

Lec 2 11/01/17 (Wed)

①

• Finite dim'l Hamiltonian dynamics on \mathbb{R}^{2n}

$$\partial_t p_j = \frac{\partial H}{\partial q_j}, \quad \partial_t q_j = -\frac{\partial H}{\partial p_j}, \quad j = 1, \dots, n$$

Vec. field = X , $X_j = \frac{\partial H}{\partial q_j}$

$$X_{n+j} = -\frac{\partial H}{\partial p_j}$$

① • Liouville's thm

$$\begin{aligned} \frac{d}{dt} \text{vol} &= \text{div } X = \sum_{j=1}^n \left[\frac{\partial}{\partial p_j} X_j + \frac{\partial}{\partial q_j} X_{n+j} \right] \\ &= \sum_{j=1}^n \left[\frac{\partial}{\partial p_j} \frac{\partial H}{\partial q_j} + \frac{\partial}{\partial q_j} \left(-\frac{\partial H}{\partial p_j} \right) \right] = 0. \end{aligned}$$

$$\Rightarrow dp dq = \prod_{j=1}^n dp_j dq_j \text{ is } \underline{\text{invariant}}.$$

② Hamiltonian H is conserved.

$$\frac{d}{dt} H(p(t), q(t)) = \frac{\partial H}{\partial p} \frac{\partial p}{\partial t} + \frac{\partial H}{\partial q} \frac{\partial q}{\partial t} = \frac{\partial H}{\partial p} \frac{\partial H}{\partial q} + \frac{\partial H}{\partial q} \left(-\frac{\partial H}{\partial p} \right) = 0$$

Gibbs meas.: $d\mu = Z_\beta^{-1} e^{-\beta H(p, q)} dpdq$, $\beta > 0$ ②

$$Z_\beta = \int_{\mathbb{R}^{2n}} e^{-\beta H(p, q)} dpdq < \infty$$

↑ inverse temp

= partition function

is invariant.

Invariance: $\Phi(t) = (p(0), q(0)) \mapsto (p(t), q(t))$

$$\mu(\Phi(t)A) = \mu(\{(p_0, q_0) \in \Phi(t)A\})$$

$$= \mu(\{\Phi(t)(p_0, q_0) \in A\})$$

$$= Z_\beta^{-1} \int_A e^{-\beta \underbrace{H(p(t), q(t))}_{H(p_0, q_0)}} \underbrace{dp(t) dq(t)}_{= dp_0 dq_0} \overset{\text{"}\Phi(t)_* \text{"}}{(dpdq)}$$

$$= \mu(A)$$

Rmk: Suppose that F is conserved under the dynamics (3)

$$\Rightarrow d\mu_F = Z^{-1} e^{-F(p,q)} dp dq$$

is invariant (for nice F).

Q: Why do we care about invariant meas?

① can view the system as a dynamical system.

with meas-preserving transf $T := \Phi(1)$

$$T: (p(0), q(0)) \mapsto (p(t), q(t)) \big|_{t=1}$$

Poincaré recurrence thm:

(i) \forall meas A with $\mu(A) > 0$.

$\exists N \in \mathbb{N}$ $\mu(A \cap T^{-N}A) > 0$.

(ii) a. e. $(p, q) \in \mathbb{R}^{2n}$ are stable according to Poisson.

i.e. $\exists \{t_j\}_{j \in \mathbb{N}} \rightarrow \infty$

s.t. $(p(t_j), q(t_j)) \rightarrow (p(0), q(0))$



← Zhidkov '01
Lec. Note in Math

Also, Furstenberg's multiple recurrence thm

(4)

$\forall A$ with $\mu(A) > 0$ and $k \in \mathbb{N}$.

$\exists N \in \mathbb{N}$ s.t.

$$\mu(A \cap T^{-N}A \cap T^{-2N}A \cap \dots \cap T^{-kN}A) > 0.$$

(2) μ is "supposed" to describe a "typical" long time behavior of solutions.

$$(NLS) \quad i\partial_t u + \Delta u = \pm |u|^{p-1}u \quad \text{on } \mathbb{T}^d.$$

$$H(u) = \frac{1}{2} \int |\nabla u|^2 \pm \frac{1}{p+1} \int |u|^{p+1}$$

$$(NLS) \Leftrightarrow \partial_t u = -i \frac{\partial H}{\partial \bar{u}}.$$

Symplectic space $L^2(\mathbb{T}^d)$

$$\text{Symplectic form } \omega(f, g) = \text{Im} \int_{\mathbb{T}^d} f(x) \overline{g(x)} dx$$

$$dH|_u(\phi) = \omega(\phi, -i \frac{\partial H}{\partial \bar{u}})$$

$$\uparrow \text{Gateaux derivative} = \frac{d}{d\varepsilon} H(u + \varepsilon\phi) \Big|_{\varepsilon=0}$$

On Fourier side:

(5)

$$p_n = \operatorname{Re} \hat{u}_n, \quad q_n = \operatorname{Im} \hat{u}_n$$

Linear case: $H(p, q) = \frac{1}{2} \sum |m|^2 (p_m^2 + q_m^2)$

↑
Plancherel

Plancherel: $\int_{\mathbb{T}^d} |\nabla u|^2 dx = \sum_{n \in \mathbb{Z}^d} |m|^2 |\hat{u}_n|^2$

Lin Schrödinger eqn: $\partial_t u = i \Delta u$

$$\Leftrightarrow \partial_t \hat{u}_n = -i |m|^2 \hat{u}_n, \quad n \in \mathbb{Z}^d$$

$$\Leftrightarrow \partial_t \begin{pmatrix} p_n \\ q_n \end{pmatrix} = \begin{pmatrix} \frac{\partial H}{\partial q_n} \\ -\frac{\partial H}{\partial p_n} \end{pmatrix}$$

Gibbs meas $d\mu = Z^{-1} e^{-H(u)} du$ (6)

$$= Z^{-1} e^{-\frac{1}{p+1} \int |u|^{p+1}} e^{-\frac{1}{2} \int |u|^2} du$$

Wiener meas on \mathbb{T}^d .

In fact, we consider

$$d\mu = Z^{-1} e^{-H(u) - \frac{1}{2} M(u)} du, \quad M(u) = \int |u|^2 dx$$

in order to avoid a problem at $n=0$.

$$e^{-\frac{1}{p+1} \int |u|^{p+1}} e^{-\frac{1}{2} \|u\|_{H^1}^2} du$$

Gaussian measures on periodic functions/distributions

Consider $d\mu_s = Z^{-1} e^{-\frac{1}{2} \|u\|_{H^s}^2} du, \quad s \in \mathbb{R} \quad \mathcal{L}(\mathbb{T}^d)$

(on \mathbb{R}^n , $d\mu = \frac{1}{(2\pi)^{n/2}} e^{-\frac{|x|^2}{2}} dx$)

as a limit of

$$d\mu_{S,N} = Z_N^{-1} e^{-\frac{1}{2} \|P_{\leq N} u\|_{H^s}^2} d(P_{\leq N} u) \quad (7)$$

$$= Z_N^{-1} e^{-\frac{1}{2} \sum_{|n| \leq N} \langle m \rangle^{2s} |\hat{u}_n|^2} \prod_{|n| \leq N} d\hat{u}_n$$

$\left(P_{\leq N} u = \sum_{|n| \leq N} \hat{u}_n e^{in \cdot x} \right)$

Lebesgue meas
on $\mathbb{C} \cong \mathbb{R}^2$.

$$= Z_N^{-1} \prod_{|n| \leq N} \underbrace{e^{-\frac{1}{2} \langle m \rangle^{2s} |\hat{u}_n|^2} d\hat{u}_n}_{\text{Gaussian meas on } \mathbb{C}}$$

$$e^{-\frac{1}{2}(x^2+y^2)} dx dy$$

Gaussian meas on \mathbb{C}
mean 0, var = $2 \langle m \rangle^{-2s}$

$$c_n e^{-\frac{1}{2} |g_n|^2} dg_n$$

$$g_n = g_n(\omega) = \omega \in \Omega \mapsto g_n(\omega) \in \mathbb{C}$$

\mathbb{C} -valued Gaussian random variable.

$$\Rightarrow \hat{u}_n = \frac{g_n(\omega)}{\langle n \rangle^s}, \quad |n| \leq N$$



(8)

$\{g_n\}_{|n| \leq N}$ = seq of indep standard \mathbb{C} -valued Gaussian r.v.'s.

$$\text{mean} = \mathbb{E}[g_n] = \int_{\Omega} g_n(\omega) dP(\omega) = 0$$

$$\text{variance} = \mathbb{E}[|g_n - \mathbb{E}[g_n]|^2] = \int_{\Omega} |g_n|^2 dP = 2.$$

$$\begin{aligned} &= \int_{\Omega} (\text{Re } g_n)^2 dP + \int_{\Omega} (\text{Im } g_n)^2 dP \\ &= 1 + 1 \end{aligned}$$

$$e^{-\frac{1}{2}|g_n|^2} dg_n = e^{-\frac{1}{2}(\text{Re } g_n)^2} d(\text{Re } g_n) \times e^{-\frac{1}{2}(\text{Im } g_n)^2} d(\text{Im } g_n)$$

Take $N \rightarrow \infty$.

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Problem: We can not take a limit as $N \rightarrow \infty$ in $H^s(\mathbb{T}^d)$.

$$f_{s,N} \sim u^N = \sum_{|n| \leq N} \frac{g_n(\omega)}{\langle n \rangle^s} e^{in \cdot x}$$

Let $\sigma \in \mathbb{R}$, $M > N$

$$\mathbb{E} \left[\|u^M - u^N\|_{H^\sigma(\mathbb{T}^d)}^2 \right]$$

$$= \mathbb{E} \sum_{N < |m| \leq M} \frac{|g_m|^2}{\langle m \rangle^{2s-2\sigma}} = 2 \sum_{N < |m| \leq M} \frac{1}{\langle m \rangle^{2s-2\sigma}} \rightarrow 0$$

iff $2s - 2\sigma > d$

\Leftrightarrow

$$\sigma < s - \frac{d}{2}$$

Moral: topology matters!

Under $\sigma < s - \frac{d}{2}$,

(10)

u^N converges in $L^2(\Omega; H^\sigma(\mathbb{T}^d))$.

$$\text{to } u = \sum_{n \in \mathbb{Z}^d} \frac{g_n}{\langle n \rangle^s} e^{in \cdot x}.$$

• $df_s = Z^{-1} e^{-\frac{1}{2} \|u\|_{H^s}^2} du$ is the induced probability meas under the map

$$\omega \in \Omega \xrightarrow{u} u = \sum_{n \in \mathbb{Z}^d} \frac{g_n}{\langle n \rangle^s} e^{in \cdot x} \in H^\sigma(\mathbb{T}^d)$$

\mathcal{P}

$$f_s = \mathcal{P} \circ u^{-1}$$

Abstract Wiener space: Gross '65, Kuo '75 Lec. Notes in Math

$H = \infty$ -dim'l separable Hilbert space

Consider " $df = Z^{-1} e^{-\frac{1}{2} \|u\|_H^2} du$ "

⇐ NOT countably additive if $\dim H = \infty$

(11)

• Must enlarge the space $H \subset B$
Banach space
dense & conti.

Def: We say (H, B, ρ) is an abstract Wiener space

if
$$\int_B e^{i \langle u, \varphi \rangle_{B^*}} d\rho(u) = e^{-\frac{1}{2} \|\varphi\|_{H^*}^2}$$

for all $\varphi \in B^* \subset H^* = H$

(i.e. $\langle u, \varphi \rangle$ is a stand. Gaussian r.v. $\forall \varphi$.
 \uparrow
random

Another equiv def: $B =$ completion of H under a "measurable"
seminorm $\|\cdot\|_B$

$\forall \varepsilon > 0 \exists P_\varepsilon \in \{ \text{collection of finite dim'l proj} \}$

s.t. $\rho(\|Pu\|_B > \varepsilon) < \varepsilon \quad \forall P \perp P_\varepsilon$

e.g. H^s .

$$P_\varepsilon \sim P_{\leq N}$$

$$P \perp P_\varepsilon \sim P_{N < \cdot \leq M}$$

$$\text{Want } \int (\|P_{>N} u\|_B > \varepsilon) < \varepsilon.$$

ex: $B = W^{\sigma, p}$, $\sigma < s - \frac{d}{2}$

H , Hilbert space

$$" dp = z^{-1} e^{-\frac{1}{2} \|u\|_H^2} du "$$

\Rightarrow enlarge H to $B =$ completion of H under a measurable norm $\|\cdot\|_B$

ex: $H = H^s(\mathbb{T}^d)$

$B = H^\sigma(\mathbb{T}^d)$,

$W^{\sigma,p}(\mathbb{T}^d)$

$\sigma < s - \frac{d}{2}$, $u^\omega(x) = \sum_{n \in \mathbb{Z}^d} \frac{g_n(\omega)}{\langle n \rangle^s} e^{in \cdot x}$

$\|u\|_{W^{\sigma,p}} = \| \mathcal{F}^{-1}(\langle n \rangle^\sigma \hat{u}(n)) \|_{L_x^p(\mathbb{T}^d)}$, $p \leq \infty$

Besov space

$B_{p,q}^\sigma(\mathbb{T}^d)$,

$p \leq \infty, q < \infty$

$\|u\|_{B_{p,q}^\sigma} = \| N^\sigma \| P_N u \|_{L_x^p(\mathbb{T}^d)} \|_{\ell_q^q(\mathbb{Z}_{\geq 0})}$ $\left(P_N u = \sum_{|n| \sim N} \hat{u}_n e^{in \cdot x} \right)$

" N , dyadic"

$N = 2^k, k \geq 0$

$k \in \mathbb{Z}$

$\frac{1}{2}N \leq |n| \leq 2N$

$\bullet B_{2,2}^\sigma = H^\sigma$

• Fourier - Lebesgue space: $\mathcal{FL}^{\sigma, p}(\mathbb{T}^d)$

(2)

$$\|u\|_{\mathcal{FL}^{\sigma, p}(\mathbb{T}^d)} = \|\langle m \rangle^{\sigma} \hat{u}_n\|_{\ell_m^p(\mathbb{Z}^d)}$$

• $\mathcal{FL}^{\sigma, 2} = H^{\sigma}$.

$(\sigma - s)p < -d$

• $\mathbb{E}[\|u^{\omega}\|_{\mathcal{FL}^{\sigma, p}}^p] = \mathbb{E} \sum_{n \in \mathbb{Z}^d} \langle m \rangle^{p(\sigma - s)} |g_n|^p$

$\sim \sum_{n \in \mathbb{Z}^d} \langle m \rangle^{p(\sigma - s)} < \infty$ iff $(\sigma - s)p < -d$.

Lemma 1.1 (Tail estimate)

Let $\sigma < s - \frac{d}{2}$. Then,

$$P_s(\|u\|_{H^{\sigma}} > K) \leq C e^{-cK^2}$$

for all $K > 0$.

Rmk: This follows from Fernique's integrability theorem:

$$\int_B e^{c\|u\|_B^2} dP(u) < \infty \text{ for some } c > 0.$$

$$\Leftrightarrow P(\|u\|_B > K) \leq c e^{-cK^2}$$

(3)

Pf of Lemma 1.1: By Chebyshev's inequality $\sum_n \langle m \rangle^{2\sigma} |\hat{u}_n|^2$

$$e^{cK^2} P_S(B_K^c) \leq \int_{H^\sigma} e^{c\|u\|_{H^\sigma}^2} dP_S(u)$$

$$\begin{aligned} \Leftrightarrow P_S(\|u\|_{H^\sigma} > K) &= P_S(e^{c\|u\|_{H^\sigma}^2} > e^{cK^2}) \\ &= \int_{H^\sigma} \mathbb{1}_{\frac{e^{c\|u\|_{H^\sigma}^2}}{e^{cK^2}} > 1} dP_S(u) \\ &< \frac{1}{e^{cK^2}} \text{ (RHS)} \end{aligned}$$

usual Chebyshev: $P(f(\omega) > K) \leq \frac{\mathbb{E}[|f|^2]}{K^2}$

$$= \prod_{m \in \mathbb{Z}^d} \int_{\mathbb{C}} e^{c \langle m \rangle^{2\sigma-2s} |g_n|^2} e^{-\frac{1}{2}|g_n|^2} \frac{dg_n}{2\pi}$$

$$= \prod_{n \in \mathbb{Z}^d} \frac{1}{1 - 2c \langle m \rangle^{2\sigma-2s}}$$

$$\left(\mathbb{E} \left[e^{aX^2} \right] = \int e^{ax^2} \frac{e^{-\frac{1}{2}x^2}}{\sqrt{2\pi}} dx = \frac{1}{\sqrt{1-2a}}, \quad a < \frac{1}{2} \right. \quad (4)$$

$$X \sim \underbrace{N_{\mathbb{R}}(0, 1)}$$

real-valued Gaussian r.v.

mean 0, var 1.

$$y = \sqrt{1-2a} x$$

$$= \prod_{n \in \mathbb{Z}^d} \left(1 + \frac{2c \langle M \rangle^{2\sigma-2s}}{1 - 2c \langle M \rangle^{2\sigma-2s}} \right) < \infty$$

$$\text{iff } \sigma < s - \frac{d}{2}$$

$$\left(a_n > 0. \text{ Then } \prod (1 + a_n) < \infty \iff \sum a_n < \infty \right.$$

□

• Construction of Gibbs measure on Π

(5)

$d=1$: Recall

$$d\mu = Z^{-1} e^{-H(u) - \frac{1}{2}M(u)} du$$

$$H(u) = \frac{1}{2} \int |\partial_x u|^2 + \frac{1}{p+1} \int |u|^{p+1}$$

$$M(u) = \int |u|^2$$

$$= Z^{-1} e^{\frac{1}{p+1} \int |u|^{p+1}} \underbrace{e^{-\frac{1}{2} \|u\|_{H^1}^2}}_{df_1} du$$

$$= Z^{-1} e^{\frac{1}{p+1} \int |u|^{p+1}} df_1 \quad \text{on } H^\sigma(\Pi), \quad \sigma < 1 - \frac{1}{2} = \frac{1}{2}$$

• Defocusing case: By Sobolev ineq,

$$\int |u|^{p+1} = \|u\|_{L^{p+1}}^{p+1} \lesssim \|u\|_{H^{\frac{1}{2}}}^{p+1} < \infty, \text{ a.s.}$$

$$\Rightarrow 0 < e^{-\frac{1}{p+1} \int |u|^{p+1}} \leq 1, \text{ a.s.}$$

$$\Rightarrow \mu \text{ is a prob. meas on } H^\sigma(\Pi), \quad \sigma < \frac{1}{2}.$$

Sobolev ineq: On \mathbb{R}^d or \mathbb{T}^d ,

(6)

• \mathbb{R}^d : $\|u\|_{L^q(\mathbb{R}^d)} \lesssim \|u\|_{\dot{W}^{s,p}(\mathbb{R}^d)}$

homogeneous Sobolev space

use $|\xi|^s$

instead of $\langle \xi \rangle^s$

When $\frac{s}{d} = \frac{1}{p} - \frac{1}{q}$

• \mathbb{T}^d : $\|u\|_{L^q(\mathbb{T}^d)} \lesssim \|u\|_{W^{s,p}(\mathbb{T}^d)}$

When $\frac{s}{d} \geq \frac{1}{p} - \frac{1}{q}$.

ex: $d=1$: $\|u\|_{L^q(\mathbb{T})} \lesssim \|u\|_{H^{\frac{1}{2}-\varepsilon}(\mathbb{T})}$
by choosing $\varepsilon = \varepsilon(q) > 0$ suff small.

• Focusing case: let $p > 1$.

(7)

$$\int_{\mathbb{T}} |u|^{p+1} \geq \|u\|_{L^2}^{p+1} = \left(\sum |\hat{u}_n|^2 \right)^{\frac{p+1}{2}}$$

$$\geq \sum \left| \frac{g_n}{\langle m \rangle} \right|^{p+1} \quad l^2 \subset l^{p+1}$$

$$\Rightarrow \mathbb{E}_p \left[e^{\frac{1}{p+1} \int |u|^{p+1}} \right] \geq \prod_{n \in \mathbb{Z}} \mathbb{E} \left[e^{\left| \frac{g_n}{\langle m \rangle} \right|^{p+1}} \right] = \infty \quad \text{b/c } p+1 > 2.$$

\Rightarrow can not construct μ as it is

Idea: Mass is conserved

\Rightarrow Introduce a mass cutoff.

$$"d\mu = z^{-1} \mathbb{1}_{\{\|u\|_{L^2} \leq r\}} \underbrace{e^{\frac{1}{p+1} \int |u|^{p+1}}}_{= e^{-H(u) - \frac{1}{2} M(u)}} d\rho_1"$$

Prop 1.2 (Lebowitz - Rose - Speer '88, Bourgain '94)

(8)

$p > 1$

(i) $p < 5$

(*)
$$R(u) = R_r(u) := \mathbb{1}_{\{\|u\|_{L^2} \leq r\}} e^{\frac{1}{p+1} \int |u|^{p+1}} \in L^q(d\beta_r)$$

 $\forall r, q < \infty$

(ii) $p=5$: (*) holds if r is sufficiently small.

In fact, $r < \|Q\|_{L^2(\mathbb{R})}$
 \uparrow
ground state on \mathbb{R}

$p=5$: mass-critical NLS $i\partial_t u + \partial_x^2 u = -|u|^4 u$ on \mathbb{T}

\Leftarrow lowest power for which we have a finite time blowup solution.

$M(Q) \leftarrow$ smallest mass for a finite time blowup solution

Lemma 1.3: $\{\tilde{g}_n\}$, indep standard \mathbb{R} -valued Gaussian r.v.'s

Then, $P \left[\left(\sum_{n=1}^M \tilde{g}_n^2 \right)^{1/2} \geq R \right] \leq e^{-\frac{1}{4}R^2}$, $R \geq 3M^{1/2}$
 $M \geq 1$.

Pf:

$$\begin{aligned} \text{(LHS)} &\stackrel{\text{Chebyshev}}{\leq} e^{-tR^2} \mathbb{E} \left[e^{t \sum_{n=1}^M \tilde{g}_n^2} \right] \\ &= (1-2t)^{-M/2} e^{-tR^2} = \prod_{n=1}^M \mathbb{E} \left[e^{t \tilde{g}_n^2} \right] \end{aligned}$$

Choose $t = \frac{1}{2} \left(1 - \frac{M}{R^2} \right)$.

$$\begin{aligned} \text{(LHS)} &\leq \left(\frac{R^2}{M} \right)^{M/2} e^{-\frac{1}{2}R^2 + \frac{1}{2}M} \\ &\leq e^{\frac{M}{2} \ln \frac{R^2}{M} + \left(\frac{1}{18} - \frac{1}{2} \right) R^2} \leq e^{-\frac{1}{4}R^2} \end{aligned}$$

($\ln x \leq \frac{x}{4}$ for $x \geq 9$)

□

Pf of Prop 1.2:

(2)

By Bernstein's ineq

$$\|P_{\leq M_0} u\|_{L^{p+1}} \leq c M_0^{\frac{1}{2} - \frac{1}{p+1}} \underbrace{\|P_{\leq M_0} u\|_{L^2}}_{\leq r}$$

$$\left(\begin{array}{l} \frac{s}{1} \geq \frac{1}{2} - \frac{1}{p+1} \\ \text{(LHS)} \underset{\text{Sobolev}}{\lesssim} \|\langle \partial_x \rangle^{\frac{1}{2} - \frac{1}{p+1}} P_{\leq M_0} u\|_{L^2} \\ \qquad \qquad \qquad \searrow \\ \qquad \qquad \qquad \langle m \rangle^{\frac{1}{2} - \frac{1}{p}} \lesssim M_0^{\frac{1}{2} - \frac{1}{p}} \end{array} \right)$$

Given $\lambda > 1$, choose M_0 s.t.

$$\frac{1}{2} \lambda = c M_0^{\frac{1}{2} - \frac{1}{p+1}} r$$

$$\text{i.e. } M_0 \sim \left(\frac{\lambda}{r}\right)^{\frac{1}{\frac{1}{2} - \frac{1}{p+1}}}$$

Let $q = p+1$.

