in $C_T W_x^{s,r}$ for some $s \geq 0$, $z \in r < \infty$.

\Rightarrow By BDG ineq, we need to study

$$\| P(t-t') \circ M_{\sigma(u)(t)} \|_{\gamma(L^2; W^{s,r})}$$

$$\sim \left(\mathbb{E}' \left[\left\| \sum_k \underbrace{g_k}_{\text{for } \{g_k\}} \langle \nabla \rangle^s P(t-t') \circ M_{\sigma(u)(t)}(e_k) \right\|_{L_x^r}^2 \right] \right)^{1/2}$$

$e_k = e^{2\pi i k x}$

$$\sim \| \dots \|_{L_x^r} \| \dots \|_{L^r(\Omega')} = \| \dots \|_{L^r(\Omega')} \| \dots \|_{L_x^r}$$

u on $\Omega \leftarrow$ indep
 $\{g_k\}$ on $\Omega' \leftarrow$ indep

$$\sim \| \dots \|_{L^2(\Omega')} \| \dots \|_{L_x^r}$$

$$\sim \left\| \left(\sum_k \left| \langle \nabla \rangle^s P(t-t') \circ M_{\sigma(u)(t)}(e_k) \right|^2 \right)^{1/2} \right\|_{L_x^r}$$

In general, $\|\Phi\|_{\mathcal{S}(L^2; L^r)} \sim \left\| \left(\sum_k |\Phi(e_k)|^2 \right)^{1/2} \right\|_{L^r_x}$.

$$\begin{aligned} & \left\| \left(\sum_k \left| \langle \nabla \rangle^s P(t-\tau) \circ M_{\sigma(u)(\tau)}(e_k) \right|^2 \right)^{1/2} \right\|_{L^r_x} \\ &= \sum_{n \in \mathbb{Z}^d} e^{-(t-\tau)|n|^2} \langle n \rangle^s \widehat{\sigma(u)(\tau)}(n-k) e_n(x) \end{aligned}$$

Minkowski

$$\leq_{r \geq 2} \left\| \left\| \dots \right\|_{L^r_x} \right\|_{l^2_k}$$

Hausdorff-Young

$$\leq \left\| \left\| e^{-(t-\tau)|n|^2} \langle n \rangle^s \widehat{\sigma(u)(\tau)}(n-k) \right\|_{l^{r'}_n} \right\|_{l^2_k} \quad \underline{r' \leq 2}$$

Minkowski

$$\begin{aligned} & \leq \left\| e^{-(t-\tau)|n|^2} \langle n \rangle^s \underbrace{\left\| \widehat{\sigma(u)(\tau)}(n-k) \right\|_{l^2_n}}_{= \|\sigma(u)(\tau)\|_{L^2_x}} \right\|_{l^{r'}_n} \end{aligned}$$

Now, we estimate

$$\| e^{-(t-t')|m|^2} \langle m \rangle^s \|_{l_n^{r'}} = \left(\sum_{n \in \mathbb{Z}^d} e^{-r'(t-t')|m|^2} \langle m \rangle^{sr'} \right)^{1/r'}$$

(13)

$$\lesssim \left(\sum_{n \in \mathbb{Z}^d} \frac{1}{(t-t')^\alpha \langle m \rangle^{2d-sr'}} \right)^{1/r'} \quad e^{-r'(t-t')|m|^2} \lesssim \min \left(\frac{1}{(t-t')|m|^2}, 1 \right)^\alpha$$

for any $\alpha \geq 0$

$$\lesssim \frac{1}{(t-t')^{\alpha/r'}} , \text{ provided that } 2d - sr' > d$$

\Leftrightarrow

$$s < \frac{2}{r'} d - \frac{d}{r'}$$

(14)