Let $M_j = 2^j M_0$. Then, we have

$$\| P_{M_j} u \|_{L^q} \leq c M_j^{\frac{1}{2} - \frac{1}{q}} \| P_{M_j} u \|_{L^2}$$

Let $\{\sigma_j\}$ s.t. $\sum_{j \geq 1} \sigma_j = \frac{1}{2}$

(\Leftarrow set $\sigma_j = c 2^{-\epsilon j} = c M_0^\epsilon M_j^{-\epsilon}$)

\Rightarrow By dyadic pigeon hole principle,

$$P_1 (\| u \|_{L^q} > \lambda, \| u \|_{L^2} \leq r)$$

$$\leq \sum_{j=1}^{\infty} P_1 (\| P_{M_j} u \|_{L^q} > \sigma_j \lambda) \quad \text{Recall } \| P_{\leq M_0} u \|_{L^q} \leq \frac{1}{2} \lambda$$

$$\sum_{j=1}^{\infty} \sigma_j \lambda = \frac{1}{2} \lambda < \| P_{> M_0} u \|_{L^q} \leq \sum_{j=1}^{\infty} \| P_{M_j} u \|_{L^q}$$

$$\leq \sum_{j=1}^{\infty} P_1 (\| P_{M_j} u \|_{L^2} > c \sigma_j M_j^{\frac{1}{q} - \frac{1}{2}} \lambda)$$

$$= \left(\sum_{|n| \sim M_j} \frac{|g_n|^2}{\langle m \rangle^2} \right)^{1/2} \sim M_j^{-1} \left(\sum_{|n| \sim M_j} |g_n|^2 \right)^{1/2}$$

(4)

$$P \left(\left(\sum_{|n| \sim M_j} |g_n(\omega)|^2 \right)^{1/2} \geq \underbrace{\sigma_j M_j^{\frac{1}{q} + \frac{1}{2}} \lambda}_{=: R_j} \right) \lesssim M_0^2 M_j^{\frac{1}{2} + \frac{1}{q} - \varepsilon} \Rightarrow M_j^{\frac{1}{2} +}$$

⇒ Apply Lemma 1.3.

$$P_1 \left(\|u\|_{L^q} > \lambda, \|u\|_{L^2} \leq r \right)$$

$$\leq \sum_{j=1}^{\infty} e^{-c R_j^2} \quad R_j^2 \sim M_0^{\frac{2}{q} + 1} \lambda^2 \underline{\underline{2^{(\frac{2}{q} + 1 - 2\varepsilon)j}}}}$$

$$\sim e^{-c \lambda^2 M_0^{\frac{2}{q} + 1}} \quad M_0 \sim \left(\frac{\lambda}{r} \right)^{\frac{1}{\frac{1}{2} - \frac{1}{q}}}$$

$$\sim e^{-c \lambda^{\frac{4q}{q-2}} r^{-\frac{2q+4}{q-2}}}$$

$$\frac{4q}{q-2} > q \text{ for } q < 6.$$

$$\lesssim \begin{cases} e^{-c \lambda^{q+\delta}}, & q < 6 \Leftrightarrow p < 5. \\ e^{-c \lambda^6}, & q = 6 \Leftrightarrow p = 5 \end{cases}$$

with $C \gg 1$ by taking $r \ll 1$

$$\begin{aligned} \|f\|_{L^p}^p &= \int |f(x)|^p dx \\ &= p \int_0^\infty \alpha^{p-1} \mu(|f(x)| > \alpha) d\alpha \end{aligned}$$

(X, μ) meas space

$$\| \mathbb{1}_{\{\|u\|_2 \leq r\}} \|_{L^{\tilde{q}}(dP_i)} \sim e^{\frac{1}{p+1} \int |u|^{p+1} dx}$$

$$\sim 1 + \int_1^\infty e^{-c \lambda^{p+1}} d\lambda$$

$e^{-c \lambda^{p+1+\delta}}$
 or
 $e^{-c \lambda^{p+1}}$

$d\lambda$

$\lambda^{p+1} \sim \ln \alpha$

$p < 5$

$p = 5$

□

Rmk: We went over the harmonic analytic proof
by Bourgain '94.

(6)

The proof by LRS '88 is more probabilistic.

$$R_r(u) = \mathbb{1}_{\{\|u\|_{L^2} \leq r\}} e^{\frac{1}{p+1} \int |u|^{p+1}}$$

$$R_{N,r}(u) = \mathbb{1}_{\{\|P_{\leq N} u\|_{L^2} \leq r\}} e^{\frac{1}{p+1} \int |P_{\leq N} u|^{p+1}}$$

Cor 1.4: $\forall q < \infty$

$$R_{N,r}(u) \longrightarrow R_r(u) \text{ in } L^q(d\rho_1) \\ \text{as } N \rightarrow \infty$$

In particular, with $d\mu_N = Z_N^{-1} R_{N,r}(u) d\rho_1$,
 μ_N converges "uniformly" to μ
i.e. $\forall \varepsilon > 0$, $\exists N_0 \in \mathbb{N}$ s.t.

$$|\mu_N(A) - \mu(A)| < \varepsilon, \quad \forall \text{ measurable } A \\ \forall N \geq N_0$$

(7)

⇐ stronger than weak convergence:

$$\lim_{N \rightarrow \infty} \int f(u) d\mu_N(u) = \int f(u) d\mu(u)$$

for any $f \in C_b(H^\sigma(\mathbb{T}))$, $\sigma < \frac{1}{2}$

Pf: $\int |P_{\leq N} u|^{p+1} \rightarrow \int |u|^{p+1}$ a.s.

(⇐ $\|P_{>N} u\|_{L^{p+1}} \lesssim \|P_{>N} u\|_{H^{\frac{1}{2}-}} \rightarrow 0$.)

⇒ $R_{N,r}(u) \rightarrow R_r(u)$ a.s.

⇒ By Egoroff's thm,

$R_{N,r}(u) \rightarrow R_r(u)$ almost uniformly

⇒ in meas

• Given $\varepsilon > 0$,

let $A_{N,\varepsilon} = \{ u \in H^{\frac{1}{2}-}(\mathbb{T}) : |R_{N,r}(u) - R_r(u)| \leq \frac{1}{2} \varepsilon \}$.

Then, $P_1(A_{N,\varepsilon}^c) \rightarrow 0$ as $N \rightarrow \infty$

(8)

$$\| R_{N,r} - R_r \|_{L^q(dP_i)}$$

$$\leq \| (R_{N,r} - R_r) \mathbb{1}_{A_{N,\varepsilon}} \|_{L^q}$$

$$+ \| (R_{N,r} - R_r) \mathbb{1}_{A_{N,\varepsilon}^c} \|_{L^q}$$

$$\leq \frac{1}{2} \varepsilon + \underbrace{\left(\| R_{N,r} \|_{L^{2q}} + \| R_r \|_{L^{2q}} \right)}_{\leq C < \infty} \underbrace{\left\{ P_i(A_{N,\varepsilon}^c) \right\}^{\frac{1}{2q}}}_{\rightarrow 0}$$

$< \varepsilon$

for $N \gg 1$. \square

• Digression: Proof of Fernique's integrability theorem

(9)

(H, B, ρ) , abstract Wiener space.

$$d\rho = Z^{-1} e^{-\frac{1}{2}\|x\|_H^2} dx$$

Define Wiener measure with variance t .

$$d\rho_t = Z_t^{-1} e^{-\frac{\|x\|^2}{2t}} dx, \quad t > 0$$

• $\{\rho_t\}_{t>0}$ forms a contraction semigroup in the Banach space of bounded, unif conti functions on B . $\rightarrow \|\rho_t\| \leq 1$

Riesz rep thm: X , LCH.

$$(C_0(X))^* = M(X)$$

\uparrow Radon meas

vanishing at ∞

$\forall \varepsilon > 0. \{x: |f(x)| \geq \varepsilon\}$ is cpt.

$$C_0(X) = \overline{C_c(X)}^{\|\cdot\|_{\text{unif}}}$$

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(1)

$$dP_t = Z_t^{-1} e^{-\|x\|^2/2t} dx, \quad t > 0 \quad (H, B, P)$$

• Integrability of $e^{\alpha\|x\|^2}$

Ω = space of conti func ω on $[0, \infty)$
with values in B and $\omega(0) = 0$.

$\exists!$ prob meas P on the σ -field generated by the coordinate functions $\omega \mapsto \omega(t), t > 0$, s.t.

if $0 = t_0 < t_1 < \dots < t_n$, then

① $\omega(t_j) - \omega(t_{j-1}), j = 1, \dots, n$, are indep

② $\omega(t_j) - \omega(t_{j-1})$ is distributed in B

according to $P_{t_j - t_{j-1}}$.

• Define $W(t) : \Omega \rightarrow B$ by $W(t)(\omega) = \omega(t)$

↑ Wiener process in B (starting at 0)

Note: $\int_B e^{\alpha \|x\|^2} P_i(dx) = \mathbb{E}[e^{\alpha \|W(1)\|^2}]$

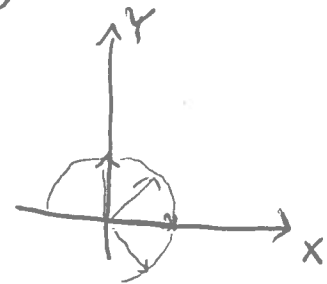
(2)

Pf of Fernique's Thm:

Let $X = W(1)$
 $Y = W(2) - W(1)$ indep, distributed by P_i

Moreover, $\frac{X+Y}{\sqrt{2}}$ and $\frac{X-Y}{\sqrt{2}}$ are indep & distributed by P_i

(just rotate the coord axes in the XY plane.)



For $t > s$, we have

$$P(\|W(1)\| \leq s) P(\|W(1)\| > t)$$

$$= P\left(\left\|\frac{X-Y}{\sqrt{2}}\right\| \leq s\right) P\left(\left\|\frac{X+Y}{\sqrt{2}}\right\| > t\right)$$

$$\stackrel{\text{indep}}{=} P\left(\frac{\|X-Y\|}{\sqrt{2}} \leq s \text{ and } \frac{\|X+Y\|}{\sqrt{2}} > t\right)$$

$$\stackrel{\text{triangle ineq}}{\leq} P\left(|\|X\| - \|Y\|| \leq \sqrt{2}s \text{ and } \|X\| + \|Y\| > \sqrt{2}t\right)$$

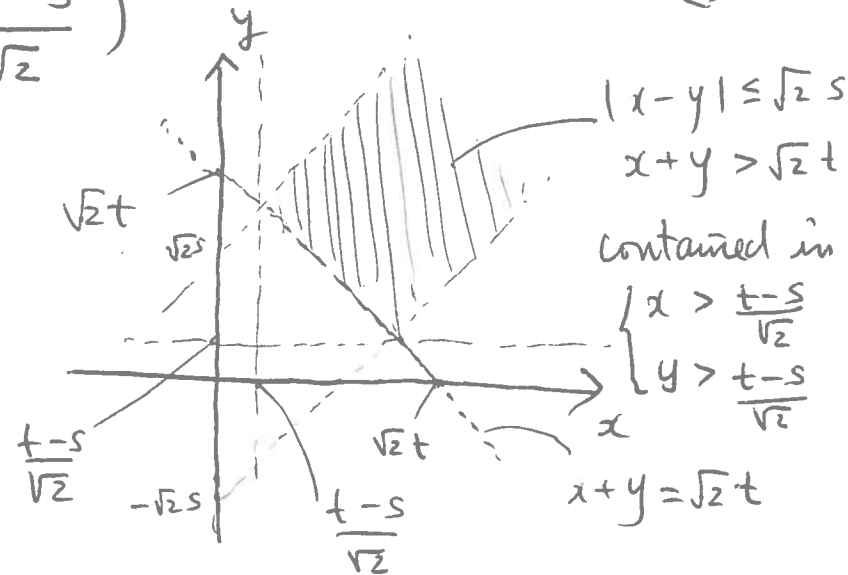
(*)

$$\leq P\left(\|X\| > \frac{t-s}{\sqrt{2}} \text{ and } \|Y\| > \frac{t-s}{\sqrt{2}}\right)$$

indep

$$= P\left(\|W(t)\| > \frac{t-s}{\sqrt{2}}\right)^2$$

(3)



Define t_n by $t_0 = s > 0$

$$t_{n+1} = s + \sqrt{2} t_n$$

$$\Rightarrow t_n = (1 + \sqrt{2} + \dots + \sqrt{2}^n) s$$

$$= \frac{\sqrt{2}^{n+1} - 1}{\sqrt{2} - 1} s.$$

Let $\alpha_m = \frac{P(\|W(t)\| > t_m)}{P(\|W(t)\| \leq s)}$, $m \geq 0$

By (X)

$$\Rightarrow \alpha_{m+1} \leq \alpha_m^2 \Rightarrow \alpha_m \leq \alpha_0^{2^m} = e^{2^m \log \alpha_0}$$

$$\Rightarrow P(\|W(t)\| > t_n) \leq P(\|W(t)\| \leq s) e^{2^n \log \alpha_0}$$

Let $u_n = \sqrt{2}^{n+4} s$

$$\Rightarrow u_n > (\sqrt{2}^{n+1} - 1)(\sqrt{2} + 1) s = t_n$$

$$\Rightarrow P(\|W(u)\| > u_n) \leq P(\|W(u)\| \leq s) e^{\frac{u_n}{16s^2} \log d_0} \quad (4)$$

• Choose $s \gg 1$ s.t. $\frac{u_n}{16s^2} \log d_0 < 1$.

(By conti from above, $P(\|W(u)\| > s) \rightarrow 0$ as $s \rightarrow \infty$.

(By conti from below, $P(\|W(u)\| \leq s) \rightarrow 1$ as $s \rightarrow \infty$

let $a = -\frac{1}{16s^2} \log d_0 > 0$

$$b = P(\|W(u)\| \leq s)$$

$$\Rightarrow P(\|W(u)\| > u_n) \leq b e^{-a u_n^2}$$

$$\Rightarrow \int_{\|x\| > N} e^{\alpha \|x\|^2} \rho_i(dx) \leq \sum_{k=0}^{\infty} \int_{2^k N \leq \|x\| \leq 2^{k+1} N} e^{\alpha \|x\|^2} f_i(dx)$$

$$\leq b \sum_{k=0}^{\infty} e^{\frac{\alpha (2^{k+1} N)^2}{(4d-a)(2^k N)^2}} e^{-\frac{a(2^k N)^2}{(4d-a)(2^k N)^2}} < \infty$$

by choosing $\alpha \ll a$.

□

• Invariance of the Gibbs measure ($d=1$)

(5)

• 3 scenarios

CASE 1: We have an a priori (deterministic)
global well-posedness in $\text{supp}(\mu) \subset H^{\frac{1}{2}^-}(\mathbb{T})$

ex: KdV, cubic, GWP in $L^2(\mathbb{T})$ (Bourgain'93)

\Rightarrow Invariance of μ follows from the invariance of
the "finite-dim'l" Gibbs meas μ_N associated to
the truncated equation. (easy)

CASE 2: We only know a priori local well-posedness in $\text{supp}(\mu)$
(in subcritical sense).

\downarrow
local existence time $\delta > 0$ depends only on the size
of initial data: $\delta \sim \|u_0\|_{H^\sigma}^{-\theta}$ for some $\theta > 0$.

$$\underline{\text{NLS}}: \quad i \partial_t u + \partial_x^2 u = \pm |u|^{p-1} u$$

(6)

We say u is a solution to NLS if u satisfies the following Duhamel formulation:

$$u(t) = S(t) u_0 \mp i \int_0^t S(t-t') |u|^{p-1} u(t') dt'$$

Note: $i \partial_t u + \partial_x^2 u = F$

$$\Rightarrow \text{F.T. in } x \quad i \partial_t \hat{u}(m) - m^2 \hat{u}(m) = \hat{F}(m) \quad \times e^{itm^2}$$

$$\Rightarrow i \partial_t (e^{itm^2} \hat{u}(m)) = e^{itm^2} \hat{F}(m)$$

\Rightarrow integrate from 0 to t . & invert F.T.

$$\Rightarrow u(t) = S(t) u_0 - i \int_0^t S(t-t') F(t') dt'$$

Let $\Gamma_{u_0}(u)(t) = S(t) u_0 \mp i \int_0^t S(t-t') |u|^{p-1} u(t') dt'$

u , soln to NLS $\iff \Gamma_{u_0}(u) = u$. i.e. fixed pt for Γ .

Suppose that we have the following 2 estimates:

(7)

① Linear estimate:

$$\| \Gamma u \|_{X^s(\tau_0, \delta)} \lesssim \| u_0 \|_{H^s} + \| |u|^{p-1} u \|_{N^s}$$

② Nonlinear estimate:

$$\| |u|^{p-1} u \|_{N^s} \lesssim \delta^\theta \| u \|_{X^s}^p$$

$$\Rightarrow \| \Gamma u \|_{X^s(\tau_0, \delta)} \leq C_1 \| u_0 \|_{H^s} + C_2 \delta^\theta \| u \|_{X^s(\tau_0, \delta)}^p$$

$\Rightarrow \Gamma$ is a contraction on $B_R \subset X^s(\tau_0, \delta)$

$$\text{Choose } R = 2C_1 \| u_0 \|_{H^s}$$

$$\cdot \| \Gamma u \|_{X^s(\tau_0, \delta)} \leq \frac{1}{2} R + C_2 \delta^\theta R^p \leq R \text{ for } \delta = \delta(R) \ll 1$$

($\Leftarrow C_2 \delta^\theta R^{p-1} \leq \frac{1}{2}$)

$$\cdot \| \Gamma u - \Gamma v \|_{X^s(\tau_0, \delta)} \leq \frac{1}{2} \| u - v \|_{X^s(\tau_0, \delta)} \text{ for } \delta = \delta(R) \ll 1$$

$$\Rightarrow \exists! u \in B_R \quad \text{s.t.} \quad \Gamma_{u_0}(u) = u.$$

(8)

—
• examples of X^s :

Strichartz space: $\| \langle \nabla \rangle^s u \|_{L_t^q L_x^r(\mathbb{I}_0, \delta) \times M}$

Strichartz estimate for Schrödinger eqn on \mathbb{R}^d :

$$\| S(t) u_0 \|_{L_t^q L_x^r} \lesssim \| u_0 \|_{L_x^2(\mathbb{R}^d)}$$

where $q, r \geq 2$ satisfies

$$\underline{\frac{2}{q} + \frac{d}{r} = \frac{d}{2}}, \quad (q, r, d) \neq (2, \infty, 2)$$

(q, r) is (Schrödinger) admissible

- $X^{s,b}$ - space (Fourier restriction norm method by Bourgain '93)

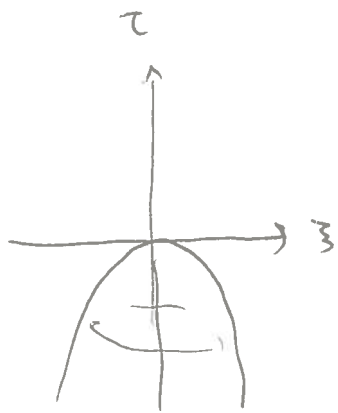
(9)

$$\|u\|_{X^{s,b}} = \left\| \langle \xi \rangle^s \langle \tau + |\xi|^2 \rangle^b \hat{u}(\tau, \xi) \right\|_{L^2_{\xi, \tau}}$$

linear Sch. eqn: $i \partial_t u + \Delta u = 0$

$$\Rightarrow \text{space-time F.T.} \quad -(\tau + |\xi|^2) \hat{u}(\tau, \xi) = 0.$$

$\Rightarrow \hat{u}(\tau, \xi) = \int_{t,x} (S(t) u_0)(\tau, \xi)$ is a measure supported on the paraboloid $\tau = -|\xi|^2$.



"measures" how far/close u is to being a linear soln.

When $b > 0$, $\langle \tau + |\xi|^2 \rangle^b$ penalizes functions away from linear solutions

\Rightarrow suitable for study Γu in a perturbative manner.

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①

• LWP in a subcritical sense:

$$\begin{array}{l} \text{Linear estimate} \\ \text{nonlin estimate} \end{array} \Rightarrow \|\Gamma_{u_0} u\|_{X^s([0, \delta])} \lesssim \|u_0\|_{H^s} + \delta^\theta \|u\|_{X^s([0, \delta])}^p$$

⇒ By a fixed pt argument (Banach's contraction mapping thm)

⇒ ∃! soln u with $u|_{t=0} = u_0$ on $[0, \delta)$

$$\delta \sim (1 + \|u_0\|_{H^s})^{-\gamma}, \quad \gamma > 0.$$

• Blowup alternative: let T^* be the forward maximal time of existence.
Then, $[0, T^*)$

$$\text{either } T^* = \infty \text{ or } \lim_{t \uparrow T^*} \|u(t)\|_{H^s} = \infty$$

• The "only" way to construct global-in-time solns is to use conservation laws: (defocusing NLS)

$$H(u) = \frac{1}{2} \int |\nabla u|^2 + \frac{1}{p+1} \int |u|^{p+1}, \quad M(u) = \int |u|^2 dx$$

$$\Rightarrow \|u(t)\|_{H^1}^2 \lesssim H(u)(t) + M(u)(t) = H(u_0) + M(u_0) < \infty \quad (2)$$

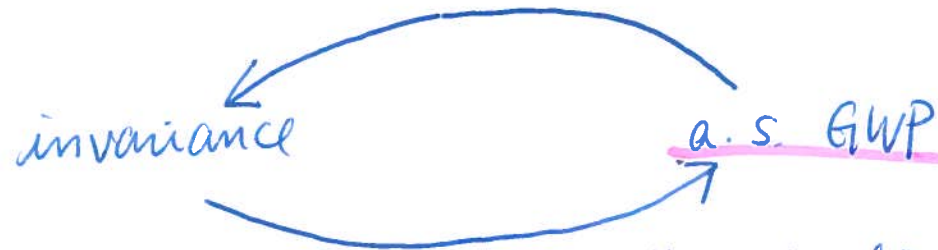
Namely,

LWP in H^1 in a subcritical sense

\Rightarrow GWP in H^1 .

Issue: \nexists conservation law at the level of the Gibbs meas μ
 $s = \frac{1}{2}^-$.

idea: use invariance of μ (in place of a conservation law)
 to construct global-in-time dynamics (on $\text{supp } \mu$)



Bourgain '94: use invariance of the "finite dim'l" Gibbs meas μ_n
 associated to the truncated dynamics

\Rightarrow a.s. GWP \Rightarrow invariance

CASE 3: No LWP (by a deterministic method)

(3)

(3.a): probabilistic local well-posedness

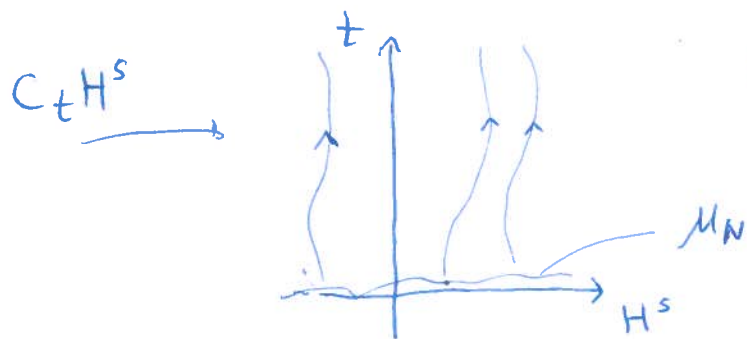
McKean '95, Bourgain '96, Burq-Tzvetkov '07.

Recall $\mu \ll \rho_1 \rightsquigarrow u_0^\omega(x) = \sum_{n \in \mathbb{Z}} \frac{g_n(\omega)}{\langle n \rangle} e^{inx}$

Aim: Exploit the randomness of initial data.

(3.b): compactness argument

(for measures on space-time functions)



$$\Phi_N = u_{0,N} \mapsto u_N$$

Define a prob meas ν_N by

$$\nu_N = \mu_N \circ \Phi_N^{-1}$$

↑

prob meas on space-time functions

⇒ a.s. global existence (without uniqueness)

& "invariance" (in some mild sense)

We first focus on CASE 2.

$$(NLS) \quad i \partial_t u + \partial_x^2 u = \pm |u|^{p-1} u \quad \text{on } \mathbb{T}.$$

Bourgain '93: LWP in $H^s(\mathbb{T})$ for some $s = s(p) < \frac{1}{2}$ in a subcritical sense

$$\delta \sim (1 + \|u_0\|_{H^s})^{-\delta}$$

Consider the following truncated dynamics:

$$(FNLS) \quad \begin{cases} i \partial_t u_N + \partial_x^2 u_N = \pm P_{\leq N} (|P_{\leq N} u_N|^{p-1} P_{\leq N} u_N) \\ u_N|_{t=0} = u_0 \end{cases}$$

$$P_{\leq N} f(x) = \sum_{|n| \leq N} \hat{f}(n) e^{inx}$$

(FNLS) is NOT finite dim'l.

high freq \sim linear \swarrow
low freq \sim nonlinear \swarrow decoupled

(FNLS) is locally well-posed by apply the LWP argument for (NLS):

$$\delta \sim (1 + \|u_0\|_{H^s})^{-\delta}, \quad \underline{\text{indep of } N}.$$

$$\text{Let } H_N(u_N) = \frac{1}{2} \int |\partial_x u_N|^2 \pm \frac{1}{p+1} \int |P_{\leq N} u|^{p+1} \quad (5)$$

$$\Rightarrow \partial_t u_N = -i \frac{\partial H_N}{\partial \bar{u}_N} \quad \text{lin. dynamics on high freq.}$$

① $u_{\text{high}} = P_{>N} u_N$ evolves linearly.

$$\partial_t \hat{u}_{\text{high}}(n) = -i n^2 \hat{u}_{\text{high}}(n), \quad |n| > N$$

$$\Rightarrow \hat{u}_{\text{high}}(t, n) = e^{-itn^2} \hat{u}_0(n), \quad |n| > N$$

In particular, u_{high} exists globally.