\Rightarrow Under $s < \frac{2}{r'} d - \frac{d}{r'}$ we then have

$$\int_0^t \| P(t-t') \circ M_{\sigma(u)(t')} \|_{\gamma(L^2; W^{s,r})}^2 dt'$$

$$\lesssim \underbrace{\int_0^t \frac{1}{(t-t')^{2d/r'}} dt'}_{\lesssim 1 \text{ iff } \underline{2d < r'}} \times \| \sigma(u) \|_{C_T L_x^2}^2, \quad 0 \leq t \leq T$$

$$\lesssim \| \sigma(u) \|_{C_T L_x^2}^2$$

$$\begin{cases} s < \frac{2}{r'} d - \frac{d}{r'} \\ 2d < r' \end{cases} \Rightarrow \underline{0 \leq s < 1 - \frac{d}{r'}}, \quad 1 < r' \leq 2$$

$$\Rightarrow \boxed{d=1}$$

$$\begin{array}{c} \Updownarrow \\ \underline{0 \leq s < \frac{1}{r'}} \text{ when } d=1. \end{array}$$

With $\sigma(u) = u^\gamma$, we have

$$\|\sigma(u)\|_{C_T L_x^2} = \|u\|_{C_T L_x^{2\gamma}}^\gamma$$

$$\lesssim \|u\|_{C_T W_x^{s,r}}^\gamma$$

• $s \geq 0$ if $r \geq 2\gamma$

• otherwise, by Sobolev ($d=1$),

$$s \geq \frac{1}{r} - \frac{1}{2\gamma} \quad \left(< \frac{1}{r} \right)$$

• Note: The BDG ineq with the computations on pp. 11-15 shows

$$\mathbb{E} \left[\|\Psi[u](t)\|_{W^{s,r}}^p \right] \lesssim \mathbb{E} \left[\|u\|_{C_T W_x^{s,r}}^{\gamma p} \right]$$

for any fixed $0 \leq t \leq T$. BUT, we need to insert

$\sup_{0 < t < T}$ under the expectation on LHS. See pp. 18-23 below.

• As for the nonlinear part, we have

$$\| \int_0^t P(t-t') N(u)(t') dt' \|_{C_T W_x^{s,r}} \leq \| \int_0^t \| P(t-t') N(u)(t') \|_{W^{s,r}} dt' \|_{C_T}$$

Schauder

$$\lesssim \| \int_0^t (t-t')^{-\frac{s}{2} - \frac{1}{2}(1-\frac{1}{r})} \| N(u)(t') \|_{L_x^1} dt' \|_{C_T}$$

Note that

$$\frac{s}{2} + \frac{1}{2}(1-\frac{1}{r}) < 1$$

since $0 < s < \frac{1}{r} < 1$.

$$\leq \| u \|_{C_T L_x^k}^k$$

$$\lesssim \| u \|_{C_T W_x^{s,r}}^k$$

$$N(u) = u^k$$

for $\begin{cases} s \geq 0 & \text{if } r \geq k \\ s \geq \frac{1}{r} - \frac{1}{k} & (< \frac{1}{r}) \end{cases}$

$$\lesssim T^\theta \| u \|_{C_T W_x^{s,r}}^k$$

With a truncation, we can perform a contraction argument in $L_{ad}^p(\Omega; C_T W_x^{s,r})$ to construct U_R by choosing

$p \gg 1$. see p.20

$$\textcircled{1} \quad 0 \leq s \leq \frac{1}{r}$$

$$\textcircled{2} \quad r \geq 2\sigma \quad \text{or} \quad s \geq \frac{1}{r} - \frac{1}{2\sigma}$$

$$\textcircled{3} \quad r \geq k \quad \text{or} \quad s \geq \frac{1}{r} - \frac{1}{k}$$

\Rightarrow Then, LWP of (SNLH) in $W_x^{s,r}(\Pi)$.