② $u_{\text{low}} = P_{\leq N} u_N$ satisfies

$$i \partial_t u_{\text{low}} + \partial_x^2 u_{\text{low}} = \pm P_{\leq N} (|u_{\text{low}}|^{p-1} u_{\text{low}})$$

\Leftarrow finite dim'l system of (nonlin) ODEs (on the Fourier side)
(with Lipschitz v.f.) \Rightarrow local existence by Cauchy-Lipschitz thm.

• L^2 -norm: $\|u_{\text{low}}\|_{L^2} = \left(\sum_{|n| \leq N} |\hat{u}_{\text{low}}(n)|^2 \right)^{1/2}$ is conserved.
= Euclidean distance on \mathbb{C}^{2N+1} .

$\Rightarrow U_{low}$ exists globally.

(6)

$\Rightarrow U_N = U_{low} + U_{high}$ exists globally in time.

Issue: (FNLS) is GWP for each $N \in \mathbb{N}$,

but there is NO uniform (in N) control on $\|U_N(t)\|_{H^s}$.

Write $P_1 = P_N \otimes P_N^\perp$

$$dP_N = Z_N^{-1} e^{-\frac{1}{2} \|P_{\leq N} u\|_{H^1}^2} d(P_{\leq N} u)$$

$$= Z_N^{-1} \prod_{|m| \leq N} e^{-\frac{1}{2} \langle n \rangle^2 |\hat{u}(n)|^2} d\hat{u}(n)$$

$$dP_N^\perp = \tilde{Z}_N^{-1} e^{-\frac{1}{2} \|P_{>N} u\|_{H^1}^2} d(P_{>N} u) \quad \text{on } H^s, s < \frac{1}{2}$$

Then,

① P_N^\perp is invariant under (FNLS)_{high}.

$$P_{>N} U_0(x) = \sum_{|n| > N} \frac{g_n(\omega)}{\langle n \rangle} e^{inx} \Rightarrow U_{high}(t, x) = \sum_{|n| > N} \frac{e^{-itn^2} g_n(\omega)}{\langle n \rangle} e^{inx}$$

\Rightarrow Since ρ_N is invariant under a rotation,
 ρ_N^\perp is invariant.

(7)

$$(2) \quad d\mu_{N, \text{low}} = Z_N^{-1} R_{N, r}(u_{\text{low}}) d\rho_N$$

is invariant under (FNLS_{low}). $\mathbb{1}_{\{\|u_{\text{low}}\|_{L^2} \leq r\}} e^{\frac{1}{p+1} |u_{\text{low}}|^{p+1}}$

$$d\mu_{N, \text{low}} = Z_N^{-1} e^{-\underbrace{H_N(u_{\text{low}})}_{\text{conserved}}} \underbrace{d u_{\text{low}}}_{\text{inv by Liouville's thm}}$$

conserved. inv by Liouville's thm

$\Rightarrow \mu_N = \mu_{N, \text{low}} \otimes \rho_N^\perp$ is invariant under (FNLS)

Key Proposition (Bourgain '94)

$\forall T > 0, \varepsilon > 0, \exists \Omega_N = \Omega_N(T, \varepsilon)$ s.t.

(i) $\mu_N(\Omega_N^c) < \varepsilon$

(ii) For $u_0 \in \Omega_N$, the soln u_N to (FNLS) with $u_N|_{t=0} = u_0$
satisfies $\|u_N(t)\|_{H^s} \lesssim \left(\log \frac{T}{\varepsilon}\right)^{\frac{1}{2}}, (t) \leq T.$

\Leftarrow implicit const is independent of N

(8)

Pf: By local theory,

$$\|u_0\|_{H^s} \leq K \Rightarrow \|u_N(t)\|_{H^s} \leq CK$$

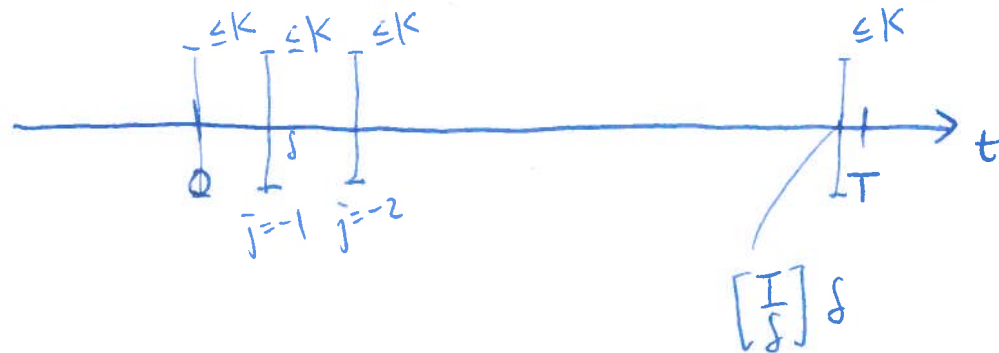
for $|t| \leq \delta \sim K^{-\sigma}$, (indep of N)

Let

$$\Omega_N = \bigcap_{\bar{t} = -[\frac{T}{\delta}]}$$

$$\Phi_N(j\delta) \underbrace{\left(\{ \|u_0\|_{H^s} \leq K \} \right)}_{= B_K}$$

$\Phi_N(t) : u_0 \mapsto u_N(t)$, soln map for (FNLS)



(i) $\mu_N(\Omega_N^c) \leq \sum_{j=-\lceil \frac{T}{\delta} \rceil}^{\lceil \frac{T}{\delta} \rceil} \mu_N(\Phi_N(j\delta)(B_K^c))$ using uniqueness $\textcircled{9}$

By invariance of μ_N $\mu_N(\Phi(-t)A) = \mu_N(A)$

$$\sim \frac{T}{\delta} \mu_N(B_K^c)$$

$$\lesssim T \cdot K^\sigma e^{-cK^2} < \varepsilon$$

Lemma 1.1 & Prop 1.2

$$\left(\mu_N(B_K^c) \stackrel{C-S}{\leq} \underbrace{\|R_{N,r}\|_{L^2(dP_1)}}_{\leq C \text{ indep. of } N} \left(\int_1(B_K^c) \right)^{1/2} \lesssim e^{-cK^2} \right)$$

by choosing $K \sim \left(\log \frac{T}{\varepsilon} \right)^{1/2}$

(ii) By local theory,

$$\|u_N(t)\|_{H^s} \leq CK \sim \left(\log \frac{T}{\varepsilon} \right)^{1/2}, \quad \forall |t| \leq T$$

□

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①

Lemma 1.6: (Approximation lemma) $s < 1/2$

$$u_0 \in H^s \text{ with } \|u_0\|_{H^s} \leq K$$

Suppose soln u_N to (FNLS) with $u_N|_{t=0} = u_0$

satisfies

$$\|u_N(t)\|_{H^s} \leq K, \quad |t| \leq T.$$

Then, $\exists!$ soln u to (NLS) on $[-T, T]$ with $u|_{t=0} = u_0$.

Moreover, we have

$$\|u(t) - P_{\leq N} u_N(t)\|_{H^{s_1}} \leq C_0 e^{C_1(1+K)^{C_2} T} K \underbrace{N^{s_1-s}}_{\rightarrow 0}$$

for $s_1 < s$. (for suff. large $N \in \mathbb{N}$.)

sketch Pf: (FNLS) & (NLS) with $u_N|_{t=0} = u|_{t=0} = u_0$ are locally well-posed on $[-\delta, \delta]$, $\delta \sim (1+K)^{-\delta}$, indep of N .

$$\text{Let } v_N = P_{\leq N} u_N.$$

↑
In fact, choose this as $\delta \sim (1+2K)^{-\delta}$

$$\|u - v_N\|_{X^{s_1}([0, \delta])} \lesssim \|u_0 - P_{\leq N} u_0\|_{H^{s_1}} + \text{nonlin term.} \quad (2)$$

$$\begin{aligned} &= P_{> N} u_0 \\ &\leq N^{s_1 - s} \|u_0\|_{H^s} \\ &\leq N^{s_1 - s} K \end{aligned}$$

Write $|u|^{p-1}u = (\text{Id} - P_{\leq N})(|u|^{p-1}u) + P_{\leq N}(|u|^{p-1}u)$

(1) $(\text{Id} - P_{\leq N})(|u|^{p-1}u)$

$$\lesssim \int^{\theta} N^{s_1 - s} K^p.$$

\uparrow LWP of u in $X^s([0, \delta])$

$$\|u\|_{X^s([0, \delta])} \leq CK.$$

+ (2) $P_{\leq N}(|u|^{p-1}u - |v_N|^{p-1}v_N)$

$$\lesssim \int^{\theta} K^{p-1} \|u - v_N\|_{X^{s_1}([0, \delta])}$$

$$\Rightarrow \|u - v_N\|_{X^{s_1}([0, \delta])} \leq c K N^{s_1 - s} + \frac{1}{2} \|u - v_N\|_{X^{s_1}([0, \delta])} \quad (3)$$

$$\left(\ll \delta^\theta K^{p-1} \ll 1 \right)$$

$$\Rightarrow \|u - v_N\|_{X^{s_1}([0, \delta])} \leq 2c K N^{s_1 - s}$$

• Now iterate the argument $\sim T/\delta$ many times.

$$\|u - v_N\|_{X^{s_1}([\delta, 2\delta])} \approx \underbrace{\|u(\delta) - v_N(\delta)\|_{H^{s_1}}}_{\approx K N^{s_1 - s}} + \underbrace{\text{nonlin term}}_{\text{can be handled exactly as before.}}$$

$$\left(X^{s_1} \subset C_t H^{s_1} \right)$$

etc.

$$\Rightarrow \|u(t) - v_N(t)\|_{H^{s_1}} \lesssim \underbrace{e^{\frac{cT}{\delta}}}_{\rightarrow e^{c(1+K)T}} K N^{s_1 - s}$$

□

Prop 1.7 (Almost a.s. GWP) $s < 1/2$

Given $T, \varepsilon > 0$, $\exists \Omega_{T, \varepsilon} \subset H^s(\mathbb{T})$ s.t.

(i) $\mu(\Omega_{T, \varepsilon}^c) < \varepsilon$

(ii) For $u_0 \in \Omega_{T, \varepsilon}$, $\exists!$ soln u to (NLS) on $[-T, T]$

s.t. $\|u(t)\|_{H^s} \lesssim \left(\log \frac{I}{\varepsilon}\right)^{1/2}, |t| \leq T.$

Pf:

Let $\Omega_N(T, \varepsilon)$ be as in Key Prop.

$\Rightarrow \|\Phi_N(t)(u_0)\|_{H^s} \leq CK$ for $|t| \leq T$ and $u_0 \in \Omega_N$

By Lemma 1.6, $\exists N_1 \gg 1$ s.t.

$\|u(t) - u_N(t)\|_{H^s} \ll 1, |t| \leq T.$

for $N \geq N_1$,

$\Rightarrow \|u(t)\|_{H^s} \lesssim K \sim \left(\log \frac{I}{\varepsilon}\right)^{1/2}, |t| \leq T.$

Also $\mu(\Omega_N^c) \stackrel{C-s}{\leq} \|R_N\|_{L^2(d\mathcal{P}_1)} \left(\int_{\Omega_N^c} \mathbb{1}_{\{\|u\|_{L^2} \leq r\}} d\mathcal{P}_1 \right)^{1/2}$
 $\leq \mathbb{1}_{\{\|P_{\leq N} u\|_{L^2} \leq r\}}$

$$\left(\int \mathbb{1}_{\{\|P_{\leq N} u\|_{L^2} \leq r\}} d\mu \ll \mu_N \right)$$

$$\stackrel{C-S}{\approx} \left(\mu_N(\Omega_N^c) \right)^{\frac{1}{4}} < \varepsilon.$$

(5)

□

Thm 1.8: NLS is globally well-posed almost surely with respect to the Gibbs measure μ .

Pf: Fix $\varepsilon > 0$. Let $T_j = 2^j$, $\varepsilon_j = \varepsilon / 2^j$.

⇒ Construct $\Omega_j = \Omega_{T_j, \varepsilon_j}$

$$\text{Let } \Omega_\varepsilon = \bigcap_{j=1}^{\infty} \Omega_j$$

$$(i) \mu(\Omega_\varepsilon^c) < \sum_{j=1}^{\infty} \frac{\varepsilon}{2^j} = \varepsilon$$

(ii) If $u_0 \in \Omega_\varepsilon$, then the soln u to (NLS) with $u|_{t=0} = u_0$ exists globally in time.

• Let $\Sigma = \bigcup_{\varepsilon > 0} \Omega_\varepsilon$.

$$(i) \mu(\Sigma^c) \leq \inf_{\varepsilon > 0} \varepsilon = 0.$$

⑥

(ii) If $u_0 \in \Sigma$, then the soln u to (NLS) with $u|_{t=0} = u_0$ exists globally in time.

□

Rmk: We have the following probabilistic bound: $s < 1/2$

$$\|u(t)\|_{H^s} \lesssim C(u_0) (\log(1+|t|))^{1/2}, \quad \forall t \in \mathbb{R}.$$

Thm 1.9: The Gibbs measure μ is invariant under the flow of NLS.

Pf: By time reversibility of $\Phi(t)$, it suffices to show

$$(*) \quad \mu(A) \leq \mu(\Phi(t)A)$$

for all measurable set $A \subset H^s$ and $t \in \mathbb{R}$

(if u solves NLS, so does $\bar{u}(-t)$)

By inner regularity,

$$\mu(A) = \sup_{\substack{F \subset A \\ \text{closed in } H^s}} \mu(F)$$

i.e. $\exists \{F_n\}$ closed sets in H^s s.t.

$$F_n \subset A \text{ and } \mu(A) = \lim_{n \rightarrow \infty} \mu(F_n)$$

• Claim: Suffices to prove $(*)$ for closed sets.

$$\Leftarrow \mu(A) = \lim_{n \rightarrow \infty} \mu(F_n)$$

$$\leq \overline{\lim}_{n \rightarrow \infty} \mu(\Phi(t)F_n)$$

$$\leq \mu(\Phi(t)A)$$

\uparrow
 $F_n \subset A$ and uniqueness.

• Given a closed set $F \subset H^s$,

$$\text{let } K_n = \{u \in F : \|u\|_{H^\sigma} \leq n\}, \quad s < \sigma < \frac{1}{2}.$$

Then, K_n is compact in H^s . (Rellich's lemma)

(7)

\Rightarrow Suffices to prove $\textcircled{*}$ for compact sets.

$\textcircled{8}$

$$\mu(F) = \lim_{n \rightarrow \infty} \mu(K_n)$$

\uparrow
Lem 1.1. & Prop 1.2

• Let K be a cpt set in H^S .

Since $\mu_N \rightarrow \mu$ & Portmanteau theorem,

we have

$$(1) \quad \mu(\Phi(t)K + \overline{B_\varepsilon}) \geq \overline{\lim} \mu_N(\Phi(t)K + \overline{B_\varepsilon})$$

• Fix $t \ll 1$. Then, \swarrow local theory

$$\Phi_N(t)(K + B_\varepsilon) \subset \Phi_N(t)K + B_{\varepsilon/2}$$

$$\subset \Phi(t)K + B_\varepsilon$$

\uparrow
Approximation lemma.

\Rightarrow By invariance of μ_N ,

$$(2) \quad \mu_N(K + B_\varepsilon) \leq \mu_N(\Phi(t)K + B_\varepsilon)$$

Therefore, $\mu(K) \leq \mu(K + B_\varepsilon)$

(9)

$$\leq \underline{\lim} \mu_N(K + B_\varepsilon)$$

$$\leq \underline{\lim}_{(2)} \mu_N(\Phi(t)K + B_\varepsilon)$$

$$\leq \overline{\lim} \mu_N(\Phi(t)K + \overline{B_\varepsilon})$$

$$\stackrel{(1)}{\leq} \mu(\Phi(t)K + \overline{B_\varepsilon})$$

\Rightarrow Let $\varepsilon \rightarrow 0$. $\mu(K) \leq \mu(\Phi(t)K)$ for $t \ll 1$

\Rightarrow for all $t \geq 0$.

\Rightarrow for all $t \in \mathbb{R}$

□

Lec 8: 01 / 02 / 17 (Wed)

①

Chap 2: Probabilistic well-posedness of dispersive PDEs

• Nonlinear wave equations on \mathbb{T}^d (or \mathbb{R}^d)

$$(NLW) \quad \underbrace{(-\partial_t^2 + \Delta)}_{=\square} u = \pm |u|^{p-1} u$$

2.1) Review on deterministic theory of NLW.

• Consider the non-homog linear wave equation

$$\begin{cases} \square u = \pm F \\ (u, \partial_t u)|_{t=0} = (u_0, u_1) \end{cases}$$

Duhamel

$$\Leftrightarrow u(t) = \underbrace{\cos(t|\nabla|) u_0 + \frac{\sin t|\nabla|}{|\nabla|} u_1}_{S(t)(u_0, u_1)} \mp \int_0^t \frac{\sin(t-t')|\nabla|}{|\nabla|} F(t') dt'$$

$$\left(\begin{array}{l} \text{With } v = \partial_t u, \\ \partial_t \begin{pmatrix} u \\ v \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ \Delta & 0 \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} = \begin{pmatrix} 0 \\ F \end{pmatrix} \end{array} \right) \quad (2)$$

Basic tool: Strichartz estimates: $d \geq 2$

$$s \geq 0, \quad 2 \leq q \leq \infty, \quad 2 \leq r < \infty, \quad (q, r, d) \neq (2, \infty, 3)$$

$$\text{Scaling condition: } \frac{1}{q} + \frac{d}{r} = \frac{d}{2} - s$$

$$\text{admissibility condition: } \frac{1}{q} + \frac{d-1}{2r} \leq \frac{d-1}{4}. \quad (\Leftarrow \text{Knapp's counterexample})$$

- Suppose (q, r) , s -admissible
 (\tilde{q}, \tilde{r}) , $(1-s)$ -admissible

$$0 \leq s \leq 1$$

Then, we have

On \mathbb{R}^d

$$\begin{aligned} & \| (u, \partial_t u) \|_{L_t^\infty \dot{H}^s} + \| u \|_{L_t^q L_x^r} \\ & \lesssim \| (u_0, u_1) \|_{\dot{H}^s} + \| F \|_{L_t^{\tilde{q}'} L_x^{\tilde{r}'}} \end{aligned}$$

$$\dot{H}^s = \dot{H}^s \times \dot{H}^{s-1} \\ \Downarrow \\ (u_0, u_1)$$

Rmk: Thanks to $|\nabla|^{-1}$ in the Duhamel term,
Sobolev inequality is also very effective.

(3)

ex: Cubic NLW on \mathbb{R}^3 :

$$-\partial_t^2 u + \Delta u = \pm u^3$$

LWP in $\dot{H}^1(\mathbb{R}^3)$

$$\Gamma_{(u_0, u_1)}(u) = S(t)(u_0, u_1) \mp \int_0^t \frac{\sin(t-t')|\nabla|}{|\nabla|} u^3(t') dt'$$

Sobolev: $\|f\|_{L_x^6(\mathbb{R}^3)} \lesssim \|f\|_{\dot{H}^1(\mathbb{R}^3)}^{\frac{1}{3} \leftarrow \frac{5}{d}} = \frac{1}{2} - \frac{1}{6}$

$$\|\Gamma u\|_{L_T^\infty \dot{H}^1} \lesssim \|(u_0, u_1)\|_{\dot{H}^1} \left(2^* = \frac{2d}{d-2} \right) \leftarrow \text{By unitarity of } e^{\pm it|\nabla|}$$

$$+ \int_0^T \underbrace{\|u^3\|_{L_T^\infty L_x^2}}_{= \|u\|_{\infty, 6}^3} dt' \lesssim \|u\|_{\infty, 6}^3$$

$$\Rightarrow \| \Gamma u \|_{L_T^\infty \dot{H}^1} \lesssim \| (u_0, u_1) \|_{\dot{H}^1} + T \| u \|_{L_T^\infty \dot{H}^1}^3 \quad (4)$$

$$\| \Gamma u - \Gamma v \|_{L_T^\infty \dot{H}^1} \lesssim T \left(\| u \|_{L_T^\infty \dot{H}^1}^2 + \| v \|_{L_T^\infty \dot{H}^1}^2 \right) \| u - v \|_{L_T^\infty \dot{H}^1}$$

\Rightarrow By the fixed pt argument, cubic NLW on \mathbb{R}^3 is LWP in $\dot{H}^1(\mathbb{R}^3)$

• the same proof works on \mathbb{T}^3 .

Q: How low can we go?

scaling/dilation symmetry = $u(x, t)$ soln to NLW (with u^p)

$$\Rightarrow u_\lambda(x, t) = \frac{1}{\lambda^{\frac{d}{2} - \frac{1}{p-1}}} u\left(\frac{x}{\lambda}, \frac{t}{\lambda}\right) \text{ is also a soln.}$$

We say that s is a critical Sobolev index if

$$\| f_\lambda \|_{\dot{H}^s(\mathbb{R}^d)} = \| f \|_{\dot{H}^s(\mathbb{R}^d)}, \quad f_\lambda(x) = \frac{1}{\lambda^{\frac{d}{2} - \frac{1}{p-1}}} f\left(\frac{x}{\lambda}\right)$$

$$s_{\text{crit}} = \frac{d}{2} - \frac{2}{p-1}$$

$$d=3, p=3; \underline{S_{crit} = \frac{1}{2}}$$

(5)

Moral: Given $(u_0, u_1) \in \dot{H}^s(\mathbb{R}^d)$,

- (i) $S > S_{crit}$: subcritical \Rightarrow (expect) well-posedness
- (ii) $S = S_{crit}$: critical \Rightarrow (expect) well-posedness
but more delicate
- (iii) $S < S_{crit}$: supercritical \Rightarrow bad behavior (ill-posedness)

Norm inflation: $\forall \varepsilon > 0, \exists$ ^{smooth} soln u_ε and $t_\varepsilon \in (0, \varepsilon)$

s.t. ① $\| (u_\varepsilon(0), \partial_t u_\varepsilon(0)) \|_{\dot{H}^s} < \varepsilon$

$S < S_{crit}$

but ② $\| (u_\varepsilon(t_\varepsilon), \partial_t u_\varepsilon(t_\varepsilon)) \|_{\dot{H}^s} > \varepsilon^{-1}$

\Rightarrow failure of continuity of the soln map $\Phi: (u_0, u_1) \mapsto (u, \partial_t u)$
(at $(0, 0)$)

$$\dot{H}^s \xrightarrow{\quad} C_t \dot{H}^s$$

Christ - Colliander - Tao '03 arXiv

LWP of cubic NLW in $\dot{H}^{\frac{1}{2}}(\mathbb{R}^3)$

(6)

↑ lowest possible regularity.

Use Strichartz with

$$s = \frac{1}{2}, \quad (q, r) = (4, 4) \\ = (\hat{q}, \hat{r})$$

$$\frac{1}{4} + \frac{3}{4} = \frac{3}{2} - \frac{1}{2}$$

$$\frac{1}{4} + \frac{3-1}{2 \cdot 4} = \frac{1}{2} \leq \frac{3-1}{4}$$

$$\Rightarrow \| \Gamma u \|_{L_T^\infty \dot{H}^{\frac{1}{2}} \cap L_T^4 L_x^4} = \dot{X}^{\frac{1}{2}}$$

(*)

$$\lesssim \| (u_0, u_1) \|_{\dot{X}^{\frac{1}{2}}} + \| u^3 \|_{L_T^{\frac{4}{3}} L_x^{\frac{4}{3}}}$$

No T^θ !!

$$= \| u \|_{L_T^4 L_x^4}^3 \leq \| u \|_{L_T^\infty \dot{H}^{\frac{1}{2}} \cap L_T^4 L_x^4}^3$$

$$\leq \gamma^2 \| u \|_{\dot{X}^{\frac{1}{2}}}$$

← manifestation of criticality.