· Back to

$$\mathbb{E} \left[\underbrace{\sup_{0 < t < T}}_{\text{set } T=1} \|\Psi[u](t)\|_{W^{s,r}}^p \right]$$

(18)

We repeat the argument on pp. 18-20 in Lec 3.

$$\text{Let } t_{l,k} = \frac{l}{2^k}, \quad l = 0, 1, 2, \dots, 2^k.$$

and write

$$(*) \quad \Psi[u](t) = \sum_{k=1}^{\infty} \left(\Psi[u](t_{l_k, k}) - \Psi[u](t_{l_{k-1}, k-1}) \right)$$

for some $l_k = l_k(t) \in \{0, 1, \dots, 2^k\}$

Note: In (*), we used continuity (in t) of $\Psi[u]$

(which we want to show). Strictly speaking, we perform the following argument to $\pi_N u$ in place of u and take $N \rightarrow \infty$ ($\pi_N =$ smooth freq. projection onto freq $\{|n| \leq N\}$)

$$\Rightarrow \sup_{0 \leq t \leq 1} \|\Psi[u](t)\|_{W_x^{s,r}}$$

$$\leq \sum_{k=1}^{\infty} \max_{0 \leq l_k \leq 2^k} \|\Psi[u](t_{l_k, k}) - \Psi[u](t_{l_{k-1}, k-1})\|_{W_x^{s,r}}$$

where $|t_{l_k, k} - t_{l_{k-1}, k-1}| \leq 2^{-k}$.

$$\Rightarrow \left\| \sup_{0 \leq t \leq 1} \|\Psi[u](t)\|_{W_x^{s,r}} \right\|_{L^p(\Omega)}$$

$$\leq \sum_{k=1}^{\infty} \left\| \Psi[u](t_{l_k, k}) - \Psi[u](t_{l_{k-1}, k-1}) \right\|_{L^p(\Omega; l_k^p W_x^{s,r})}$$

$$\lesssim \sum_{k=1}^{\infty} 2^{k/p} \max_{0 \leq l_k \leq 2^k} \left\| \Psi[u](t_{l_k, k}) - \Psi[u](t_{l_{k-1}, k-1}) \right\|_{L^p(\Omega; W_x^{s,r})}$$

Claim:

$$\sup_{0 \leq t_1 < t_2 \leq T} \|\Psi[u](t_2) - \Psi[u](t_1)\|_{L^p(\Omega; W_x^{s,r})} \lesssim (t_2 - t_1)^\theta \|\sigma(u)\|_{L^p(\Omega; C_T L_x^2)}$$

for some $\theta > 0$, indep of p .

⇒ By assuming claim, we have

$$\begin{aligned} & \|\sup_{0 \leq t \leq 1} \|\Psi[u](t)\|_{W^{s,r}}\|_{L^p(\Omega)} \\ & \lesssim \underbrace{\sum_{k=1}^{\infty} 2^{k/p} 2^{-k\theta}}_{\lesssim 1} \|\sigma(u)\|_{L^p(\Omega; C([0,1]); L_x^2)} \end{aligned}$$

by choosing $p \gg 1$.

Pf of Claim: For $0 \leq t_1 < t_2 \leq T$, we have

(2)

$$\Psi[u](t_2) - \Psi[u](t_1)$$

$$= \int_{t_1}^{t_2} P(t_2 - t') \sigma(u)(t') dW(t')$$

$$+ \int_0^{t_1} [P(t_2 - t') - P(t_1 - t')] \sigma(u)(t') dW(t')$$

$$=: I(t_1, t_2) + II(t_1, t_2).$$

By BDG ineq and repeating the computations on pp. 11-14

$$\| I(t_1, t_2) \|_{L^p(\Omega; W_x^{s,r})} \lesssim \left(\int_{t_1}^{t_2} \frac{1}{(t_2 - t')^{1 - \frac{2\theta}{p}}} dt' \right)^{p/2} \| \sigma(u) \|_{L^p(\Omega; C_T W_x^{s,r})}$$

$$\sim (t_2 - t_1)^\theta \quad \text{for some small } \theta > 0.$$

As for II, first note that

$$| \mathcal{F}_x ([P(t_2 - t') - P(t_1 - t')] f)(n) |$$

(*)

$$= | e^{-(t_2 - t_1)|m|^2} - 1 | \cdot e^{-(t_1 - t')|m|^2} | \hat{f}(n) |$$

$$\stackrel{MVT}{\lesssim} (t_2 - t_1)^\theta |m|^{2\theta} e^{-(t_1 - t')|m|^2} | \hat{f}(n) |$$

Using (*), we repeat the computation on p. 13.

$$\| (e^{-(t_2 - t')|m|^2} - e^{-(t_1 - t')|m|^2}) \langle m \rangle^s \|_{l_n^{r'}}$$

$$\lesssim (t_2 - t_1)^\theta \left(\sum_{n \in \mathbb{Z}^d} \frac{1}{(t_1 - t')^\alpha} \langle m \rangle^{2\alpha - sr' - 2\theta} \right)$$

$$\lesssim (t_2 - t_1)^\theta \frac{1}{(t_1 - t')^{\frac{\alpha}{r'}}}, \text{ if } \boxed{2\alpha - sr' - 2\theta > d}$$

(A)

Also, note $\int_0^{t_1} \frac{1}{(t_1 - t')^{\frac{2d}{r'}}} dt' \lesssim 1$

(23)

if $\frac{2d}{r'} < 1$

(B)

- By choosing $\theta > 0$ suff. small, the conditions (A) & (B) are satisfied in view of the conditions on pp 13-14.

□

Rmk: • On p. 12, we put $\sigma(u)(t')$ in L^2_x

when this term has more regularity. Thus, we can improve the argument a bit but it seems that we can not close the argument in $C_T C_x^s$ via the BD&G ineq on p. 8 (since we would get two contradictory conditions $s > \frac{1}{2} - \frac{1}{r} > \frac{1}{2} - \varepsilon$ and $s < \frac{1}{2} - \varepsilon$.)

\uparrow
 $\varepsilon r > 1$

• In the following, we directly show $\Psi[u]$ in $C_T C_x^{s_0}$ for some $s_0 > 0$, where $u \in L^p_{\text{ad}} C_T H_x^{\frac{1}{2}-}$, $p \gg 1$.

In view of Kolmogorov's conti. criterion, it suffices to show (25)

$$\mathbb{E} \left[\left| \Psi[u](t_1, x_1) - \Psi[u](t_2, x_2) \right|^p \right] \\ \lesssim \left| (t_1, x_1) - (t_2, x_2) \right|^{1+\theta} \text{ for some } \theta > 0 \\ \text{and } p \gg 1$$

$$\cdot \Psi[u](t_1, x_1) - \Psi[u](t_2, x_2)$$

$$= \left(\Psi[u](\underline{t_1}, \underline{x_1}) - \Psi[u](\underline{t_2}, \underline{x_1}) \right) \leftarrow \begin{array}{l} \text{use the ideas from} \\ \text{the pf of Claim} \\ \text{on p. 20} \end{array} \\ + \left(\Psi[u](\underline{t_2}, \underline{x_1}) - \Psi[u](\underline{t_2}, \underline{x_2}) \right)$$

• We focus only on

$$\mathbb{E} \left[\left| \Psi[u](t, x_1) - \Psi[u](t, x_2) \right|^p \right]$$

(26)

$$\Psi[u](t, x_1) - \Psi[u](t, x_2)$$

$$= \sum_{n \in \mathbb{Z}^d} \sum_{k \in \mathbb{Z}^d} \int_0^t e^{-(t-t')|m|^2} \widehat{\sigma(u)(t')} (n-k) d\beta_k(t') (e_n(x_1) - e_n(x_2))$$

$$= \sum_{k \in \mathbb{Z}^d} \int_0^t \left[\sum_{n \in \mathbb{Z}^d} e^{-(t-t')|m|^2} \widehat{\sigma(u)(t')} (n-k) (e_n(x_1) - e_n(x_2)) \right] d\beta_k(t')$$