Can carry out a contraction mapping principle in B_γ

where $\gamma = 2C \| (u_0, u_1) \|_{\dot{X}^{\frac{1}{2}}} \ll 1$
 ↑
 const on lin soln.

Rmk: can take $T = \infty$ if initial data is suff. small.
(small data global well-posedness)

⑦

For large data, note that

$$\|S(t)(u_0, u_1)\|_{L_{T,x}^4} \rightarrow 0 \text{ as } T \rightarrow 0 \text{ (by DCT)}$$

$$(\varepsilon \| (u_0, u_1) \|_{\dot{H}^{1/2}} < \infty.)$$

By Strichartz estimate,

$$\| \Gamma u \|_{L_{T,x}^4} \leq \| S(t)(u_0, u_1) \|_{L_{T,x}^4} + C \| u \|_{L_{T,x}^4}^3$$

\Rightarrow contraction on B_{γ} in $L_{T,x}^4$

$$\gamma = 2 \| S(t)(u_0, u_1) \|_{L_{T,x}^4} \ll 1.$$

\Rightarrow By $\textcircled{*}$, $u \in C_T \dot{H}^{1/2}$.

Rmk: ① The same LWP holds on \mathbb{T}^3

⑧

② Local existence time $T = T((u_0, u_1))$.

(i.e. more info than $\|(u_0, u_1)\|_{H^{1/2}}$ is needed.)

Let $s < 1/2$. Consider $(u_0, u_1) \in \mathcal{H}^s(\mathbb{T}^3) \setminus \mathcal{H}^{1/2}(\mathbb{T}^3)$

Q: Can we construct a soln u with $(u, \partial_t u)|_{t=0} = (u_0, u_1)$?

Idea: "Randomize" (u_0, u_1)

$$f \text{ on } \mathbb{T}^d, \quad f(x) = \sum_{n \in \mathbb{Z}^d} \hat{f}(n) e^{in \cdot x}$$

Randomization of f :

$$f^\omega(x) = \sum_{n \in \mathbb{Z}^d} \underline{g_n(\omega)} \hat{f}(n) e^{in \cdot x}$$

Lec 9: 06/02/17 (Mon)

①

2.2 Randomization of ϵ function on \mathbb{T}^d (and \mathbb{R}^d)

$$f(x) = \sum_{n \in \mathbb{Z}^d} \hat{f}(n) e^{in \cdot x}, \quad x \in \mathbb{T}^d.$$

Let $\{g_n\}_{n \in \mathbb{Z}^d}$ be a collection of independent ^{mean-zero} random variables with distributions of real and imaginary parts $\mu_n^{(r)}, \mu_n^{(i)}$.
assume indep.

$$\int e^{\gamma \cdot x} d\mu_n^{(r)}(x) \leq e^{c\gamma^2}$$
$$\int e^{\gamma \cdot x} d\mu_n^{(i)}(x) \leq e^{c\gamma^2}, \quad \forall n \in \mathbb{Z}^d, \gamma \in \mathbb{R}$$

\Leftarrow satisfied by standard \mathbb{C} -valued Gaussian r.v.'s.,
Bernoulli (or uniform distri on $\mathbb{S}^1 \subset \mathbb{C}$).

Rmk: In many situations, it suffices to assume (uniform) (2)
 boundedness of the k^{th} moment $\mathbb{E}[|g_n|^k] \leq c < \infty$

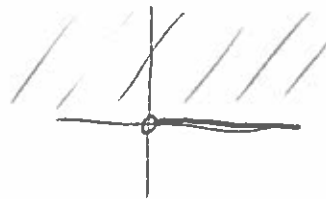
\Rightarrow Randomization f^ω of f :

$$f^\omega(x) = \sum_{n \in \mathbb{Z}^d} \underline{g_n(\omega)} \hat{f}(n) e^{in \cdot x}$$

Rmk: If f is real, i.e. $\hat{f}(-n) = \overline{\hat{f}(n)}$, $\forall n \in \mathbb{Z}^d$.

Let $I = \bigcup_{k=0}^{d-1} \mathbb{Z}^k \times \mathbb{Z}_+ \times \{0\} \stackrel{d-k-1}{=} \mathbb{Z}^d / 2$

$\Rightarrow \mathbb{Z}^d = I \cup \{-I\} \cup \{0\}$



Then, introduce $\{g_n\}_{n \in I \cup \{0\}}$

and set $g_{-n} = \overline{g_n}$, $\forall n \in \mathbb{Z}^d$. (g_0 is real-valued.)

$\Rightarrow f^\omega(x) = \sum_{n \in \mathbb{Z}^d} g_n(\omega) \hat{f}(n) e^{in \cdot x}$ is real-valued.

Moral: If $f \in H^s \setminus H^{s+\varepsilon}$, then $f^\omega \in H^s \setminus H^{s+\varepsilon}$ a.s. (3)

i.e. there is no gain in differentiability.

• If $f \in L^2$, then $f^\omega \in L^p$, $\forall 2 \leq p < \infty$, a.s.

\Rightarrow gain of integrability.

Lemma 2.1: (i) Given $f \in H^s(\mathbb{T}^d)$, let f^ω be the randomization defined above. Then, the following tail estimate holds:

$$P(\|f^\omega\|_{H^s} > \lambda) \leq C \exp\left(-c \frac{\lambda^2}{\|f\|_{H^s}^2}\right) \xrightarrow{\text{as } \lambda \rightarrow \infty} 0$$

for all $\lambda > 0$. In particular, $f^\omega \in H^s$, a.s.

(ii) Let $f \in L^2(\mathbb{T}^d)$. Then, for any $\sqrt[p]{p} \geq 2$, we have

$$P(\|f^\omega\|_{L^p} > \lambda) \leq C \exp\left(-c \frac{\lambda^2}{\|f\|_{L^2}^2}\right).$$

for all $\lambda > 0$. In particular, $f^\omega \in L^p$, a.s.

Paley-Zygmund '30

Kahane's book

Lemma 2.2: Let $\{g_n\}_{n \in \mathbb{Z}^d}$ be as above. Then, we have ④

$$\left\| \sum_{n \in \mathbb{Z}^d} c_n g_n \right\|_{L^p(\Omega)} \lesssim \sqrt{p} \|c_n\|_{\ell_n^2}$$

Pf: We first prove

$$P(|\sum c_n g_n| > \lambda) \leq c e^{-c \frac{\lambda^2}{\|c_n\|_{\ell_n^2}^2}}$$

$$\int e^{t \sum c_n g_n} dP \stackrel{\text{indep}}{=} \prod_n \int e^{t c_n g_n} dP$$

$$\leq \prod_n e^{\alpha (t c_n)^2} = e^{\alpha^2 t^2 \sum c_n^2}$$

$$\Rightarrow P(\sum c_n g_n > \lambda) \leq \frac{\mathbb{E}[e^{t \sum c_n g_n}]}{e^{t\lambda}} \leq \frac{e^{\alpha^2 t^2 \sum c_n^2}}{e^{t\lambda}}$$

$$\text{choose } t = \frac{\lambda}{2\alpha^2 \sum c_n^2} = e^{-\frac{\lambda^2}{4\alpha^2 \sum c_n^2}}$$

Similarly, $P(\sum c_n g_n < -\lambda) \leq c e^{-c \frac{\lambda^2}{\|c_n\|_{\ell_n^2}^2}}$

$$\Rightarrow \|\sum c_n g_n\|_{L^p(\Omega)}^p \leq p \int_0^\infty \lambda^{p-1} e^{-c \frac{\lambda^2}{\|c_n\|_{\ell_n^2}^2}} d\lambda \quad (5)$$

$$\stackrel{\text{ch. of var}}{\sim} p \left(\sum c_n^2\right)^{p/2} \int_0^\infty \lambda^{p-1} e^{-\lambda^2/2} d\lambda$$

$$\sim \left(p \sum c_n^2\right)^{p/2} \sim (p-1)!! = \frac{(2p)!}{2^p \cdot p!}$$

□

Aside $\int_{\mathbb{R}} x^k e^{-x^2/2} dx$, k even.

⇐ IBP does not help. Use the moment generating function

$$\mathbb{E}[e^{tx}] \stackrel{\text{ch. of var}}{=} e^{\frac{1}{2}t^2} = \sum_{n=0}^{\infty} \frac{1}{n! 2^n} t^{2n}$$

$$\mathbb{E}\left[\sum_{k=0}^{\infty} \frac{(tX)^k}{k!}\right] = \sum_{k=0}^{\infty} \frac{\mathbb{E}[X^k]}{k!} t^k. \quad \text{Compare the coefficients.}$$

pf of Lemma 2.1: By Minkowski's integral inequality.

(6)

$$\forall p \geq 2 \quad \left\| \left\| f^\omega \right\|_{H^s} \right\|_{L^p(\Omega)} \leq \left\| \left\| \sum_n \underbrace{\langle m \rangle^s \hat{f}(m)}_{c_n(x)} e^{in \cdot x} g_m \right\|_{L^p(\Omega)} \right\|_{L^2_x}$$

$$\| \langle \nabla \rangle^s f^\omega \|_{L^2_x} \stackrel{\text{Lem 2.2}}{\lesssim} \sqrt{p} \left\| \underbrace{\| \langle m \rangle^s \hat{f}(m) \|_{l_n^2}}_{= \| f \|_{H^s}} \right\|_{L^2_x(\mathbb{T}^d)}$$

$$\sim \sqrt{p} \| f \|_{H^s}.$$

By Chebyshev's ineq,

$$P(\| f^\omega \|_{H^s} > \lambda) < \left(\frac{C_0 p^{1/2} \| f \|_{H^s}}{\lambda} \right)^p, \quad \forall p \geq 2.$$

Let $p = \left(\frac{\lambda}{C_0 e \| f \|_{H^s}} \right)^2$. If $p \geq 2$, then we have

$$P(\| f^\omega \|_{H^s} > \lambda) < e^{-p} = e^{-c \frac{\lambda^2}{\| f \|_{H^s}^2}}$$

If $p = \left(\frac{\lambda}{C_0 e \|f\|_{H^s}} \right)^2 \leq 2$, then

(7)

choose C s.t. $C e^{-2} \geq 1$. Then, we have

$$P(\|f^\omega\|_{H^s} > \lambda) \leq 1 \leq C e^{-2} \leq C e^{-C \lambda^2 / \|f\|_{H^s}^2}$$

(Alternative way: Establish $\mathbb{E}[e^{\alpha \|f^\omega\|_{H^s}^2}] < \infty$

by Taylor expansion + $\| \|f^\omega\|_{H^s} \|_{L^p(\Omega)} \lesssim \sqrt{p} \|f\|_{H^s}$

Given $p \geq 2$,
let $q \geq p$.

$$\| \|f^\omega\|_{L^p_x} \|_{L^q(\Omega)} \leq \| \| \sum_n \underbrace{\hat{f}(n)}_{c_n} e^{in \cdot x} \cdot g_n \|_{L^q(\Omega)} \|_{L^p_x}$$

$$\lesssim \sqrt{q} \| \| \hat{f}(n) \|_{\ell_n^2} \|_{L^p_x} \|$$

$$= \| f \|_{L^2}$$

$$\sim \sqrt{q} \| f \|_{L^2} \quad \forall q \geq p.$$

□

Lemma 2.3: Suppose $f = \sum \widehat{f}(m) e^{in \cdot x} \in H^s \setminus H^{s+\varepsilon}$ for some $\varepsilon > 0$. ⑧

Suppose $\exists c > 0$ s.t.

$$\limsup_{|n| \rightarrow \infty} P(\underbrace{\{|g_n| \leq c\}}_{\mu_n(-c, c)}) \leq 1 - \delta < 1$$

(e.g. satisfied if g_n is i.i.d. and $g_n \neq 0$)

Then, $P(f^\omega \in H^{s+\varepsilon}) = 0$

Pf: For simplicity, assume g_n is real-valued.

$$\begin{aligned} \int e^{-\|f^\omega\|_{H^{s+\varepsilon}}^2} dP & \stackrel{\text{indep}}{=} \prod_n \int e^{-\langle n \rangle^{2(s+\varepsilon)} |\widehat{f}(m)|^2 |g_n|^2} d\mu_n \\ & \leq \prod_n \left(\mu_n(-c, c) + \underbrace{e^{-c^2 \langle n \rangle^{2(s+\varepsilon)} |\widehat{f}(m)|^2}}_{=: \alpha_n} \int_{|g_n| \geq c} (1 - \mu_n(-c, c)) \right) \\ & = \prod_n (\mu_n(-c, c)(1 - \alpha_n) + \alpha_n) \\ & \leq \prod_n ((1 - \delta)(1 - \alpha_n) + \alpha_n) = \prod_n (1 - \delta(1 - \alpha_n)) \stackrel{**}{=} 0 \end{aligned}$$

$$\Rightarrow \|f^\omega\|_{H^{s+\varepsilon}} = \infty, \text{ a. s.}$$

(9)

Proof of $(*)$: By assumption, $\sum \langle n \rangle^{2(s+\varepsilon)} |\hat{f}(n)|^2 = \infty$.

Then, we claim $\sum (1 - \alpha_n) = \infty$

$$\sum (1 - e^{-c^2 \langle n \rangle^{2(s+\varepsilon)} |\hat{f}(n)|^2})$$

- If $\langle n \rangle^{2(s+\varepsilon)} |\hat{f}(n)|^2 \rightarrow 0$ as $|n| \rightarrow \infty$, then $\sum (1 - \alpha_n) \geq \sum (1 - \varepsilon) = \infty$.
 - Otherwise, $\exists N$ s.t. $c^2 \langle n \rangle^{2(s+\varepsilon)} |\hat{f}(n)|^2 \leq 1, \forall |n| \geq N$
- $$\Rightarrow \sum_{|n| \geq N} (1 - \alpha_n) \leq C \sum_{|n| \geq N} \langle n \rangle^{2(s+\varepsilon)} |\hat{f}(n)|^2 = \infty$$
- (b/c $e^{-x} \leq 1 - Cx$ for $0 \leq x \leq 1$ (for some $C \ll 1$))

Then,
$$\left(\prod_n (1 - \delta(1 - \alpha_n)) \right)^{-1} = \prod_n \left(1 + \frac{\delta(1 - \alpha_n)}{1 - \delta(1 - \alpha_n)} \right)$$

$$\leq \prod_n (1 + 2\delta(1 - \alpha_n)) \quad \text{by choosing } \delta \leq \frac{1}{2}.$$

$$= \infty.$$

$$\left(\prod_n (1 + a_n) < \infty \iff \sum a_n < \infty. \right)$$

Lec 10 = 08/02/17 (Wed)

①

Probabilistic Strichartz estimate:

$S(t) = e^{it\Delta}$ linear Schrödinger semigroup

• Strichartz estimates on \mathbb{R}^d

$$\|S(t)f\|_{L_t^q L_x^r(\mathbb{R} \times \mathbb{R}^d)} \lesssim \|f\|_{L^2(\mathbb{R}^d)}$$

$$2 \leq q, r \leq \infty, (q, r, d) \neq (2, \infty, 2)$$

$$\frac{2}{q} + \frac{d}{r} = \frac{d}{2}$$

Prop 2.4: $f \in L^2(\mathbb{T}^d)$

Given $2 \leq q, r < \infty$, we have

$$P\left(\|S(t)f^w\|_{L_t^q L_x^r([-T, T] \times \mathbb{T}^d)} > \lambda\right) \leq C \exp\left(-c \frac{\lambda^2}{T^{\frac{2}{q}} \|f\|_{L^2}^2}\right)$$

for all $\lambda > 0$ and all $T > 0$.

$$\left(\begin{array}{l} \rightarrow 0 \text{ as } \lambda \rightarrow \infty \\ \text{or } T \rightarrow 0 \end{array}\right)$$

Rmk: • Set $\lambda = T^\theta \|f\|_{L^2}$, $\theta < \frac{1}{8}$.

(2)

Prop 2.4 $\Rightarrow \|S(t)f^\omega\|_{L_T^q L_x^r} \leq T^\theta \|f\|_{L^2}$

outside a set of prob $\leq c e^{-c T^{2(\theta - \frac{1}{8})}}$ ($\rightarrow 0$ as $T \rightarrow 0$)

• Fix $T > 0$. Given $\varepsilon > 0$,

$\|S(t)f^\omega\|_{L_T^q L_x^r} \leq C_T \left(\log \frac{1}{\varepsilon}\right)^{1/2} \|f\|_{L^2}$ outside a set of prob $< \varepsilon$.
 (or $\leq C_T K \|f\|_{L^2}$ prob $< e^{-cK^2}$)

Pf: Let $p \geq \max(q, r)$.

$\| \|S(t)f^\omega\|_{L_T^q L_x^r} \|_{L^p(\Omega)} \stackrel{\text{Mink}}{\leq} \| \|S(t)f^\omega\|_{L^p(\Omega)} \|_{L_T^q L_x^r}$

$\| \sum e^{-itm^2} \widehat{f}(m) e^{in \cdot x} g_n \| = C_n(t, x)$

$\lesssim \sqrt{p} \| \|\widehat{f}\|_{l^n} \|_{L_T^q L_x^r}$

$= \sqrt{p} T^{\frac{1}{4}} \|f\|_{L_x^2} \Rightarrow \text{Apply Chebyshev} \quad \square$

• The same estimate holds for the half wave operator $e^{\pm it|\nabla|}$ (3)

$$\Rightarrow \text{Also for } \cos t|\nabla| = \frac{e^{it|\nabla|} + e^{-it|\nabla|}}{2}$$

• Next, we consider $\frac{\sin t|\nabla|}{|\nabla|}$.

Prop 2.5: (probabilistic Strichartz estimate for wave eqn)

$$\text{Let } (f_0, f_1) \in \mathcal{H}^0 = L^2 \times H^{-1}$$

$$P \left(\left\| \begin{matrix} S(t) (f_0^\omega, f_1^\omega) \\ \cos t|\nabla|, \frac{\sin t|\nabla|}{|\nabla|} \end{matrix} \right\|_{L_t^q L_x^r(I \times \mathbb{T}^d)} > \lambda \right)$$

$$I = [a, b]$$

$$\leq C \exp \left(-c \frac{\lambda^2}{\max(1, a^2, b^2) |I|^{2/q} \|(f_0, f_1)\|_{\mathcal{H}^0}^2} \right)$$

Pf: Repeat the proof of Prop 2.4, noting

$$\left| \frac{\sin t|n|}{|n|} \right|_{n=0} = |t| \leq \max(1, |a|, |b|)$$

See O. Pocovnicu

arXiv 15

\mathbb{R}^d ?

$$f(x) = \int \hat{f}(\xi) e^{ix \cdot \xi} d\xi$$

$$= \sum_{n \in \mathbb{Z}^d} \int_{Q_n} \hat{f}(\xi) e^{ix \cdot \xi} d\xi$$

$$Q_n = m + \left(-\frac{1}{2}, \frac{1}{2}\right]^d$$

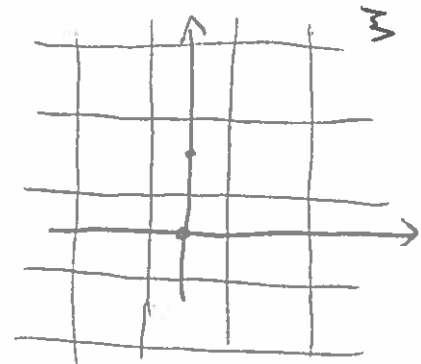
$$\Rightarrow \mathbb{R}^d = \bigcup_{n \in \mathbb{Z}^d} Q_n \quad \text{Wiener decomposition.}$$

For practical purpose, let $\psi \in C_c^\infty(\mathbb{R}^d; [0, 1])$ s.t

$$\text{supp } \psi \subset [-1, 1]^d$$

$$\text{and } \sum_{n \in \mathbb{Z}^d} \underbrace{\psi(\xi - n)}_{\text{smoothed version of } \mathbf{1}_{Q_n}} \equiv 1$$

$$\begin{aligned} \Rightarrow f(x) &= \sum_{n \in \mathbb{Z}^d} \underbrace{\psi(\xi - n)}_{\text{smoothed version of } \mathbf{1}_{Q_n}} f \\ &= \int \psi(\xi - n) \hat{f}(\xi) e^{ix \cdot \xi} d\xi. \end{aligned}$$



Randomization of a function on \mathbb{R}^d

(5)

$$f^w = \sum_{n \in \mathbb{Z}^d} g_n \psi(D-n)f.$$

Rmk: Why unit scale? \leftarrow a matter of choice.

Can introduce a similar randomization based on dilated cubes. (e.g. Bényi-O-P '15)

• translation on $\mathbb{R}^d_{\frac{1}{3}}$ \iff modulation $e^{ix \cdot \xi_0}$ on \mathbb{R}^d_x

$$\rightsquigarrow \text{modulation space } \|f\|_{M^p_s} = \left\| \left\| \langle n \rangle^s \psi(D-n)f \right\|_{L^p_x} \right\|_{l^n_q}$$

Wiener amalgam space

$$\|f\|_{W^{p,q}} = \left\| \left\| \psi(D-n)f \right\|_{L^p_x} \right\|_{l^n_q}$$

Rmk: On \mathbb{T}^d ,

$$\|f\|_{M^{p,q}} = \|f\|_{W^{p,q}} = \|f\|_{F^{p,q}} = \|\hat{f}(m)\|_{l^n_q}$$

• What about randomization based on the LP-decomposition?

$$f = \sum_{\substack{N \geq 1 \\ \text{dyadic}}} P_N f, \quad P_N = \overset{LP}{\text{proj onto } \{|\xi| \sim N\}}$$

Let's consider

$$f^\omega = \sum_{\substack{N \geq 1 \\ \text{dyadic}}} \pm P_N f.$$

$1 < p < \infty$

$$\|f^\omega\|_{L^p} \stackrel{\text{LP theory}}{\sim} \left\| \left(\sum_{\substack{N \geq 1 \\ \text{dyadic}}} |P_N f|^2 \right)^{1/2} \right\|_{L^p}$$

$\sim \|f\|_{L^p} \Rightarrow$ No gain of integrability!

• Back to Wiener randomization on \mathbb{R}^d .

We can not use the finiteness of the domain.

\Rightarrow Use Bernstein's inequality

$$\|P_N f\|_{L^q} \lesssim N^{\frac{d}{p} - \frac{d}{q}} \|f\|_{L^p}$$

In fact,

$$\| \mathcal{F}^{-1}(\mathbb{1}_Q f) \|_{L^q} \lesssim |Q|^{\frac{1}{p} - \frac{1}{q}} \|f\|_{L^p}$$

really smooth cutoff

$\xi_0 =$ center of Q

$$\| e^{ix \cdot \xi_0} \mathcal{F}^{-1}(\mathbb{1}_Q f) \|_{L^q}$$

centered at 0 on \mathbb{R}^d

$$\Rightarrow \| \Psi(D-n) f \|_{L_x^r} \lesssim \| \Psi(D-n) f \|_{L_x^2} \quad (7)$$

\Rightarrow For $p \geq \max(q, r)$

$$\| \| S(t) f^\omega \|_{L_T^q L_x^r} \|_{L^p(\Omega)} \stackrel{\text{Mink}}{\leq} \| \| \underbrace{\sum_n (\Psi(D-n) S(t) f) g_n}_{= C_n} \|_{L^p(\mu)} \|_{L_T^q L_x^r}$$

$$\leq \sqrt{p} \| \| \Psi(D-n) S(t) f \|_{l_n^2} \|_{L_T^q L_x^r}$$

$$\stackrel{\text{Mink}}{\leq} \sqrt{p} \| \| \underbrace{\Psi(D-n) S(t) f}_{L_x^r} \|_{L_T^q l_n^2}$$

$$\stackrel{\text{Bernstein}}{\lesssim} \| \Psi(D-n) \underbrace{S(t) f}_{L_x^2} \|_{L_x^2}$$

$$\lesssim \sqrt{p} \| \| f \|_{L^2(\mathbb{R}^d)} \|_{L_T^q} \quad \begin{matrix} \uparrow \\ \text{unitarity} \end{matrix}$$

$$\lesssim \sqrt{p} T^{\frac{1}{q}} \| f \|_{L^2(\mathbb{R}^d)} \Rightarrow \text{Prob. Strichartz estimate on } \mathbb{R}^d.$$

Prop 2.6: (Global-in-time prob. Strichartz estimate on \mathbb{R}^d)

⑧

Let (q, r) be Schrödinger admissible,

$$q, r < \infty.$$

Let $\tilde{r} \geq r$ finite. Then,

$$P \left(\| S(t) f^w \|_{L_t^q L_x^{\tilde{r}}(\mathbb{R} \times \mathbb{R}^d)} > \lambda \right) \leq C \exp \left(-c \frac{\lambda^2}{\|f\|_{L^2}^2} \right)$$

← Proof uses deterministic Strichartz estimate.