• By the BDG inequality for scalar martingales, we have

$$\mathbb{E} \left[\left| \Psi[u](t, x_1) - \Psi[u](t, x_2) \right|^p \right]$$

$$\lesssim \mathbb{E} \left[\left(\sum_{k \in \mathbb{Z}^d} \int_0^t \left| \sum_{n \in \mathbb{Z}^d} \dots \right|^2 dt' \right)^{p/2} \right]$$

By MVT, we have $|e_n(x_1) - e_n(x_2)| \lesssim |m|^\delta |x_1 - x_2|^\delta$ (27)

for any $0 \leq \delta \leq 1$.

$$\sum_k \int_0^t \left| \sum_n \quad \right|^2 dt' \\ \lesssim |x_1 - x_2|^{2\delta} \int_0^t \|a_n * b_n\|_{L_n^2}^2 dt'$$

$$a_n = e^{-(t-t')|m|^2} |m|^\delta, \quad b_n = \widehat{\sigma(u)(t')}(n)$$

$$\lesssim \frac{1}{(t-t')^\theta \langle m \rangle^{2\theta-\delta}}$$

$$\frac{1}{2} + 1 = \frac{1}{1+\delta} + \frac{1}{2-\delta}$$

Young

$$\lesssim |x_1 - x_2|^{2\delta} \underbrace{\int_0^t \frac{1}{(t-t')^{2\theta}} dt'}_{\lesssim 1 \text{ for } \theta < 1/2} \underbrace{\left\| \frac{1}{\langle m \rangle^{2\theta-\delta}} \right\|_{L_n^1}^2}_{\lesssim 1 \text{ by choosing } \theta = \frac{1}{2} - \delta} \underbrace{\| \widehat{\sigma(u)(t')}(n) \|_{L_n^{2-\delta}}^2}_{\text{by Hölder}}$$

Hölder

$$\lesssim |x_1 - x_2|^{2\delta} \| \sigma(u) \|_{C_T H_x^\alpha}^2, \quad \alpha = 0+$$

$\delta > 0$ small.

Lastly, by the fractional Leibniz rule with $\sigma(u) = u^\sigma$ (28)

$$\|\sigma(u)\|_{C_T H_x^\alpha} \lesssim \|u\|_{C_T W_x^{\alpha, 2\sigma}}^\sigma$$

$$\stackrel{\substack{\text{Sobolev} \\ \text{see p.15}}}{\lesssim} \|u\|_{C_T H_x^s}^\sigma, \quad s \geq \alpha + \frac{1}{2} - \frac{1}{2\sigma} \begin{cases} < \frac{1}{2} \\ \text{b/c } \alpha = 0+ \end{cases}$$

Hence, we proved

$$\begin{aligned} & \mathbb{E} \left[\left| \Psi[u](t, x_1) - \Psi[u](t, x_2) \right|^p \right] \\ & \lesssim |x_1 - x_2| \stackrel{(28)}{=} \mathbb{E} \left[\|u\|_{C_T H_x^s}^{\sigma p} \right] \\ & \quad \text{for } p \text{ by choosing } p \gg 1. \end{aligned}$$

As for the nonlinear part,

$$\| \int_0^t P(t-t') N(u)(t') \|_{C_T C_x^{s_0}}$$

$$\lesssim \left\| \int_0^t \| P(t-t') N(u)(t') \|_{W_x^{s_0, \infty}} dt' \right\|_{C_T}$$

$$\stackrel{\text{Schauder}}{\lesssim} \underbrace{\left\| \int_0^t (t-t')^{-\frac{1}{2} - \frac{s_0}{2}} dt' \right\|_{C_T}}_{\lesssim T^\theta} \underbrace{\| N(u) \|_{C_T L_x^k}}_{= \| u \|_{C_T L_x^k}^k}$$

$$\lesssim \| u \|_{C_T H^s}^k, \quad s \geq \frac{1}{2} - \frac{1}{k} (< \frac{1}{2})$$