Lec 11: 13/02/17 (Mon)

①

Prop 2.7: $f \in H^\varepsilon(\mathbb{T}^d)$, $\varepsilon > 0$.

Given $2 \leq q < \infty$, we have

$$P\left(\|S(t)f^\omega\|_{L_t^q L_x^\infty([-T, T] \times \mathbb{T}^d)} > \lambda\right) \leq C \exp\left(-c \frac{\lambda^2}{T^{2/q} \|f\|_{H^\varepsilon}^2}\right)$$

Pf: By Sobolev embedding,

$$\begin{aligned} \|S(t)f^\omega\|_{L_T^q L_x^\infty} &\lesssim \|\langle \nabla \rangle^\varepsilon S(t)f^\omega\|_{L_T^q L_x^r} \quad \varepsilon r > d \\ &= \|S(t)(\langle \nabla \rangle^\varepsilon f^\omega)\|_{L_T^q L_x^r} \end{aligned}$$

\Rightarrow Apply Prop 2.4.

□

- The same estimate holds if $q = \infty$.

(2)

By Sobolev in time,

$$\|S(t)f^w\|_{L_T^\infty L_x^r} \lesssim \|\langle \partial_t \rangle^\varepsilon\|_{L_T^q} \|S(t)f^w\|_{L_x^r} \quad \varepsilon q > 1$$

See also
Oh-Pocovnicu'16
JMPA.

(*)

if we can place $\langle \partial_t \rangle^\varepsilon$ inside L_x^r -norm,
we can replace $\langle \partial_t \rangle^\varepsilon$ by $\langle \nabla \rangle^{2\varepsilon}$ (Schrödinger case)

ex: $\varepsilon = 1$. consider

$$\begin{aligned} \|\partial_t\|_{L_T^q L_x^r} &= \left\| \lim_{h \rightarrow 0} \frac{\|F(t+h)\|_{L_x^r} - \|F(t)\|_{L_x^r}}{h} \right\|_{L_T^q} \\ &\leq \left\| \lim_{h \rightarrow 0} \left\| \frac{F(t+h) - F(t)}{h} \right\|_{L_x^r} \right\|_{L_T^q} \\ &= \left\| \lim_{h \rightarrow 0} \left\| \frac{F(t+h) - F(t)}{h} \right\|_{L_x^r} \right\|_{L_T^q} \end{aligned}$$

for nice F .

Justify (*) by interpolation.

(2.3) Probabilistic well-posedness of NLW on $\mathbb{T}^3/\mathbb{R}^3$ (3)

• Cubic NLW:
$$\begin{cases} -\partial_t^2 u + \Delta u = u^3 \\ (u, \partial_t u)|_{t=0} = (u_0, u_1) \end{cases}$$
 defocusing (needed only for the global argument)

Recall $S_{\text{crit}} = 1/2$.

Assume $(u_0, u_1) \in \mathcal{H}^s \setminus \mathcal{H}^{1/2}$, $0 \leq s < 1/2$

\Rightarrow Randomize (u_0, u_1) and consider

(*)
$$\begin{cases} -\partial_t^2 u + \Delta u = u^3 \\ (u, \partial_t u)|_{t=0} = (u_0^w, u_1^w) \end{cases}$$

Thm 2.8: The Cauchy problem (*) is almost surely locally well-posed.

More precisely, given small $T > 0$, $\exists \Omega_T$ (also depends on the fixed (u_0, u_1))

s.t. ① $P(\Omega_T^c) \leq c e^{-c/T^\delta}$ for some $\delta > 0$.

② For each $w \in \Omega_T$, $\exists!$ soln u to (*) with $(u, \partial_t u)|_{t=0} = (u_0^w, u_1^w)$

Rmk: • $\Sigma = \bigcup_{0 < T \ll 1} \Omega_T$. Then, $P(\Sigma) = 1$.

(4)

• For each $\omega \in \Sigma$, u exists on $[-T\omega, T\omega]$.

• u lies in

$$S(t)(u_0^\omega, u_1^\omega) + \underbrace{C_T H^1}_{\uparrow \text{smoother}}$$

PF: Write $u = \underbrace{S(t)(u_0^\omega, u_1^\omega)}_{=: Z^\omega} + v \hat{=} \int_0^t \frac{\sin(t-t')|\nabla|}{|v|} u^3(t') dt'$

$\Rightarrow v$ satisfies

$$\begin{cases} -\partial_t^2 v + \Delta v = (v + Z)^3 \\ (v, \partial_t v)|_{t=0} = 0 \end{cases}$$

Let $\Gamma v = - \int_0^t \frac{\sin(t-t')|\nabla|}{|v|} (v + Z)^3(t') dt'$.

$$\begin{aligned} \Rightarrow \| \Gamma v \|_{L_T^\infty \dot{H}^1} &\leq \left\| \int_0^t \| (v + Z)^3(t') \|_{L_x^2} dt' \right\|_{L_T^\infty} \\ &= \| (v + Z)^3 \|_{L_T^1 L_x^2} = \| v + Z \|_{L_T^3 L_x^6}^3 \end{aligned}$$

(5)

$$\lesssim \|v\|_{L_T^3 L_x^6}^3 + \|z\|_{L_T^3 L_x^6}^3$$

Sobolev

$$\lesssim T \|v\|_{L_T^\infty H^1}^3 + \|z\|_{L_T^3 L_x^6}^3$$

≤ 1 outside a set of
 prob. $\leq c e^{-\frac{c}{T^{\frac{1}{8}} \| (u_0, u) \|_{L^2}^2}$

• A difference estimate follows in a similar manner:

$$\| \Gamma v_1 - \Gamma v_2 \|_{L_T^\infty H^1} \lesssim T^{1/3} \left(T^{2/3} \|v_1\|_{L_T^\infty H^1}^2 + T^{2/3} \|v_2\|_{L_T^\infty H^1}^2 + \|z\|_{L_T^3 L_x^6}^2 \right) \|v_1 - v_2\|_{L_T^\infty H^1}$$

$\Rightarrow \Gamma^w$ is a contraction on $B_{(1)}$ in $L^\infty H^1$

for every $w \in \Omega_T$

• Also,

$$\|v(t)\|_{L_x^2} \stackrel{\text{Mink}}{\leq} \int_0^t \underbrace{\| \partial_{t'} v \|_{L_x^2}}_{\Gamma \text{ controlled}} dt'$$

FTC $\| \int_0^t \partial_{t'} v dt' \|_{L_x^2}$

$$\partial_t v = - \int_0^t \cos(t-t') |v| (v+z)^3 dt' \in C_T L_x^2$$

□

The same proof works for

(b)

$$\begin{cases} -\partial_t^2 v + \Delta v = (v + z^w)^3 \\ (v, \partial_t v)|_{t=0} = (v_0, v_1) \in \mathcal{H}' \end{cases}$$

by performing a fixed pt argument in $B_0(\|(v_0, v_1)\|_{\mathcal{H}'})$

$$\Gamma v = S(t)(v_0, v_1) - \int_0^t \frac{\sin(t-t')|\nabla|}{|t'|} (v+z)^3(t') dt'$$

Rmk: In terms of the original NLW,

we have a.s. LWP for

$$\begin{cases} -\partial_t^2 u + \Delta u = u^3 \\ (u, \partial_t u) = \underbrace{(v_0, v_1)}_{\in \mathcal{H}'} + \underbrace{(u_0^w, u_1^w)}_{\text{rough \& random}} \end{cases}$$

$0 < s \leq 1$

Thm 2.9: The defocusing cubic NLW is a.s. GWP
(with respect to (u_0^w, u_1^w) .)

⑦

For $s=0$, see Burq-Tzvetkov
JEMS '14

Pf: $E(u, \partial_t u) = \frac{1}{2} \int |\nabla u|^2 + \frac{1}{2} \int (\partial_t u)^2 + \frac{1}{4} \int u^4$

• Estimate the growth of $E(v(t))$
 $\hat{=} \text{in } H^1$

$$\Rightarrow \partial_t E(v) = \int \partial_t v (\underbrace{\partial_t^2 v - \Delta v + v^3}_{= - (v+z)^3}) dx$$

$$\stackrel{C-S}{\leq} \left(\int (\partial_t v)^2 dx \right)^{1/2} \left(\int (v^2 |z| + \underbrace{|v|z^2}_{\text{No } v^3 \text{ term}} + |z|^3) dx \right)^{1/2}$$

$$\begin{aligned} & z^{1/2} z^{3/2} \\ & \parallel \\ |v|z^2 & \lesssim v^2 |z| + |z|^3 \end{aligned}$$

$$\lesssim E(v) \left(1 + \|z\|_{L_{T,x}}^2 \right) + \|z\|_{L_T L_x}^6 \quad v^4 z^2$$

Gronwall
 \Rightarrow

$$E(v)(t) \lesssim \|z\|_{L_{T,x}}^6 e^{c \int_0^t (1 + \|z\|_{L_{T,x}}^2) dt'}$$

, $|t| \leq T$

Lec 12: 15/02/17 (Wed)

①

Our goal is to prove a.s. GWP.

⇐ suffices to prove "almost a.s. GWP."

$\forall T, \varepsilon > 0 \exists \Omega_{T, \varepsilon}$ with $P(\Omega_{T, \varepsilon}^c) < \varepsilon$

s.t. $\forall \omega \in \Omega_{T, \varepsilon}, \exists!$ soln u^ω with $(u^\omega, \partial_t u^\omega)|_{t=0} = (u_0^\omega, u_1^\omega)$
on $[-T, T]$

Fix $T, \varepsilon > 0$.

$$A_\lambda = \{ \omega \in \Omega : \|z^\omega\|_{L_{T, x}^6 \cap L_{T, x}^\infty} \leq \lambda \}$$

⇒ Can choose $\lambda = \lambda(\|(u_0, u_1)\|_{H^s}, T) \gg 1$ s.t.

$$P(A_\lambda^c) < \frac{\varepsilon}{2}$$

i.e. on $A_\lambda, E(v)(t) \leq C_T \quad \forall |t| \leq T.$

① Choose local existence time

②

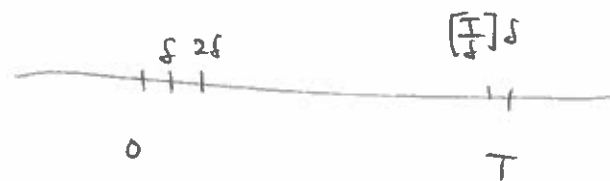
$$\delta = \delta \left(\underbrace{\sup_{|t| \leq T} \| (v, \partial_t v)(t) \|_{\mathcal{H}^1}}_{\leq C_T^{1/2}} \right) > 0$$

② By possibly making δ smaller, we have

$$P \left(\underbrace{\| z \|_{L_{I_j}^3 L_x^6}}_{\leq C_T^{1/2}} \geq 1 \right) \leq e^{-C/\delta^{2/3}}, \quad I_j = ((j-1)\delta, j\delta)$$

$$\Rightarrow \forall \omega \in A_\lambda \cap \bigcap_{j=-[\frac{T}{\delta}] - 1}^{[\frac{T}{\delta}] + 1} B_j^c,$$

$$\exists! u = u^\omega \text{ on } [-T, T]$$



$$\text{Lastly, } P \left(\left(\bigcap_j B_j \right)^c \right) \leq \sum_j P(B_j^c) \leq T \delta^{-1} e^{-C/\delta^{2/3}}$$

$< \varepsilon/2$ by choosing $\delta \ll 1$.

$$\delta(T, \varepsilon)$$

□

• Rmk: One actually need to work with $V_N \leftarrow$ smooth

(3)

$$\Leftrightarrow \text{soln to } -\partial_t^2 V_N + \Delta V_N = (V_N + \underbrace{Z_N^w})^3$$

to justify the switching of ∂_t and $\int dx = P_{\leq N} Z^w$.

in the probabilistic energy estimate (and let $N \rightarrow \infty$)

indep of N .

• $3 < p \leq 5$: $-\partial_t^2 u + \Delta u = \underbrace{|u|^{p-1} u}_{=: N(u)} \text{ on } \mathbb{R}^3 \text{ or } \mathbb{T}^3$

As before, set $u = z + v \quad =: N(u)$

$$\Rightarrow -\partial_t^2 v + \Delta v = N(v+z).$$

\uparrow only need some space-time control.

① a.s. LWP: Consider

$$\begin{cases} -\partial_t^2 v + \Delta v = N(v+f) \\ (v, \partial_t v)|_{t=t_0} = (v_0, v_1) \end{cases}$$

on $I \ni t_0$

④

$$\|(\Gamma v, \partial_t \Gamma v)\|_{C_I \dot{H}^1} + \|\Gamma v\|_{L_I^{2p/p-3} L_x^{2p}}$$

$$\lesssim \|(v_0, v_1)\|_{\dot{H}^1} + \|N(v+f)\|_{L_I^1 L_x^2}$$

$$\leq |I|^\theta \|v+f\|_{L_I^{2p/p-3} L_x^{2p}}$$

$$\lesssim \|(v_0, v_1)\|_{\dot{H}^1}$$

$$+ |I|^\theta \|v\|_{L_I^{2p/p-3} L_x^{2p}} + |I|^\theta \|f\|_{L_I^{2p/p-3} L_x^{2p}}$$

Strichartz pair
(S=1)
p factors

$$x: \frac{1}{2} = \frac{1}{2p} + \dots + \frac{1}{2p}$$

$$t: 1 = p \cdot \frac{p-3}{2p} + \theta$$

Need

$$\frac{p-3}{2} \leq 1 \Leftrightarrow p \leq 5$$

$$\theta = 0 \text{ when } p = 5$$

- p < 5: LWP on $|I| \leq \delta \sim \left(\|(v_0, v_1)\|_{\dot{H}^1} + \|f\|_{L_I^{2p/p-3} L_x^{2p}} \right)^{-(p-1)}$
- p = 5: δ depends on the profile of (v_0, v_1) .

② a.s. GWP: $3 < p \leq 5$: (p=5: Oh-Pocovnicu '16)

side computation: $\partial_t (|v|^{p+1}) = \partial_t (v^2)^{\frac{p+1}{2}} = (p+1) |v|^{p-1} v \partial_t v$

Let $E(v) = \frac{1}{2} \int (\partial_t v)^2 + \frac{1}{2} \int |\nabla v|^2 + \frac{1}{p+1} \int |v|^{p+1}$

$$\Rightarrow \partial_t E(v) = \int \partial_t v \left(\underbrace{\partial_t^2 v - \Delta v + |v|^{p-1} v}_{-|v+f|^{p-1}(v+f)} \right) \quad (5)$$

$$F(x) = |x|^{p-1} x$$

$$F'(x) = p|x|^{p-1}$$

$$F''(x) = p(p-1)|x|^{p-3} x$$

Taylor
 \Rightarrow

$$|v+f|^{p-1}(v+f) - |v|^{p-1}v = p|v|^{p-1}f + \frac{p(p-1)}{2}|v+\theta f|^{p-3}(v+\theta f) \cdot f^2.$$

$$\begin{aligned} \Rightarrow \partial_t E(v) &= -p \int \partial_t v \cdot |v|^{p-1} f - \frac{p(p-1)}{2} \int \partial_t v \underbrace{|v+\theta f|^{p-2}(v+\theta f) \cdot f^2}_{\approx |v|^{p-2} f^2 + f^p} \\ &=: \text{I} + \text{II} \end{aligned}$$

$$\cdot |\text{II}| \stackrel{C-S}{\lesssim} \left(\int (\partial_t v)^2 \right)^{1/2} \left(\|f(t)\|_{L_x^\infty}^4 \int |v|^{2(p-2)} + \|f(t)\|_{L^{2p}}^{2p} \right)^{1/2}$$

$$\lesssim \left(1 + \|f\|_{L_{T,x}^\infty}^{2-} \right) E(v)$$

$$+ \|f(t)\|_{L^{2p}}^{2p}$$

$$\frac{2(p-2) \leq p+1}{\Leftrightarrow p \leq 5}$$

6

$$\int_0^t I = - P \int_0^t \underbrace{\partial_t v |v|^{p-1} f}_{= \partial_t (|v|^{p-1} v)}$$

$$\stackrel{\text{IBP}}{=} - \int_x |v|^{p-1} v f \Big|_0^t \quad v(0) = 0$$

$$\left(\begin{array}{l} \text{Young} \\ \lesssim \\ \approx \end{array} \right) \underbrace{\varepsilon \int_x |v(t)|^{p+1}}_{\varepsilon \bar{E}(t) \leftarrow \text{hide it on LHS}} + \underbrace{\frac{1}{\varepsilon} \int_x |f(t)|^{p+1}}_{= \frac{1}{\varepsilon} \|f\|_{L_T^\infty L_x^{p+1}}^{p+1}}$$

$$+ P \int_0^t \underbrace{\int_x |v|^{p-1} v \partial_t f}_{=: \langle \nabla \rangle f}$$

① place $\langle \nabla \rangle^{s-1} \tilde{f}$ in $L_{T,x}^\infty$

$$\int_x \langle \nabla \rangle^{1-s} (|v|^{p-1} v) \langle \nabla \rangle^{s-1} \tilde{f}$$

$$\textcircled{2} \left(\int |v|^{(p-1) \frac{p+1}{p-1}} \right)^{\frac{p-1}{p+1}} \lesssim E^{\frac{p-1}{p+1}}$$

If $f = z$, then

$$\partial_t z = \langle \nabla \rangle \tilde{z}$$

$$\tilde{z} = \tilde{S}(t) (u_0^\omega, u_1^\omega)$$

$$= \frac{-|\nabla| \sin(t|\nabla|)}{\langle \nabla \rangle} u_0^\omega + \frac{\cos(t|\nabla|)}{\langle \nabla \rangle} u_1^\omega$$

Satisfies the same probabilistic Strichartz estimates

$$\textcircled{3} \quad \left(\int \langle \nabla \rangle^{1-s+} v \right)^{\frac{p+1}{2}} \Big|_{p+1}^{\frac{2}{p+1}}$$

low freq easy
 \Rightarrow Assume $\langle \nabla \rangle \sim |\nabla|$

$$= \| \langle \nabla \rangle^{1-s+} v \|_{L^{\frac{p+1}{2}}}^{\frac{2}{p+1}}$$

$$\lesssim \| \nabla v \|_{L^2}^\theta \| v \|_{L^{p+1}}^{1-\theta}$$

$$\frac{2}{p+1} = \frac{\theta}{2} + \frac{1-\theta}{p+1} \Rightarrow \theta = \frac{2}{p-1}$$

$$1-s+ = \theta \cdot 1 + (1-\theta) \cdot 0$$

$$s > 1-\theta = 1 - \frac{2}{p-1} = \frac{p-3}{p-1}$$

$p=3: s > 0$
 Burg-Tsvel '14

$p=5: s > 1/2$
 O-P '16

$3 < p < 5 = \text{Xia}$

Lührman-Meudelson (weaker)

$$\lesssim E^{\frac{\theta}{2} + \frac{1-\theta}{p+1}} \leq \frac{2}{p+1}$$

$$\Rightarrow \int_0^t I_2 \lesssim \| \langle \nabla \rangle^{s-} f \|_{L_{T,x}^\infty} \int_0^t E(t') dt'$$

\Rightarrow Apply Gronwall.

Lec 13: 20/02/17 (Mon)

①

• Dispersion: Failure of the endpoint Strichartz estimate (wave)

⊗ False: $\|S(t)(u_0, u_1)\|_{L_t^2 L_x^\infty} \lesssim \|(u_0, u_1)\|_{\dot{H}^1} \quad d=3$

$$(q, r, d) = (2, \infty, 3).$$

(Schrödinger: fails at $(2, \infty, \underline{2})$) \leftarrow ex: $\|f\|_{L_x^\infty(\mathbb{R}^2)} \not\lesssim \|f\|_{H^1(\mathbb{R}^2)}$

• Nonhomog lin wave eqn:

$$-\partial_t^2 u + \Delta u = F$$

$$(u, \partial_t u)|_{t=0} = (u_0, u_1) \quad \text{on } \mathbb{R}^{3+1}$$

$$\Rightarrow u(t, x) = \frac{\partial}{\partial t} \left[\int_{\mathbb{S}^2} u_0(x + t\omega) d\sigma(\omega) \right]$$

Kirchoff

$$+ t \int_{\mathbb{S}^2} u_1(x + t\omega) d\sigma(\omega)$$

$$+ \int_0^+ t \int_{\mathbb{S}^2} F(t-t', x+t'\omega) d\sigma(\omega) dt'$$

Idea: adaptation of Stein's counterexample for the spherical maximal function. (p469, Stein)

(2)

$$f \longmapsto \sup_{t>0} |A_t f| \quad \text{on } L^p$$

$$A_t f(x) = \int_{|y|=1} f(x-ty) d\sigma(y) = f * d\sigma_t$$

spherical average

$$\int g(x) d\sigma_t(x) = \int g(tx) d\sigma(x)$$

Bounded: $d \geq 2$, $p > \frac{d}{d-1}$

Fail: $d=1$, $p < \infty$
 $d \geq 2$; $p \leq \frac{d}{d-1}$

$d=1$: $A_t f(x) = \frac{1}{2} [f(x+t) + f(x-t)]$

f , positive, unbounded near 0.

$$\Rightarrow \sup_t A_t f(x) = \infty$$

but $f \in L^p(\mathbb{R})$

$$\underline{d \geq 2}: f(x) = \begin{cases} \frac{|x|^{1-d}}{\log 1/|x|} & \text{near } 0 \\ 0, & |x| \gg 1 \end{cases}$$

(3)

$\Rightarrow \forall x, A_t f(x)$ is unbdd near x .

$\Rightarrow \sup_t A_t f(x) = \infty$ but $f \in L^p(\mathbb{R}^d), p = \frac{d}{d-1}$.

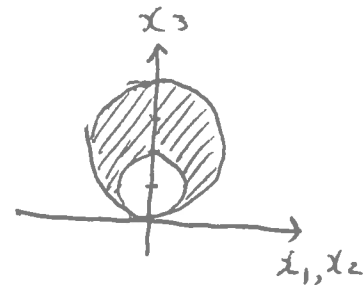
Back to wave Strichartz.

Prop 2.10: $u_0 = F = 0$.

$$u_1(x) = \frac{\mathbb{1}_A(x)}{|x|^2 (1 + \log|x|)^\alpha}, \quad \frac{1}{2} < \alpha \leq 1$$

d=3

$$A = B(2\hat{e}_3, 2) \setminus B(\hat{e}_3, 1)$$



Claim: $u_1 \in L^2$ but

$$u(t, t\hat{e}_3) = \infty, \quad \forall 1 < t < 2$$

Dirac soln

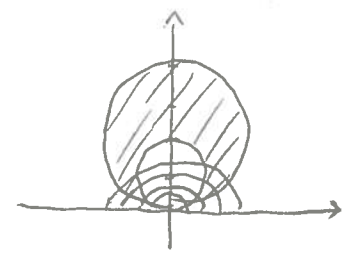
\Rightarrow (LHS) $\neq \infty \Rightarrow \text{(*) fails}$.

(4)

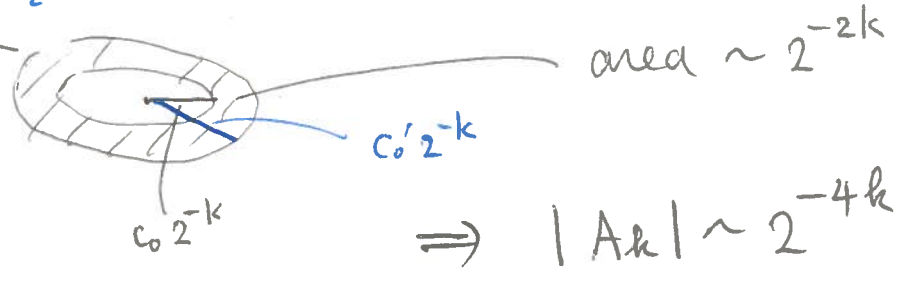
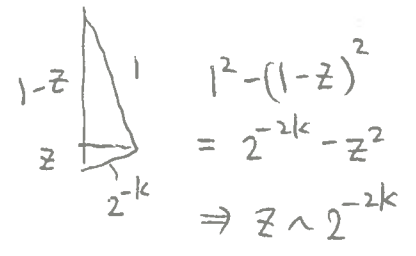
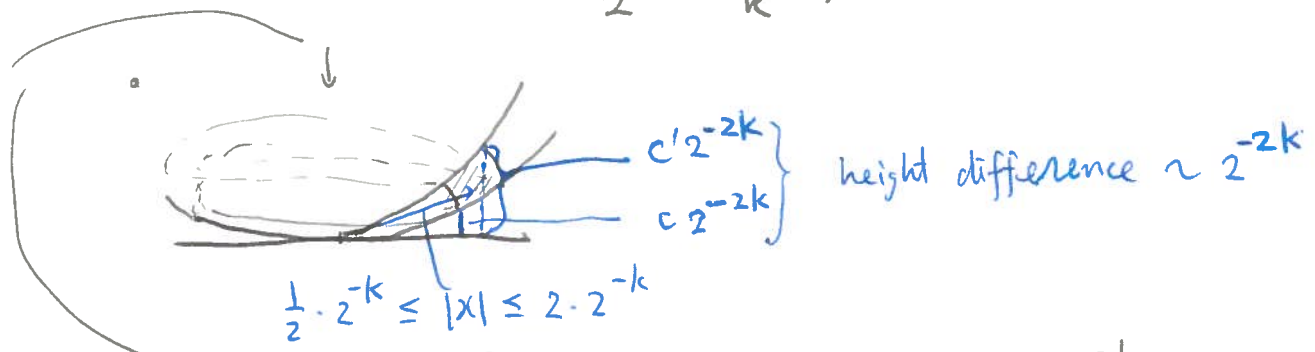
Portion of u_1 in the dyadic shell $\{|x| \sim 2^{-k}\} \cap A = A_k$ lives in a set of volume $\sim 2^{-4k}$ and has height $\sim \frac{2^{2k}}{k^\alpha}$

$$\|u_1\|_{L^2} \sim \left(\sum_k \left(\frac{2^{2k}}{k^\alpha} \right)^2 \times 2^{-4k} \right)^{1/2}$$

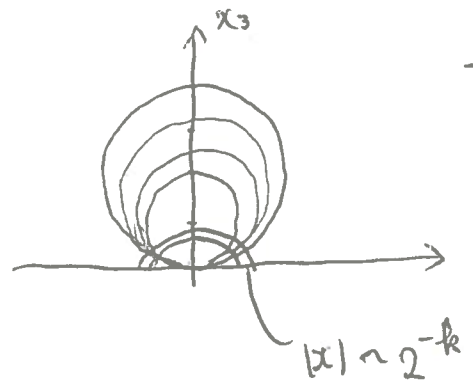
$$\sim \left(\sum \frac{1}{k^{2\alpha}} \right)^{1/2} < \infty, \quad \alpha > 1/2$$



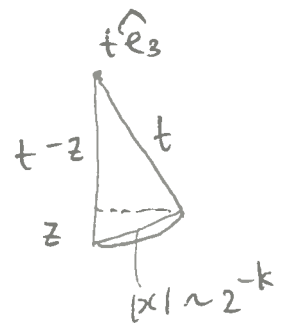
On A_k , $u_1(x) = \frac{1}{2^{-2k} k^\alpha}$



$U(t, x) = t \int_{S^2} U_1(x + t\omega) d\sigma(\omega)$



Take $x = t\hat{e}_3$, $1 < t < 2$

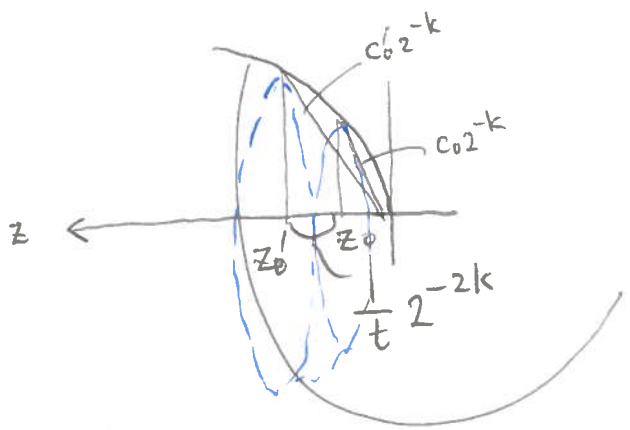


$$t^2 - (t-z)^2$$

$$\parallel$$

$$|x|^2 - z^2$$

$$\Rightarrow z \sim \frac{1}{t} 2^{-2k}$$



$f(z) = \sqrt{t^2 - (z-t)^2}$ \Leftarrow rotate around the z-axis

surface area = $\int_{z_0}^{z_0'} 2\pi f(z) \sqrt{1 + (f'(z))^2} dz \sim 2^{-2k}$

($\because 1 < t < 2$)

\Rightarrow contribution from A_k

$$2^{-2k} \times \frac{2^{2k}}{k^\alpha} = \frac{1}{k^\alpha} \notin l_k^1 \quad \alpha \leq 1$$

$$\Rightarrow u(t, t\hat{e}_3) = \infty, \quad 1 < t < 2.$$

(6)

Rmk: • localization in time does NOT help

• Assuming a higher regularity does NOT help

$$\|u\|_{L_t^2 L_x^\infty} \not\lesssim \|u\|_{H^s}, \quad s \gg 1.$$

⇐ The estimate is L^2 -scaling invariant.

• But if one assumes a higher regularity AND localization in time.

$$\Rightarrow \|u\|_{L_T^2 L_x^\infty} \lesssim C(\varepsilon, T) \|u\|_{H^\varepsilon}.$$

↳ $C_\varepsilon (\log T)^{1/2}, T \gg 1$ ← TT* argument.

↳ $C_\varepsilon T^\varepsilon, T \lesssim 1$ ← scaling the $T \gg 1$ case.

Q: What about replacing L_x^∞ by the BMO norm?

or localize $\text{supp } \hat{u}_i \subset \{|\xi| \lesssim 1\}$?

ANS: Both fail!!

BMO estimate \Rightarrow freq localized estimate.
(if true)

Suppose $\|S(t)(u_0, u_1)\|_{L_t^2 BMO_x} \lesssim \|(u_0, u_1)\|_{\dot{H}^1}$

Suppose u_0, u_1 are frequency localized.

\Rightarrow so is $u = S(t)(u_0, u_1)$

$\Rightarrow \|u(t, \cdot)\|_{BMO_x} \sim \|u(t, \cdot)\|_{L_x^\infty}$

i.e. LP proj of a BMO function is in L^∞ .

Goal: Disprove freq localized estimate. (based Montgomery-Smith '98 Duke)

Disprove

$$\left\| \int e^{it\sqrt{-\Delta}} F(t) dt \right\|_{L_x^2} \lesssim \|F\|_{L_t^2 L_x^1}$$

loc in spatial freq.
 $|\xi| \sim 1$

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$$\| e^{-it\sqrt{\Delta}} f \|_{L_t^2 L_x^\infty} \stackrel{?}{\lesssim} \| f \|_{L_x^2}$$

$$\sup_{\substack{F \in L_t^2 L_x^1 \\ \|F\|_{L_t^2 L_x^1} = 1}} | \langle e^{-it\sqrt{\Delta}} f, F \rangle_{L_{t,x}^2} |$$

$$= | \int \langle f, e^{it\sqrt{\Delta}} F \rangle dt |$$

$$= | \langle f, \int e^{it\sqrt{\Delta}} F dt \rangle_{L_x^2} |$$

$$\leq \| f \|_{L_x^2} \underbrace{\| \int e^{it\sqrt{\Delta}} F dt \|_{L_x^2}}_{\stackrel{?}{\lesssim} \| F \|_{L_t^2 L_x^1} \leq 1}$$

$$\stackrel{?}{\lesssim} \| F \|_{L_t^2 L_x^1} \leq 1$$

$$\text{supp } \hat{f}(\xi) \subset \{|\xi| \sim 1\}$$

\Downarrow

(can assume

$$\text{supp } \hat{F}(t, \xi) \subset \{|\xi| \sim 1\}$$

$\forall t$.

Goal: Disprove the freq. loc. dual estimate:

$$\textcircled{*} \quad \left\| \int e^{it\zeta_3} F(t) dt \right\|_{L_x^2} \lesssim \|F\|_{L_t^2 L_x^1}$$

$\beta_t =$ Brownian motion

$\beta_t - \beta_s =$ mean 0 Gaussian r.v.
with $\text{Var} = t - s$.

$$\mathbb{E} \left[e^{\pm 2\pi i (\beta_t - \beta_s) \zeta} \right]$$

$$= e^{-2\pi |t-s| \zeta^2}$$

$$\forall \zeta \in \mathbb{R}$$

$$\text{Let } F(t, x) = \eta\left(\frac{t}{T}\right) \Psi(x - t\hat{e}_3 - \beta_t \hat{e}_1)$$

\uparrow
non-neg. smooth cutoff
supp on $[-1, 1]$

$\hat{\Psi}(\zeta)$ supported
on $\{|\zeta| \sim 1\}$

Suppose $\textcircled{*}$ holds

$$\text{With } \hat{F}(t, \zeta) = \eta\left(\frac{t}{T}\right) e^{-2\pi i t \zeta_3} e^{-2\pi i \beta_t \zeta_1} \hat{\Psi}(\zeta)$$

$$(*) \Rightarrow \left\| \int e^{2\pi i t |\zeta|} e^{-2\pi i t \zeta_3} e^{-2\pi i \beta t \zeta_1} \eta\left(\frac{t}{T}\right) dt \cdot \hat{\psi}(\zeta) \right\|_{L^2_{\zeta}} \quad (2)$$

$$\lesssim \|F\|_{L_t^2 L_x^1} \sim \sqrt{T}.$$

$$(**) \int_{|\zeta| \sim 1} \left| \int_t \dots dt \right|^2 d\zeta \lesssim T$$

Let $A = \{ \zeta \in \mathbb{R}^3 : \zeta_3 \sim 1, |\zeta_2| \ll |\zeta_1| \ll 1 \}$.

\Rightarrow We have the following Taylor approx:

$$|\zeta| - \zeta_3 \sim \zeta_1^2, \quad \forall \zeta \in A.$$

$$\left(|\zeta| = \sqrt{\zeta_3^2 + \zeta_1^2 + \zeta_2^2} = \zeta_3 + \frac{1}{2|\zeta_3|} (\zeta_1^2 + \zeta_2^2) + \text{error} \right)$$

• Restrict the domain to A and expand the square.

(LHS) of $(**)$

(3)

$$\sim \int_A \iint_{t, t'} e^{2\pi i (t-t')(|\zeta| - \zeta_3)} \underline{e^{-2\pi i (\beta_t - \beta_{t'}) \zeta_1}} \eta\left(\frac{t}{T}\right) \eta\left(\frac{t'}{T}\right) dt' dt d\zeta$$

Take \mathbb{E}

$$\int_A \iint e^{2\pi i (t-t')(|\zeta| - \zeta_3)} e^{-2\pi |t-t'| \zeta_1^2} \eta\left(\frac{t}{T}\right) \eta\left(\frac{t'}{T}\right) dt' dt d\zeta$$

$$\lesssim T$$

let $r = t - t'$.

$$\int_A \int e^{2\pi i r (|\zeta| - \zeta_3)} e^{-2\pi |r| \zeta_1^2} \eta * \eta\left(\frac{r}{T}\right) dr d\zeta \lesssim 1$$

Let $T \rightarrow \infty$

$$\int_A \int e^{2\pi i r \underbrace{(|\zeta| - \zeta_3)}_x} e^{-2\pi |r| \underbrace{\zeta_1^2}_y} dr d\zeta \lesssim 1$$

$$\int_A \frac{1}{\zeta_1^2} d\zeta_1 d\zeta_2 d\zeta_3 \sim \infty$$

$$|\zeta_2| \ll |\zeta_1| \ll 1 \rightarrow \int \frac{1}{\zeta_1} d\zeta_2 = \infty$$

$$\left(\int e^{2\pi i r x} e^{-2\pi |r| y} dr = \frac{y}{\pi(x^2 + y^2)} \right)$$

$\Rightarrow \otimes$ contradiction.

Rmk: Without the BM factor, there would be NO $e^{-2\pi|\eta|\xi_1^2}$ in the last integral.

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$$\Rightarrow \int_A \delta_0(\xi_1^2) d\xi \lesssim 1 \quad (\Leftarrow \text{No contradiction})$$

Recall $\|f\|_{BMO} = \sup_Q \frac{1}{|Q|} \int_Q |f - f_Q|$, $f_Q = \text{ave}_Q f$.

Lemma 2.11: $\text{supp } \hat{f} \subset \{|\xi| \sim 2^k\}$, $k \in \mathbb{Z}$ (homog LP localization)

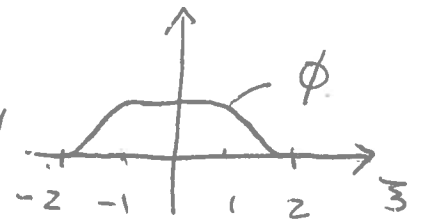
$$\Rightarrow \|f\|_{BMO} \sim \|f\|_{L^\infty}$$

Pf: By scaling, we may assume $k=0$.

• Fix x .

$$j \gg 1. \quad f(x) = P_{\leq j} f(x) = \int f(x-y) 2^{jd} \check{\Phi}(2^j y) dy$$

$$\Phi(\xi) = \begin{cases} 1, & |\xi| \leq 1 \\ 0, & |\xi| \geq 2 \end{cases}$$



We also have

$$\int f(x-y) 2^{-jd} \Psi(2^{-j}y) dy = P_{\leq -j} f(x) = 0.$$

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$$\Rightarrow f(x) = \int f(x-y) \left(2^{jd} \Psi(2^j y) - 2^{-jd} \Psi(2^{-j} y) \right) dy$$

↑ proj onto $\{|z| \leq 2^{-j}\}$

Let $Q = \{y: |y| \leq 2^\alpha\}$, and $Q_x = x + Q$.

Note that

$$\begin{aligned} & \int f_{Q_x} (2^{jd} \Psi(2^j y) - 2^{-jd} \Psi(2^{-j} y)) dy \\ &= f_{Q_x} \left[\underbrace{\int 2^{jd} \Psi(2^j y) dy}_{= \int \Psi dy} - \underbrace{\int 2^{-jd} \Psi(2^{-j} y) dy}_{= \int \Psi dy} \right] = 0 \end{aligned}$$

$$\Rightarrow f(x) = \int (f(x-y) - f_{Q_x}) (2^{jd} \Psi(2^j y) - 2^{-jd} \Psi(2^{-j} y)) dy$$

• If $y \in Q$, then $|f(x)| \lesssim 2^{d\alpha} \|f\|_{BMO}$.

• If $y \notin Q$, then $|f(x)| \leq 2 \|f\|_{L^\infty} \underbrace{\int_{|y| \geq 2^\alpha} (|2^{jd} \Psi(2^j y)| + |2^{-jd} \Psi(2^{-j} y)|) dy}_{< \frac{1}{4} \text{ by choosing } \alpha \gg 1 \text{ (index of } x \text{)}}$

$$\Rightarrow |f(x)| \leq C_\alpha \|f\|_{BMO} + \frac{1}{2} \|f\|_{L^\infty}$$

\Rightarrow Take sup in x .

□

(j is fixed)

Chap 3: Invariant Gibbs measure, Part 2

(6)

We'll focus on \mathbb{T}^2 .

Sec 3.1: Gibbs meas on \mathbb{T}^2 (for NLS, NLW)

$$H(u) = \frac{1}{2} \int |\nabla u|^2 \pm \frac{1}{p+1} \int |u|^{p+1}$$

$$\Rightarrow "d\mu = Z^{-1} e^{-H(u)} du \\ = Z^{-1} e^{\mp \frac{1}{p+1} \int |u|^{p+1}} df_1"$$

where $df_1 = Z^{-1} e^{-\frac{1}{2} \|u\|_{H^1}^2} du \leftarrow$ supported on $H^s(\mathbb{T}^2) \setminus L^2(\mathbb{T}^2)$

$$\Rightarrow \int |u|^{p+1} dx = \infty, \text{ a.s.} \quad s < 0.$$

• We need to renormalize $\int |u|^{p+1}$.

We only consider the defocusing case.

(Focusing case: Brydges - Slade '96.

We can not construct Gibbs measure for the focusing cubic NLS on \mathbb{T}^2 even if we consider the Wick ordered nonlinearity (with the Wick ordered $1^2 - \text{cutoff}$)

In the following, we focus on the real-valued setting

(7)

- Under P , we have \swarrow indep, std Gaussian r.v. (for simplicity)

$$u = \sum_{n \in \mathbb{Z}^2} \frac{g_n(\omega)}{\langle n \rangle} e^{in \cdot x}, \quad \begin{aligned} g_{-n} &= \overline{g_n} \\ \text{Var}(g_n) &= 1 \end{aligned}$$

Given $N \in \mathbb{N}$, let $u_N = P_{\leq N} u$.

\Rightarrow For each $x \in \mathbb{T}^2$, $u_N(x)$ is a mean 0 \mathbb{R} -valued Gaussian r.v. with variance:

$$\sigma_N := \mathbb{E}[u_N^2(x)] = \sum_{|n| \leq N} \frac{1}{\langle n \rangle^2} \sim \log N \uparrow \infty$$

Similarly,

$$\mathbb{E} \left[\int_{\mathbb{T}^2} u_N^2(x) dx \right] = \sum_{|n| \leq N} \frac{1}{\langle n \rangle^2} \sim \log N$$

but defining $: u_N^2(x) :$ $\stackrel{\text{def}}{=} u_N^2(x) - \sigma_N$

Wick ordered monomial

$$= H_2(u_N(x); \sigma_N)$$

\nwarrow Hermite poly $H_2(x; \sigma) = x^2 - \sigma$

We have

$$\int_{\mathbb{T}^2} : u_N^2(x) : = \sum_{|n| \leq N} \frac{|g_n|^2 - 1}{\langle m \rangle^2}$$

(8)

$$\cdot \mathbb{E} \left[\int_{\mathbb{T}^2} : u_N^2 : dx \right] = 0$$

$$\cdot \mathbb{E} \left[\left(\int_{\mathbb{T}^2} : u_N^2 : dx \right)^2 \right] = \mathbb{E} \left[(g_0^2 - 1)^2 \right]$$

$$+ 2 \mathbb{E} \sum_{n \in I} \frac{(g_n^2 - 1)^2}{\langle m \rangle^4}$$

" $I = \mathbb{Z}^2/2$ "

$$\lesssim 1 + \sum_{\substack{|n| \leq N \\ |m| \leq N}} \frac{1}{\langle m \rangle^4} \lesssim 1 \text{ indep of } N$$

$$\mathbb{E} \left[(|g_n|^2 - 1)(|g_m|^2 - 1) \right] = 0 \text{ unless } n = m.$$

$\Rightarrow \int_{\mathbb{T}^2} : u_N^2 : dx$ is a well-defined r.v.

$$\int_{\mathbb{T}^2} : u^2 : dx = \lim_{N \rightarrow \infty} \int : u_N^2 : dx \text{ exists in } L^q(\Omega)$$

for any $q < \infty$.

Lec 15 27/02/17 (Mon)

①

Sec 3.2: Hermite polynomials, white noise functional,
and Wick ordering

- Hermite polynomials $H_k(x; \sigma)$ defined through the generating function:

$$\begin{aligned} G(t, x; \sigma) &= e^{tx - \frac{1}{2}\sigma t^2} \\ &= \sum_{k=0}^{\infty} \frac{t^k}{k!} H_k(x; \sigma) \end{aligned}$$

$$H_0(x; \sigma) = 1, \quad H_1(x; \sigma) = x,$$

$$H_2(x; \sigma) = x^2 - \sigma, \quad H_3(x; \sigma) = x^3 - 3\sigma x,$$

$$H_4(x; \sigma) = x^4 - 6\sigma x^2 + 3\sigma^2.$$

- $\partial_x H_k(x; \sigma) = k H_{k-1}(x; \sigma)$

• White noise functional;

(2)

$$d\mathcal{P}_0 = Z^{-1} e^{-\frac{1}{2} \|u\|_{L^2}^2} du \quad \text{on } H^{-\frac{d}{2}-\varepsilon}(\mathbb{T}^d)$$

$$\Leftrightarrow u_w(x) = \sum_{n \in \mathbb{Z}^d} g_n(\omega) e^{inx} \Leftrightarrow \text{distribution.}$$

• $W(\cdot) : L^2(\mathbb{T}^d) \rightarrow L^2(\Omega)$

by $W_f(\omega) = \langle f, u_w(\omega) \rangle_{L^2} = \sum_{n \in \mathbb{Z}^d} \hat{f}(n) \bar{g}_n(\omega)$

$\Rightarrow W_f$ is a Gaussian r.v. with mean 0.

$$\begin{aligned} \text{Var}(W_f) &= \mathbb{E} \left[\sum_n \hat{f}(n) \bar{g}_n \quad \overline{\sum_m \hat{f}(m) \bar{g}_m} \right] \quad \text{and} \quad g_{-n} = \bar{g}_n \\ &= \|f\|_{L^2}^2 \end{aligned}$$

Moreover, $\mathbb{E}[W_f \bar{W}_h] = \langle f, h \rangle_{L^2}$ deterministic

$W(\cdot)$ is unitary. (\sim Wiener integral $\int f dB$)

Lemma 3.1: $f, h \in L^2(\mathbb{T}^2)$ with $\|f\|_{L^2} = \|h\|_{L^2} = 1$ ③

Then, $\mathbb{E} [H_k(W_f) H_m(W_h)] = \delta_{km} k! [\langle f, h \rangle_{L^2_x}]^k$

Rmk: In the complex-valued setting, we use the (generalized) Laguerre polynomials (Oh-Thomann '15)

Pf: First, recall

$$\int_{\Omega} e^{W_f} dP = e^{\frac{1}{2} \|f\|_{L^2}^2}$$

\tilde{I} = index set
= " $\mathbb{Z}^2/2$ "

= $\mathbb{Z} \times \mathbb{N} \cup \mathbb{N} \times \{0\}$
 $\cup \{(0,0)\}$

$I^* = I \setminus \{(0,0)\}$

$$\begin{aligned} \text{(LHS)} &= \prod_{n \in I^*} \frac{1}{\pi} \int_{\mathbb{C}} e^{z \operatorname{Re}(\hat{f}_n \bar{g}_n) - |g_n|^2} dg_n \\ &\quad \times \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{\hat{f}_0 g_0} e^{-\frac{1}{2} g_0^2} dg_0 \\ &= \prod_{n \in I^*} \frac{1}{\pi} \int_{\mathbb{R}} e^{z \operatorname{Re} \hat{f}_n \operatorname{Re} g_n - (\operatorname{Re} g_n)^2} d\operatorname{Re} g_n \\ &\quad \times \int_{\mathbb{R}} e^{z \operatorname{Im} \hat{f}_n \operatorname{Im} g_n - (\operatorname{Im} g_n)^2} d\operatorname{Im} g_n \end{aligned}$$

$$\left(\times \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} \dots dg_0 \right) = e^{\sum_{n \in \mathbb{I}^*} |\hat{f}^{(n)}|^2} e^{\frac{1}{2} (\hat{f}^{(0)})^2} = e^{\frac{1}{2} \|f\|_{L^2}^2}$$

$$\begin{aligned} \Rightarrow & \int_{\Omega} G(t, W_f) G(s, W_h) dP \\ &= e^{-\frac{1}{2}(t^2+s^2)} \int_{\Omega} e^{W_t f + s h} dP \\ & \quad \uparrow \text{by linearity of } W(\cdot) \quad \underbrace{\hspace{10em}}_{= e^{\frac{1}{2} \|tf + sh\|_{L^2}^2}} \\ &= e^{ts \langle f, h \rangle_{L^2}} = \sum_{l=0}^{\infty} \frac{(ts)^l}{l!} (\langle f, h \rangle)^l. \end{aligned}$$

On the other hand,

$$(LHS) = \sum_{k, m=0}^{\infty} \frac{t^k s^m}{k! m!} \int_{\Omega} H_k(W_f) H_m(W_h) dP$$

□

Under P_1 , we have

$$u = \sum_{n \in \mathbb{Z}^2} \frac{g_n}{\langle n \rangle} e^{in \cdot x}$$

Let $U_N = P_{\leq N} u$.

$\Rightarrow U_N(x)$ is a Gaussian r.v. with mean 0
 $\forall x \in \mathbb{T}^2$. and $\text{var} \sim \log N$
 σ_N

\Rightarrow Define the Wick ordered monomial

$$\underline{: U_N^k(x) :} = H_k(U_N(x); \sigma_N)$$

$$\text{Let } G_N(u) = \frac{1}{2^m} \int_{\mathbb{T}^2} : (P_{\leq N} u)^{2m} : dx$$

Prop 3.2: $\{G_N(u)\}$ is a Cauchy seq in $L^q(P_1)$

for any finite $q \geq 2$. Moreover,

$$\|G_M(u) - G_N(u)\|_{L^q(P_1)} \leq C_m (q-1)^m \frac{1}{N^{1/2}}$$

for any $q \geq 1$, $M \geq N \geq 1$.

Rmk: Prop 3.2 allows us to define

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$$G(u) = G_\infty(u)$$

$$= \frac{1}{2m} \int_{\mathbb{T}^2} : u^{2m} : dx = \lim_{N \rightarrow \infty} G_N(u) \text{ in } L^q(\beta).$$

and the Wick ordered Hamiltonian

$$H_{\text{Wick}}(u) = \frac{1}{2} \int_{\mathbb{T}^2} |\nabla u|^2 + \frac{1}{2m} \int_{\mathbb{T}^2} : u^{2m} :$$

Pf ($q=2$): $e_n(y) = e^{in \cdot y}$

$$\psi_N(x)(\cdot) = \frac{1}{\sigma_N^{1/2}} \sum_{|n| \leq N} \frac{e_n(x)}{\langle n \rangle} e_n(\cdot)$$

$$\delta_N(\cdot) = \sum_{|n| \leq N} \frac{1}{\langle n \rangle^2} e_n(\cdot)$$

$$\sigma_N = \sum_{|n| \leq N} \frac{1}{\langle n \rangle^2} \sim \log N$$

Note that

$$\|\psi_N(x)\|_{L^2(\mathbb{T}^2)} = 1 \text{ for each fixed } x \in \mathbb{T}^2.$$

and

$$(*) \quad \langle \psi_M(x), \psi_N(y) \rangle_{L^2(\mathbb{T}^2)} = \frac{1}{\sigma_M^{1/2} \sigma_N^{1/2}} \delta_N(y-x), \quad M \geq N \geq 1.$$

$$\begin{aligned} \bullet U_N(x) &= \sigma_N^{1/2} \frac{U_N(x)}{\sigma_N^{1/2}} = \sigma_N^{1/2} \overline{W_{\eta_N(x)}} \\ &= \sigma_N^{1/2} W_{\eta_N(x)} \end{aligned}$$

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$$\Rightarrow \circ U_N^{2m}(x) = H_{2m}(U_N(x); \sigma_N)$$

$$H_k(x; \sigma) = \sigma^{k/2} H_k(x; 1)$$

$$= \sigma_N^m H_{2m}\left(\frac{U_N(x)}{\sigma_N^{1/2}}; 1\right)$$

$$= \sigma_N^m H_{2m}(W_{\eta_N(x)})$$

\uparrow L^2 -norm is 1.

$$\Rightarrow (2m)^2 \|G_M(u) - G_N(u)\|_{L^2(\beta_i)}^2$$

$$= \int_{\Pi_x^2 \times \Pi_y^2} \int_{\Omega} \sigma_M^{2m} H_{2m}(W_{\eta_M(x)}) H_{2m}(W_{\eta_M(y)})$$

$$- \sigma_M^m \sigma_N^m H_{2m}(W_{\eta_M(x)}) H_{2m}(W_{\eta_N(y)})$$

$$- \sigma_N^m \sigma_M^m H_{2m}(W_{\eta_N(x)}) H_{2m}(W_{\eta_M(y)})$$

$$+ \sigma_N^{2m} H_{2m}(W_{\eta_N(x)}) H_{2m}(W_{\eta_N(y)}) dP dx dy$$

Lemma 3.1 & (*)

$$= C_m \int_{\mathbb{T}_x^2 \times \mathbb{T}_y^2} (\gamma_M(y-x))^{2m} - (\gamma_N(y-x))^{2m} dx dy$$

(8)

$$= C_m \int_{\mathbb{T}^2} (\gamma_M(x))^{2m} - (\gamma_N(x))^{2m} dx$$

$$\leq \int_{\mathbb{T}^2} |\gamma_M(x) - \gamma_N(x)| \left(|\gamma_M(x)|^{2m-1} + |\gamma_N(x)|^{2m-1} \right) dx$$

$$\stackrel{C-S}{\leq} \underbrace{\|\gamma_M - \gamma_N\|_{L^2}} \left(\|\gamma_M^{2m-1}\|_{L^2} + \|\gamma_N^{2m-1}\|_{L^2} \right)$$

$$= \left(\sum_{N \leq |m| \leq M} \frac{1}{\langle m \rangle^4} \right)^{1/2} \approx \frac{1}{N}$$

$$\cdot \|\gamma_M^{2m-1}\|_{L^2} = \|\gamma_M\|_{L^{4m-2}}^{2m-1} \stackrel{H-Y}{\leq} \left(\sum_{|m| \leq M} \frac{1}{\langle m \rangle^{2 \cdot \frac{4m-2}{4m-3}}} \right)^{\frac{4m-3}{2}} \leq C_m < \infty$$

indep of M.

□

01 / 03 / 17 (Wed)

①

Sec 3.3: Hypercontractivity of the Ornstein-Uhlenbeck semigroup
and Wiener chaos estimate

$$H = L^2(\mathbb{R}^d, \mu_d)$$

\uparrow standard Gaussian meas on \mathbb{R}^d .

$$L = \Delta - x \cdot \nabla$$

= Hontree - Fock operator (OU operator)

Consider

$$\partial_t u = Lu \quad \text{on } \mathbb{R}^d.$$

Lemma 3.3 (hypercontractivity of OU semigroup) Nelson '65
 $p > 1, q \geq 1$. Then, we have

$$\|e^{tL} f\|_{L^q(\mathbb{R}^d, \mu_d)} \leq \|f\|_{L^p(\mathbb{R}^d, \mu_d)}$$

for $t \geq \frac{1}{2} \log\left(\frac{q-1}{p-1}\right)$

Rmk: • The estimate holds indep of dim $d \geq 1$
(even $d = \infty$.)

(2)

• This is equivalent to Log Sobolev inequality.

Define a homog Wiener chaos of order k

$$HI_{\vec{k}}(x) = \prod_{j=1}^d H_{k_j}(x_j) \quad x = (x_1, \dots, x_d)$$

$|\vec{k}| = k_1 + \dots + k_d$

$\mathcal{H}_k =$ closure of homog Wiener \vec{k}
chaoses of order k under $L^2(\mathbb{R}^d, \mu_d)$
(or any $L^p(\mathbb{R}^d, \mu_d)$.)

Itô - Wiener decomposition,

$$L^2(\mathbb{R}^d, \mu_d) = \bigoplus_{k=0}^{\infty} \mathcal{H}_k$$

• $F \in H_k$ is an eigenfunction of L with eigenvalue $-k$.

⇒ Cor 3.4: $F \in H_k$. Then for $q \geq 2$, we have

$$\|F\|_{L^q(\mathbb{R}^d, \mu_d)} \leq (q-1)^{k/2} \|F\|_{L^2(\mathbb{R}^d, \mu_d)}$$

($\Leftarrow p=2, t = \frac{1}{2} \log(q-1)$ in Lemma 3.3)

Lemma 3.5: $S_k(\omega) = \sum_{\Gamma(k,d)} C(m_1, \dots, m_k) \underbrace{g_{n_1}(\omega) \dots g_{n_k}(\omega)}_{\in \bigoplus_{j=0}^k H_j}$

where $\Gamma(k,d) = \{ (m_1, \dots, m_k) \in \mathbb{Z}^k : |n_j| \leq d \}$

⇒ Then, for $q \geq 2$

$$\|S_k\|_{L^q(\Omega)} \leq C \sqrt{k+1} (q-1)^{k/2} \|S_k\|_{L^2(\Omega)}$$

↑ can be removed

(\Leftarrow Cor 2.4 and $x^j = \sum_{m=0}^{[j/2]} c_{m,j} \sigma^m H_{j-2m}(x; \sigma)$)

$\{g_n\}$
std indep Gaussian r.v.'s

Thomann-Tzetkov

ex: $n \neq m$

$$g_n^2 g_m^3 = (g_n^2 - 1 + 1) (g_m^3 - 3g_m + 3g_m)$$

$$\Rightarrow g_n^2 g_m^3 = \underbrace{H_2(g_n) H_3(g_m)}_{\in \mathcal{H}_5} + \underbrace{H_0(g_n) H_3(g_m)}_{\in \mathcal{H}_3} + \underbrace{3H_2(g_n) H_1(g_m)}_{\in \mathcal{H}_3} + \underbrace{3H_0(g_n) H_1(g_m)}_{\in \mathcal{H}_1}$$

Back to Prop 3.2:

$$u_N = \sum_{|n| \leq N} \frac{g_n}{\langle n \rangle} e^{in \cdot x}$$

$$u_N^{2m} = \sum_{|n| \leq N} e^{in \cdot x} \left(\sum_{\substack{n = n_1 + \dots + n_{2m} \\ |n_j| \leq N}} \prod_{j=1}^{2m} \frac{g_{n_j}}{\langle n_j \rangle} \right) \leftarrow 2m \text{ fold product of Gaussian}$$

$$G_N(u) = \frac{1}{2m} \int : u_N^{2m} : dx \in \bigoplus_{k=0}^{2m} \mathcal{H}_k \quad (\text{actually in } \mathcal{H}_{2m})$$

→ projection onto \mathcal{H}_{2m}

⇒ Use Lemma 3.5 and Prop 3.2 with $q=2$.

Namely,

$$\|G_M(u) - G_N(u)\|_{L^q(\rho_i)} \leq C_m (q-1)^m \|G_M(u) - G_N(u)\|_{L^2(\rho_i)} \lesssim \frac{1}{N^{1/2}}$$

□

Sec 3.4: Construction of the Gibbs measure on \mathbb{T}^2

Goal: Construct

$$\begin{aligned} d\mu &= Z^{-1} e^{-H_{\text{wick}}(u)} du \\ &= Z^{-1} e^{-\frac{1}{2m} \int_{\mathbb{T}^2} :u^{2m}: dx} d\rho_i \end{aligned}$$

($\overset{\text{deg}}{\downarrow} \Phi_2^{2m}$ or φ_2^m Euclidean QFT)
 \uparrow dim

$$\text{Let } R_N(u) = e^{-G_N(u)} = e^{-\frac{1}{2m} \int :u_N^{2m}: dx}$$

Prop 3.6:
• $R_N(u) \in L^q(\rho_i)$, for any finite $q \geq 1$,
with a unif bd in N (depends on q)
• $R_N(u)$ converges to some $R(u)$ in $L^q(\rho_i)$

• This allows us to define the Gibbs meas

(6)

$$d\mu = Z^{-1} R(u) d\rho_1 = Z^{-1} e^{-\frac{1}{2m} \int : u^{2m} : dx} d\rho_1$$

$$= \lim_{N \rightarrow \infty} \underbrace{Z_N^{-1} R_N(u) d\rho_1}_{= d\mu_N}$$

↑
"unif" conv.

PF: Nelson's estimate. Next time.

Lec 17 06/03/17 (Mon)

①

Pf of Prop 3.6: $G_N(u)$ is not sign definite

Nonetheless, the defocusing nature plays an important role.

Main observation: - $G_N(u)$ has a logarithmic upper bound.

$$\begin{aligned} -G_N(u) &= -\frac{1}{2m} \int_{\mathbb{R}^2} \underbrace{H_{2m}(u_N; \sigma_N)}_{=} dx \\ &= \sigma_N^m H_{2m}\left(\frac{u_N}{\sigma_N^{1/2}}\right) \end{aligned}$$

$$\leq \underline{b_m (\log N)^m}$$

$$H_{2m}(x; 1) \geq \underbrace{-a_m}_{>0}$$

$$\begin{aligned} \cdot \frac{\|R_N(u)\|_{L^q(\rho_i)}^q}{e^{-G_N(u)}} &= \int_0^\infty \rho_i(e^{-q G_N(u)} > \alpha) d\alpha \\ &\leq 1 + \int_1^\infty \rho_i(-q G_N(u) > \log \alpha) d\alpha \end{aligned}$$

⊗ \Rightarrow Suffices to show $\rho_i(-q G_N(u) > \log \alpha) \leq C \alpha^{-(1+\delta)}$, $\forall \alpha \geq 1, N \in \mathbb{N}$

(2)

Given $\lambda = \log \alpha$,

Choose $N_0 \in \mathbb{R}$ s.t. $\lambda = 2q \text{bm} (\log N_0)^m$.

Then, by the log upper bound,

$$\underbrace{P_i(-q G_N(u) > \lambda)} = 0 \quad \text{for all } N < N_0$$

$$\leq q \text{bm} (\log N)^m \leq \frac{1}{2} \lambda.$$

For $N \geq N_0$,

$$P_i(-q G_N(u) > \lambda) \leq P_i(-q G_N(u) + q G_{N_0}(u) > \lambda - \overbrace{q \text{bm} (\log N_0)^m}^{= \frac{1}{2} \lambda})$$

$$-q \text{bm} (\log N_0)^m - q G_{N_0}(u) \leq 0.$$

$$\leq P_i(q |G_N(u) - G_{N_0}(u)| \geq \frac{1}{2} \lambda)$$

Prop 3.2 & Chebyshev

$$\leq C_{m,q} e^{-c N_0^{1/2m} \lambda^{1/m}}$$

$N \geq N_0$

$$\ll e^{-(1+\delta)\lambda} \sim \alpha^{-(1+\delta)} \Rightarrow (*)$$

In $(**)$, we used

$$\begin{aligned} P_i (|G_M(u) - G_N(u)| > \tilde{\lambda}) \\ \leq C_m e^{-C_m N^{\frac{1}{2}m} \tilde{\lambda}^{\frac{1}{m}}} \end{aligned}$$

$G_N(u)$ converges in measure

$\Rightarrow R_N = e^{-G_N(u)}$ converges in measure

By repeating the argument in Cor 1.4,

$$R_N \rightarrow R = e^{-G(u)} \text{ in } L^q(P_i), \quad \forall q \geq 1$$

□

(3)

Sec 3.5: On the dynamical problem.

(4)

- Wick ordered NLS (WNLS):

$$i\partial_t u + \Delta u = : |u|^{2(m-1)} u : \quad \text{on } \mathbb{T}^2$$

- Wick ordered NLW (NLKG)

$$-\partial_t^2 u + \Delta u - u = : u^{2m-1} : \quad \text{on } \mathbb{T}^2.$$

-
- NLS: Scaling critical regularity $s_{crit} = 1 - \frac{1}{m-1}$ ($m \geq 2$)

cubic: $s_{crit} = 0$

quintic: $s_{crit} = 1/2$

but ρ_1 and μ are supported on

$$H^{-\epsilon}(\mathbb{T}^2) \setminus L^2(\mathbb{T}^2)$$

$$u = z + v$$

- Bourgain 196: The cubic WNLS is almost locally locally well-posed w.r.t. ρ_1 (and hence μ)

"inv. meas argument"
in Chap 1

(Also, see Colliander-Oh, Duke '12)

\Rightarrow a.s. GWP & invariance of μ .

When $m \geq 3$, i.e. (super)quintic

(5)

\Rightarrow regularity gap $> \frac{1}{2} \Leftarrow$ too large!!

Compactness argument: (on measures on spacetime functions)

$$(FWNLS) \quad i\partial_t u^N + \Delta u^N = F_N(u^N),$$

$$\text{where } F_N(u) = P_{\leq N} \left(: |P_{\leq N} u|^{2(m-1)} P_{\leq N} u : \right)$$

$$= (-1)^{m-1} (m-1)! \sigma_N^{m-1} \cdot P_{\leq N} \left(L_{m-1}^{(u)} \left(\frac{|u|^2}{\sigma_N} \right) u_N \right)$$

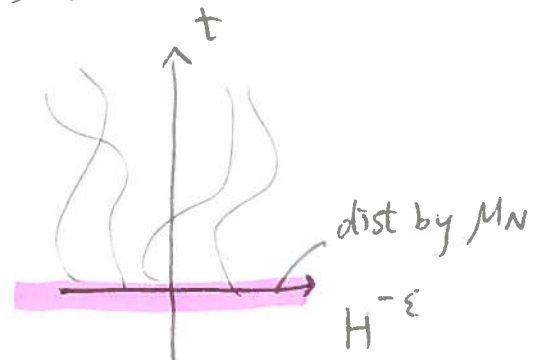
generalized Laguerre poly

$$\cdot d\mu_N = Z_N^{-1} R_N(u) dP,$$

= invariant Gibbs meas for (FWNLS).

① Extend $\mu_N =$ prob. meas on initial data

to $\nu_N =$ prob. meas on space-time functions



• (FWNLS) is globally well-posed
a.s.

(6)

$$\Rightarrow \exists \Phi_N : u_0 \in H^{-\varepsilon} \mapsto u^N \in C(\mathbb{R}; H^{-\varepsilon})$$

• Define a prob. meas ν_N on $C(\mathbb{R}; H^{-\varepsilon})$

$$\text{by } \underline{\nu_N = \mu_N \circ \Phi_N^{-1}}$$

= induced prob meas under Φ_N .

• Note that by invariance of μ_N ,

$$\underline{L(u^N(t)) = \mu_N} \text{ for any } t \in \mathbb{R}$$

\uparrow law

(2) (By a soft argument), show $\{\nu_N\}$ is tight (= precompact)

Prokhorov

$$\Rightarrow \underline{\nu_{N_j} \rightarrow \nu}$$

(3) Skorokhod's Thm: \exists another prob. space $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{\mathbb{P}})$

$\tilde{u}^{N_j}, C(\mathbb{R}; H^{-\varepsilon})$ -valued r.v.'s

$u, \quad =$

s.t. $L(\tilde{u}^{N_j}) = L(u^{N_j}) = \nu_{N_j}$

(7)

$$L(u) = \nu$$

and \tilde{u}^{N_j} converges to u in $C(\mathbb{R}; H^{-\varepsilon})$ almost surely.

We "upgrade" weak conv to a.s. conv.

Rmk: $L(\tilde{u}^{N_j}(t)) = L(u^{N_j}(t)) = \mu_{N_j}$

• By $\mu_N \xrightarrow{\text{"unif"}} \mu$, $L(u(t)) = \mu$ = Gibbs meas

Oh-Thomann '15

THM 3.7: \exists a set Σ of full probability s.t.

$\forall \phi \in \Sigma$, \exists global soln $u \in C(\mathbb{R}; H^{-\varepsilon})$ to WNLS
with $u|_{t=0} = \phi$.

Moreover, $L(u(t)) = \mu$, $\forall t \in \mathbb{R}$

Rmk: ① only a.s. global existence (NO uniqueness)

← "fitness"
 "energy" solns

② only invariance of μ in mild sense

Idea of the proof of Thm 3.7:

$$\text{Let } X_j = i \partial_t \tilde{u}^{N_j} + \Delta \tilde{u}^{N_j} - F_{N_j}(\tilde{u}^{N_j})$$

↑ $\mathcal{D}'_{t,x}$ -valued r.v.

$$\Rightarrow \mathcal{L}(X_j) = \delta_0$$

$$\Rightarrow \mathcal{L}(X_\infty) = \delta_0, \text{ i.e.}$$

↑

$u = \lim_{j \rightarrow \infty} \tilde{u}^{N_j}$ is a distributional soln to (WNLS).

$$\bullet \quad i \partial_t \tilde{u}^{N_j} + \Delta \tilde{u}^{N_j} \longrightarrow i \partial_t u + \Delta u, \text{ easy}$$

$$\bullet \quad \langle |u|^{2(m-1)} u \rangle = \lim_{j \rightarrow \infty} F_{N_j}(\tilde{u}^{N_j}) \text{ exists in } L^q(\rho_t; H^{-\varepsilon}(\mathbb{T}^2)), \varepsilon > 0.$$

← analogous to Prop 3.2

First, study $\langle F_{N_j}(\tilde{u}^{N_j}), e_n \rangle_{L^2_x}$

• (defocusing) WNLW on \mathbb{T}^2

(2)

Oh-Thomann '15

Thm 3.8: WNLW is almost surely locally well-posed w.r.t. μ .

\Rightarrow a.s. GWP & invariance of μ .

Key ingredient

① $Z_N = P_{\leq N} S(t) (u_0^\omega, u_1^\omega)$

$\Rightarrow P \left(\| \langle \nabla \rangle^{-\varepsilon} : Z_N^k : \|_{L_T^q L_x^r} > \lambda \right) \leq C \exp \left(-c \frac{\lambda^{2/k}}{T^{2/qk}} \right)$

② $: u^{2m-1} : = \lim_{N \rightarrow \infty} : u_N^{2m-1} :$

$u_N = Z_N + V_N$

$= \lim_{N \rightarrow \infty} \sum_{k=0}^{2m-1} \binom{2m-1}{k} : Z_N^k : V_N^{2m-1-k}$

\uparrow
 H^s for some $s > 0$

Sec 3.6: Further topics

(3)

① μ , invariant. $L(u(t)) = \mu = L(u(0))$

How are they related?

Difficult question: study space-time covariance

$$\mathbb{E}[u(t, x) \overline{u(s, y)}]$$

- Weakly nonlinear, and large box limit.
($\varepsilon \rightarrow 0$) ($|A| \rightarrow \infty$)
- Dynamical property?
 - recurrence, ✓
 - ergodicity? (difficult)
 - Known for some SPDEs
 - \Leftrightarrow uniqueness of an invariant meas

↑

NOT true for (deterministic) nonlin Hamil PDEs.

e.g. δ_0 .

$$d\delta_0 = Z^{-1} e^{-\frac{1}{2} \|u\|_{L^2}^2} du \quad \leftarrow \text{white noise inv for KdV.}$$

Quastel-Valko, Oh QdV.

② \mathbb{T}^3 .

④

• Gibbs meas on \mathbb{T}^3 is renormalizable

only for cubic & defocusing

i.e. $\begin{matrix} \textcircled{4} \\ \textcircled{3} \end{matrix}$ — deg of nonlin in Hamil
dim

• Wick ordering is NOT enough.

$$\sigma_N = \mathbb{E}_{P_t} [u^2(x)] \sim N \text{ when } d=3.$$

but need to replace it by $C_1 N + C_2 \log N$.

• Stochastic quantization eqn

space-time white noise

$$\partial_t u = \Delta u - u - u^3 + \infty \cdot u + \xi$$

"LWP": Hairer '14 (regularity structure)

Catellier - Chouk '14 (paracontrolled distribution)

à la Gubinelli - Imkeller - Perkowski

Kupiainen '14 (renormalization group method)

Invariance: Hairer - Matetski '15

GWP: Mourrat - Weber '16.

Q: NLS / NLW on \mathbb{T}^3 ?

Chap 4: Introduction to stochastic dispersive PDEs

(5)

Stochastic NLS (SNLS):

$$i \partial_t u + \Delta u = \pm |u|^{p-1} u + \phi \Xi$$

bdd op on L^2 (smoothing)
(actually,
Hilbert-Schmidt: $L^2 \rightarrow H^s$)

Stochastic NLW (SNLW):

$$-\partial_t^2 u + \Delta u = u^p + \Xi$$

• $\Xi =$ space time white noise. = " $\frac{\partial^2 B}{\partial x \partial t}$ "

• Ito's formulation:

$$i du = (-\Delta u \pm |u|^{p-1} u) dt + \phi dW$$

$W(t) = L^2$ -cylindrical Wiener process

$$= \sum_{n \in \mathbb{Z}^d} \beta_n(t) e^{in \cdot x}, \quad \{\beta_n\} = \text{indep } \mathbb{C}\text{-valued Brownian motions.}$$

$$\beta_n = \beta_n^{(r)} + i \beta_n^{(i)}$$

↑ ↑
indep indep
ℝ-valued BM

Note: $X_t(\omega) = X(t, \omega)$

(6)

Sec 4.1 Basic stochastic calculus

(Ω, \mathcal{F}, P) , prob space.

A BM. on \mathbb{R}_+ ($t \geq 0$) is a stochastic process, satisfying

(i) $B(0) = 0$, a.s.

(ii) $B(t) - B(s) \sim N(0, t-s)$, $t > s$

↑ normal distri

↓ var

(iii) $B(t_1) - B(s_1), B(t_2) - B(s_2)$, $t_1 \geq s_1 \geq t_2 \geq s_2$

are indep (i.e. indep increment over disjoint intervals.)

Properties: ① BM is a.s. conti. (i.e. \exists "conti version".)

$$\mathbb{E} \left(|B(t) - B(s)|^{2k} \right) = \frac{(2k)!}{2^k k!} |t-s|^k$$

$= (2k-1)!!$

i.e. $\mathbb{E} \left(|B(t) - B(s)|^p \right) \sim |t-s|^{p/2}$

Kolmogorov continuity criterion (Bass, EX 8.2)

⑦

$\{X_t\}$ with values in a metric space S . (separable?)

$p, \varepsilon > 0$ Suppose
$$\mathbb{E} \left(d(X_s, X_t)^p \right) \leq C_0 |t-s|^{1+\varepsilon}, \quad \forall t, s.$$

Then,
$$P \left(\sup_{s \neq t} \frac{d(X_s, X_t)}{|t-s|^{\frac{\varepsilon}{p} - \gamma}} \geq \lambda \right) \leq \frac{C_1}{\lambda^p}.$$

$\Rightarrow X_t$ is unif conti a.s.

Back to BM. $\frac{\varepsilon}{p} = \frac{p-1}{p} = \frac{1}{2} - \frac{1}{p} \rightarrow \frac{1}{2}$ as $p \rightarrow \infty$.

$B \in \dot{W}_{loc}^{b,p}$, $b < 1/2$, $p \leq \infty$

$p < \infty$:
$$\mathbb{E} \left[\|B(t)\|_{\dot{W}^{b,p}(I)}^p \right] = \int_I \int_I \frac{\mathbb{E}(|B(t) - B(t')|^p)}{|t-t'|^{1+bp}} dt' dt$$

$$\sim \iint_{I \times I} |t-t'|^{-1-bp+p/2} dt' dt < \infty$$

iff $b < 1/2$

$$B \in B_{p, \infty}^{1/2}, \quad p < \infty \quad (\text{but } B \in B_{\infty, \infty}^{1/2-})$$

(8)

Roynette '93 SS rep

Bényi - Oh '11 Ad. Math.

(2) Covariance

$$\mathbb{E}[B(t)B(s)] = t \wedge s = \min(t, s)$$

$$\begin{aligned} & \mathbb{E}[B(s)(B(t) - B(s))] + \mathbb{E}[B^2(s)] = s, \quad t > s \\ & \quad \quad \quad \uparrow \quad \quad \uparrow \\ & \quad \quad \quad \text{indep} \quad \quad = 0 \end{aligned}$$

White noise: $\xi = dB$

$$\xi \rightsquigarrow \mathbb{E}[\xi(t)\xi(s)] = \delta(t-s), \quad \text{"}\xi \text{ scales like } \sqrt{\delta}\text{"}$$

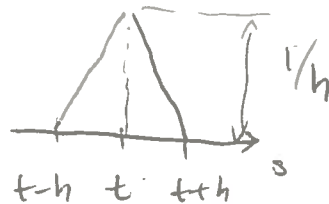
$h > 0$

$$\phi_h(s) = \mathbb{E}\left[\frac{B(t+h) - B(t)}{h} \cdot \frac{B(s+h) - B(s)}{h}\right]$$

$$= \dots = \frac{1}{h^2} \left((t+h) \wedge (s+h) - (t+h) \wedge s - t \wedge (s+h) + t \wedge s \right)$$

$$\cdot \int \phi_h(s) ds = 1$$

$$\phi_h(s) \rightarrow \delta(t-s) \quad \text{as } h \rightarrow 0$$



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①

Back to L^2 -cylindrical Wiener process.

$$W(t) = \sum_{n \in \mathbb{Z}^d} \beta_n(t) e^{in \cdot x} \in H^{-\frac{d}{2}-\varepsilon}(\mathbb{T}^d) \\ W^{-\frac{d}{2}-\varepsilon, \infty}(\mathbb{T}^d)$$

$$W \in W_{t, \text{loc}}^{\frac{1}{2}-, \infty} W_x^{-\frac{d}{2}-\varepsilon, \infty}(\mathbb{T}^d)$$

(SNLS) \Leftrightarrow study it in the mild formulation:

$$u(t) = S(t) u_0 - i \int_0^t S(t-t') |u|^{p-1} u \, dt' \\ - i \int_0^t S(t-t') \phi \, dW(t')$$

= stochastic convolution = Ψ

$$\Psi = \int_0^t S(t-t') \phi \, dW(t') = \sum_{n \in \mathbb{Z}^d} \int_0^t e^{-im^2(t-t')} \phi(e^{in \cdot x}) \, d\beta_n(t')$$

For simplicity, we only consider $\phi = \text{diag}(\phi_n)$ i.e. $\phi(e_n) = \phi_n e_n$
translation invariant.

• Wiener integral (real-valued case)

(2)

$$\int_a^b f(t) dB(t) \sim \langle f, dB \rangle_{L_t^2(a,b)}$$

↑
deterministic

Step 1: step func. $f(t) = \sum_{j=1}^m a_{j-1} \mathbb{1}_{[t_{j-1}, t_j)}(t)$

\Rightarrow Define $I(f) = \sum_{j=1}^n a_{j-1} (B(t_j) - B(t_{j-1}))$
(left endpt Riemann sum)

Then, ① $\mathbb{E}[I(f)] = 0$

② $\mathbb{E}[(I(f))^2] \stackrel{\uparrow \text{indep}}{=} \sum_{j=1}^n a_{j-1}^2 (t_j - t_{j-1})$
 $= \int_a^b f^2(t) dt$
 $= \|f\|_{L^2(a,b)}^2$

step 2: Given $f \in L^2(a, b)$,

(3)

approximate f by step functions f_n in $L^2(a, b)$

and define $I(f) = \lim_{n \rightarrow \infty} I(f_n)$ ← Gaussian

⇒ ① & ② hold.

Namely, $I: L^2(a, b) \rightarrow L^2(\Omega)$ is an isometry.
 $\rightarrow H'_1$ — Wiener homog. chaoses of order 1.

(just like the white noise functional in Chap 3.)

Rmk: If $f \in C^1$, we can define it as

Paley-Wiener-Zygmund integral:

$$I(f) = \int_a^b f dB = - \int_a^b f'(t) B(t) dt \quad (\text{if } f(b) = 0)$$

Rmk: If B is \mathbb{C} -valued,

$$\mathbb{E}[(I(f))^2] = 2 \|f\|_{L^2([a, b])}^2$$

• Ito integral:

✓ σ -field

④

filtration $\{\mathcal{F}_t\}_{t \geq 0}$ s.t. $\mathcal{F}_t \subset \mathcal{F}_{t_2} \subset \mathcal{F}$, $t_1 \leq t_2$

• We say $X(t)$ is adapted (non-anticipating)

if $X(t)$ is \mathcal{F}_t -measurable, $\forall t \geq 0$.

• progressively meas: $\forall T$,

$[0, T] \times \Omega$

↓

$(t, \omega) \longmapsto X(t, \omega)$ is $\mathcal{B}_{[0, T]} \otimes \mathcal{F}_t$ -meas.

• p.m. \rightarrow adapted

• adapted & left (or right) conti \rightarrow p.m.

e.g. adapted & càdlàg func \rightarrow p.m.

↓
continue à droite
limite à gauche

"Assume" (i) $B(t)$ is \mathcal{F}_t -meas

(5)

(ii) $B(t) - B(s)$ is indep of $\{\mathcal{F}_s\}_{s < t}$.

e.g. simply take $\mathcal{F}_t^B = \sigma(\{B(s) : s \leq t\})$

We set

$$L_{ad}^2([a, b] \times \Omega) = \{ f(t, \omega) :$$

↑

① f is adapted to $\{\mathcal{F}_t\}$

② $\int_a^b \mathbb{E}(f^2(t)) dt < \infty$

Define the Ito integral here.

Rmk: can define Ito integral on a larger class of stochastic processes.

e.g. $L_{ad}(\Omega ; L^2([a, b]))$

① f , adapted

② $\int_a^b |f(t, \omega)|^2 dt < \infty$, a.s.

Step 1: Step stoch process

$$f(t, \omega) = \sum_{j=1}^n a_{j-1}(\omega) \mathbb{1}_{[t_{j-1}, t_j)}(t)$$

"does not peek in the future"

⑥

a_j, \mathcal{F}_{t_j} -meas

$$\sum \mathbb{E}(a_j^2) < \infty$$

• Define Ito integral:

$$I(f)(\omega) = \sum_{j=1}^n a_{j-1}(\omega) (B(t_j) - B(t_{j-1}))$$

(left endpt Riemann sum)

Lemma 4.1: ① $\mathbb{E}[I(f)] = 0$

$$\textcircled{2} \mathbb{E}[(I(f))^2] = \int_a^b \mathbb{E}[f^2] dt \quad (\text{Ito isometry})$$

Recall: Conditional expectation.

(7)

If $X \in L^1(\Omega; \mathcal{F})$ and $\mathcal{G} \subset \mathcal{F}$,
 \uparrow sub σ -field,

then define the conditional expectation of X given \mathcal{G}
by the unique r.v. Y s.t.

① Y is \mathcal{G} -meas

② $\int_A X dP = \int_A Y dP, \quad \forall A \in \mathcal{G}.$

Rmk: ① Y is given by Radon-Nikodym Thm:

$$\mu(A) = \int_A X dP, \quad A \in \mathcal{G}.$$

$$\Rightarrow \mu \ll P|_{\mathcal{G}}$$

R-N
 \Rightarrow

$$d\mu = Y dP|_{\mathcal{G}}$$

\uparrow
 L^1

Denote it by $Y = \mathbb{E}[X|\mathcal{G}]$

$$\textcircled{2} \quad X \in L^2(\Omega; \mathcal{F})$$

Ⓟ

$$\Rightarrow \mathbb{E}[X | \mathcal{G}] = P_{L^2(\Omega; \mathcal{G}) \leftarrow \text{closed}}(X)$$

Properties: (i) $\mathbb{E}[\mathbb{E}[X | \mathcal{G}]] = \mathbb{E}(X)$

In general, more useful to read it as
$$\mathbb{E}[X] = \mathbb{E}[\mathbb{E}[X | \mathcal{G}]]$$

i.e. compute expectation by conditioning

(ii) If X is \mathcal{G} -meas,

$$\mathbb{E}[X | \mathcal{G}] = X$$

(iii) If X & \mathcal{G} are indep, $\left(\begin{array}{l} \{X \in U\} \text{ and } A \in \mathcal{G} \text{ are indep} \\ \forall U, V \in \mathcal{B}_{\mathbb{R}}, A \in \mathcal{G} \end{array} \right.$

$$\mathbb{E}[X | \mathcal{G}] = \mathbb{E}[X]$$

(iv) If Y is \mathcal{G} -meas and $\mathbb{E}[XY] < \infty$,

then
$$\mathbb{E}[XY | \mathcal{G}] = Y \mathbb{E}[X | \mathcal{G}]$$

pf of Lemma 4.1:

(9)

$$\textcircled{1} \quad \mathbb{E} [a_{j-1} (B(t_j) - B(t_{j-1}))]$$

$$\stackrel{\text{(ii)}}{=} \mathbb{E} [\mathbb{E} [a_{j-1} (B(t_j) - B(t_{j-1})) \mid \mathcal{F}_{t_{j-1}}]]$$

$$\stackrel{\text{(iv)}}{=} \mathbb{E} [a_{j-1} \underbrace{\mathbb{E} [B(t_j) - B(t_{j-1}) \mid \mathcal{F}_{t_{j-1}}]]}_{\stackrel{\text{(iii)}}{=} \mathbb{E} [B(t_j) - B(t_{j-1})] = 0}] = 0$$

(2) $i < j$.

$$\mathbb{E} [a_{i-1} a_{j-1} (B(t_i) - B(t_{i-1})) (B(t_j) - B(t_{j-1}))]$$

$$\stackrel{\text{(i),(iv)}}{=} \mathbb{E} [a_{i-1} a_{j-1} (B(t_i) - B(t_{i-1})) \underbrace{\mathbb{E} [B(t_j) - B(t_{j-1}) \mid \mathcal{F}_{t_{j-1}}]]}_{\stackrel{\text{(iii)}}{=} 0}]$$

= 0

(iii) 0

$$\begin{aligned} & \underline{i=j} \\ & \mathbb{E} \left[a_{j-1}^2 (B(t_j) - B(t_{j-1}))^2 \right] \\ & \stackrel{(i), (iv), (ii)}{=} \mathbb{E} \left[a_{j-1}^2 \right] (t_j - t_{j-1}) \end{aligned}$$

□

Step 2: FACT: Given $f \in L_{ad}^2([a, b] \times \Omega)$,
 $\exists \{f_n\}$ of step stoch processes
 converging to f .

Define Ito integral:

$$I(f) = \int_a^b f(t, \omega) dB(t, \omega)$$

$$\stackrel{\text{def}}{=} \lim_{n \rightarrow \infty} I(f_n).$$

Properties : ① I , linear

②

$$\textcircled{2} \quad \mathbb{E}(I(f)) = 0$$

$$\textcircled{3} \quad \mathbb{E}[(I(f))^2] = \int_a^b \mathbb{E}[f^2] dt \quad (\text{Ito isometry})$$

$$\begin{aligned} \textcircled{4} \quad \mathbb{E} \left[\int_a^b f(t) dB \int_a^b h(t) dB \right] \\ = \int_a^b \mathbb{E} [f(t) h(t)] dt. \end{aligned}$$

$I: L^2_{ad}([a, b] \times \Omega) \longrightarrow L^2(\Omega)$ is an isometry

Lec 20 15 / 03 / 17 (Wed)

①

Sec 4.2: Local well-posedness of SNLS

$$\text{(SNLS)} \begin{cases} i du = (-\Delta u \pm |u|^2 u) dt + \phi dW & \text{on } \mathbb{T} \\ u|_{t=0} = u_0 \end{cases}$$

$$\phi = \text{diag } \phi_n, \text{ HS on } L^2(\mathbb{T})$$

$$\phi(e_n) = \phi_n e_n$$

Mild formulation: $S(t) = e^{it\Delta}$

$$u(t) = S(t) u_0 \mp i \int_0^t S(t-t') (|u|^2 u)(t') dt'$$

$$- i \int_0^t S(t-t') \phi dW(t')$$

Stochastic convolution Ψ

• Review of deterministic LWP of cubic NLS on \mathbb{T}

(2)

(Bourgain '93)

• Fourier restriction norm method

$$\|u\|_{X^{s,b}} = \|\langle m \rangle^s \langle \tau + m^2 \rangle^b \widehat{u}(m, \tau)\|_{L_m^2 L_\tau^2}$$

$$\left(\begin{array}{l} \text{lin Schrödinger} = i\partial_t u + \Delta u = 0 \\ \Rightarrow -(\tau + m^2) \widehat{u}(m, \tau) = 0 \\ \text{F.T in space-time} \end{array} \right.$$

• local-in-time version: $T > 0$.

$$\|u\|_{X_T^{s,b}} = \inf \left\{ \|v\|_{X^{s,b}} : v|_{[0,T]} = u \right\}.$$

• Duhamel formula:

$$u(t) = \Gamma_{u_0}(u)(t) = \eta(t) S(t) u_0 - i \eta\left(\frac{t}{T}\right) \int_0^t S(t-t') |u|^2 u(t') dt'.$$

$$\eta(t) \equiv \begin{cases} 1 & \text{on } [0, 1] \\ 0 & \text{on } [-\frac{1}{2}, \frac{3}{2}]^c \end{cases}$$

\Rightarrow If $u = \Gamma_{u_0}(u)$, then u is a soln on $[0, T]$, $0 < T \leq 1$.

- Deterministic estimates

(3)

① homog lin $\| \eta(t) S(t) u_0 \|_{X^{s,b}} \leq C_b \| u_0 \|_{H^s}$

$$\mathcal{F}_{x,t} (\eta(t) S(t) u_0) = \widehat{\eta}(\tau + n^2) \widehat{u}_0(m)$$

② nonhomog lin:

$$\left\| \eta\left(\frac{t}{T}\right) \int_0^t S(t-t') F(t') dt' \right\|_{X^{s,b}} \lesssim T^\theta \| F \|_{X^{s,b-1+\theta}}$$

• $b > 1/2$, $\theta > 0$ small.

③ L^4 -Strichartz estimate

$$\| u \|_{L^4_{x,t}(\mathbb{T} \times \mathbb{R})} \lesssim \| u \|_{X^{0,3/8}}$$

④ $\| u \|_{C_T H^s} \lesssim \| u \|_{X^{s,b}_T}$, $b > 1/2$.

(\Leftarrow 1-d Sobolev in time.)

④

$$\Rightarrow \|\Gamma_{u_0} u\|_{X_T^{0, \frac{1}{2}+}}$$

$$\leq \| \eta(t) S(t) u_0 \|_{X^{0, \frac{1}{2}+}} + \left\| \eta\left(\frac{t}{T}\right) \int_0^t S(t-t') |v|^2 v(t') dt' \right\|_{X^{0, \frac{1}{2}+}}$$

(for any extension v of u)

$$\stackrel{\textcircled{1}, \textcircled{2}}{\lesssim} \|u_0\|_{L^2} + T^\theta \| |v|^2 v \|_{X^{0, -\frac{1}{2} + \theta +}}$$

$$(X^{s,b})^* = X^{-s,b}$$

$$\sup_{\|w\|=1} \underbrace{\left| \iint w \cdot |v|^2 v \, dx \, dt \right|}_{X^{0, \frac{1}{2} - \theta -}}$$

$$\leq \|w\|_{L_{x,t}^4} \|v\|_{L_{x,t}^4}^3$$

$$\stackrel{\textcircled{3}}{\lesssim} \|w\|_{X^{0, \frac{3}{8}}} \|v\|_{X^{0, \frac{3}{8}}}^3$$

$$\leq \|v\|_{X^{0, \frac{3}{8}}}^3$$

Taking inf over v .

$$\Rightarrow \|\Gamma_{u_0} u\|_{X_T^{0, \frac{1}{2}+}} \lesssim \|u_0\|_{L^2} + T^\theta \|u\|_{X_T^{0, \frac{1}{2}+}}$$

• A difference estimate holds in a similar manner

\Rightarrow contraction in $B_R \subset X_T^{0, \frac{1}{2}+}$ $R = C \|u_0\|_{L^2}$
 $0 < T = T(\|u_0\|_{L^2}) \ll 1$

Back to SNLS.

• method 1: Show

$$\mathbb{E} \left[\|\Psi^w\|_{X_T^{0, \frac{1}{2}-}}^2 \right] \lesssim 1 < \infty.$$

\Rightarrow fixed pt argument in $X_T^{0, \frac{1}{2}-}$

$$R^w \sim \|u_0\|_{L^2} + \|\Psi^w\|_{X_T^{0, \frac{1}{2}-}}$$

$$\Rightarrow T = T^w > 0.$$

• Then, show the continuity in time.

$$u(t) = \underbrace{\text{lin}}_{\text{conti}} + \underbrace{\text{nonlin}}_{\substack{\text{estimate} \\ \text{in } X_T^{0, \frac{1}{2}+}}} + \Psi$$

\nwarrow Kolmogorov conti criterion.

• method 2: write $u = \Psi + v$.

(6)

Need to estimate $\|\Psi\|_{L_T^q L_x^r}$

$p \geq q, v, r$

$$\left[\mathbb{E} \left(\|\Psi\|_{L_T^q L_x^r}^p \right) \right]^{1/p}$$

Minkowski
 \leq

Wiener chaos
est
 \lesssim

$\hookrightarrow \mathcal{H}^1 =$ homog Wiener chaoses of order 1

$$\|\Psi(x,t)\|_{L^p(\Omega)} \Big\|_{L_T^q L_x^r}$$

$$p^{1/2} \|\Psi(x,t)\|_{L^2(\Omega)} \Big\|_{L_T^q L_x^r}$$

$$\stackrel{\text{indep}}{=} \left(\sum_n \mathbb{E} \left(\left| \int_0^t e^{-im^2(t-t')} \phi_n e^{inx} d\beta_n(t') \right|^2 \right) \right)^{1/2}$$

$$\stackrel{\text{H isometry}}{=} 2 |\phi_n|^2 t$$

$$\lesssim p^{1/2} T^{\frac{1}{2} + \frac{1}{q}} \left(\sum |\phi_n|^2 \right)^{1/2} \Rightarrow \text{chebyshev.}$$

$$= \|\phi\|_{\text{HS}(L^2; L^2)}$$

$$\left(\|\phi\|_{\text{HS}(X; Y)} = \left(\sum \|\phi(e_n)\|_Y^2 \right)^{1/2}, \quad \{e_n\}, \text{ ONB in } X \right)$$

Fixed pt problem for v :

(7)

$$v(t) = \gamma(t) S(t) u_0 + i \gamma\left(\frac{t}{T}\right) \int_0^t S(t-t') |v + \Psi|^2 (v + \Psi) dt'$$

\Rightarrow contraction in $X_T^{0, \frac{1}{2}+}$ (a bit smoother)

only change:

$$\|v + \Psi\|_{L^4_{x,t}} \leq \|v\|_{L^4_{x,t}} + \underbrace{\|\Psi^\omega\|_{L^4_{x,t}}}_{< \infty, \text{ a.s.}}$$

Thm 4.2: Stochastic cubic NLS on \mathbb{T} (with $\phi \in \text{HS}(L^2, L^2)$)

is pathwise locally well-posed in $L^2(\mathbb{T})$

Sec 4.3: Global well-posedness in $L^2(\Pi)$

8

Idea: Control $M(u) = \sum_{n \in \mathbb{Z}} |\hat{u}_n|^2 = \sum p_n^2 + q_n^2$

$$p_n = \operatorname{Re} \hat{u}_n$$

$$q_n = \operatorname{Im} \hat{u}_n$$

$$d\hat{u}_n = (-im^2 \hat{u}_n - i \mathcal{F}_x(|u|^2 u)(m)) dt - i \Phi_n d\beta_n$$

$$\Rightarrow dp_n = (m^2 q_n + \operatorname{Im} \mathcal{F}_x(|u|^2 u)(m)) dt + \operatorname{Im}(\Phi_n d\beta_n)$$

$$dq_n = (-m^2 p_n - \operatorname{Re} \mathcal{F}_x(|u|^2 u)(m)) dt - \operatorname{Re}(\Phi_n d\beta_n)$$

$$\operatorname{Im} \Phi_n d(\operatorname{Re} \beta_n) + \operatorname{Re} \Phi_n d(\operatorname{Im} \beta_n)$$

$$- \operatorname{Re} \Phi_n d(\operatorname{Re} \beta_n) + \operatorname{Im} \Phi_n d(\operatorname{Im} \beta_n)$$

• Ito's Lemma: $dX = f dt + g dB$

(9)

Consider $F(X)$

$$\text{Then, } dF = \partial_x F dX + \frac{1}{2} \partial_x^2 F (dX)^2$$

$$= \partial_x F (f dt + g dB) + \frac{1}{2} \partial_x^2 F g^2 dt$$

$$(dt^2) = 0$$

$$dt dB = dB dt = 0$$

$$(dB)^2 = dt$$

$$\Rightarrow dM \stackrel{\text{Ito}}{=} \underbrace{2 \sum_n (p_n dp_n + q_n dq_n)} + \underbrace{\sum ((dp_n)^2 + (dq_n)^2)}$$

$$= 2 \sum_n p_n \text{Im}(\Phi_n d\beta_n)$$

$$- q_n \text{Re}(\Phi_n d\beta_n)$$

$$= 2 \sum |\Phi_n|^2 dt$$

$$= 2 \|\Phi\|_{\text{HS}(L^2; L^2)}^2 dt$$

Integrate from 0 to t (large), sup over $t \in [0, T]$, & take expectation

\Rightarrow Apply Burkholder-Davis-Gundy ineq.

$$\mathbb{E} \left[\sup_{t \in [0, T]} \|u(t)\|_{L^2}^2 \right] \leq \|u_0\|_{L^2}^2 + C(T, \|\Phi\|_{\text{HS}(L^2; L^2)})$$

\Rightarrow a.s. GWP. in $L^2(\Pi)$ by iterating the local argument.

B-D-G inequality: X , ^{local} martingale

(10)

$$\mathbb{E} \left(\sup_{t \in [0, T]} |X(t)|^p \right) \sim \mathbb{E} \left[\langle X(t) \rangle_{[0, T]}^{p/2} \right], \quad 1 \leq p < \infty$$

↑
quadratic variation

Rmk: In order to justify the application of Ito's lemma, one should first consider the finite dim'l dynamics:

$$\begin{cases} i dU^N = (-\Delta U^N \pm P_{\leq N}(|U^N|^2 U^N)) dt + P_{\leq N}(\phi dW) \\ U^N|_{t=0} = U_0^N := P_{\leq N} U_0 \end{cases}$$

and obtain an a priori bound (indep of N)

Then, proceed with an approximation argument to obtain the same a priori bound for solutions to SNLS.

• Another alternative is to apply the infinite dim'l version of Ito's lemma (such as the one in Da Prato - Zabczyk.)