

Lec 8 02 / 04 / 21 (Fri)

①

$$U(x) = \sum_{n \in \mathbb{Z}^2} \frac{g_n(\omega)}{\langle n \rangle} e^{in \cdot x}$$

law of u

$$\downarrow$$
$$L(u) = \mu_1 = \underbrace{\text{GFF on } \mathbb{T}^2}_{\text{massive}}$$

$\{g_n\}_{n \in \mathbb{Z}^2}$ indep std \mathbb{C} -valued Gaussian r.v.'s

(In the real-valued setting,
impose $g_{-n} = \overline{g_n}$)

• Last time:

$$u \in \underline{H^s}(\mathbb{T}^2) \setminus L^2(\mathbb{T}^2), \text{ a.s. } \boxed{s < 0}$$

What about $L^p(\mathbb{T}^2)$, $p < 2$?

$$L^p(\mathbb{T}^2) \supset L^2(\mathbb{T}^2)$$

• $p=1$:

$$\underline{u \notin L^1(\mathbb{T}^2), \text{ a.s.}}$$

$$\text{Let } X_n = \frac{g_n}{\langle n \rangle} e^{in \cdot x} \quad (\text{for fixed } x \in \mathbb{T}^2)$$

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$$\text{and } Y_n = X_n \cdot \mathbb{1}_{\{|X_n| \leq A\}} \quad (\text{for some } A > 0)$$

Kolmogorov three series thm: X_n indep

Then, $\sum X_n$ converges a.s. iff the following 3 cond. hold:
for ANY $A > 0$

(i) $\sum P(|X_n| \geq A)$ converges

(ii) With $Y_n = X_n \cdot \mathbb{1}_{\{|X_n| \leq A\}}$,

$$\sum \mathbb{E}[Y_n] \text{ converges.}$$

(iii) $\sum \text{var}(Y_n)$ converges.

$$|g_n| \leq A \langle n \rangle$$



For our problem, $\sum \text{var}(Y_n) \geq \sum_{n \in \mathbb{Z}^2} \frac{1}{\langle n \rangle^2} \mathbb{E}[|g_n|^2 \cdot \mathbb{1}_{\{|g_n| \leq A\}}]$
 $= \infty$

\Rightarrow By Kolmog. 3-series thm, $\sum X_n$ diverges on a set of positive prob.

\Rightarrow By Kolmog 0-1 law, $\sum X_m(x)$ diverges a.s.

Compute

$$\mathbb{E} \left[\int_{\mathbb{T}^2} |u(x)| dx \right] = \int_{\mathbb{T}^2} \overbrace{\mathbb{E}[|u(x)|]}^{", \infty"} dx = \infty$$

$\Rightarrow u_N$ does not converge in L^1 with positive prob.

$\Rightarrow u_N$ diverges in $L^1(\mathbb{T}^2)$, a.s.

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Back to $u(x) = \sum_{n \in \mathbb{Z}^2} \frac{g_n}{\langle n \rangle} e^{in \cdot x}$

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$$\mathbb{E} [u_N(x) \overline{u_N(y)}] = \sum_{|n| \leq N} \frac{1}{\langle n \rangle^2} e_n(x-y)$$

insert $\langle \nabla_x \rangle^{-s}$ $\langle \nabla_y \rangle^{-s}$

and set $x=y$.

$$\Rightarrow \mathbb{E} [\langle \nabla \rangle^{-2s} |u_N(x)|^2]$$

$$\lesssim \sum_{\substack{|n| \leq N \\ m \in \mathbb{Z}^2}} \frac{1}{\langle m \rangle^{2+2s}} < \infty \quad \text{unif in } N \in \mathbb{N}.$$

$$\Rightarrow \left\| \| u_N \|_{W^{s,r}(\mathbb{T}^2)} \right\|_{L^p(\Omega)} \stackrel{\text{Mink}}{\leq} \left\| \left\| \langle \nabla \rangle^s u_N(x) \right\|_{L^p(\Omega)} \right\|_{L_x^r(\mathbb{T}^2)}$$

$$\lesssim p^{1/2}, \text{ unif in } N \in \mathbb{N}$$

$$= p^{1/2} \| \langle \nabla \rangle^s u_N(x) \|_{L^2(\Omega)}$$

for $s < 0$

Note:

$$\mathbb{E} [g_n^k \overline{g_m^l}] = k! \delta_{kl}$$

In particular,

$$\mathbb{E} [g_n^2] = 0.$$

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A similar computation for $u_N - u_M$ holds

$\Rightarrow \{u_N\}_{N \in \mathbb{N}}$ converges in $L^p(\Omega; W_x^{s,r}(\mathbb{T}^2))$

$s < 0$ and $r \leq \infty$

Also, we have an exponential tail estimate:

$$P(\|u\|_{W_x^{s,r}} > \lambda) \leq C e^{-c\lambda^2}, \quad \forall \lambda > 0$$

Since u is NOT a function and is only a distribution, the power u^k is ill-defined.

\Rightarrow We need to introduce a renormalization.

In the following, we restrict our attention to the real-valued setting.

Hermite polynomial:

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generating function: $e^{tx - \frac{1}{2}\sigma t^2} = \sum_{k=0}^{\infty} \frac{t^k}{k!} \underline{H_k(x; \sigma)}$

• $H_k(x; \sigma) =$ Hermite poly of deg k .

$$H_0(x; \sigma) = 1, \quad H_1(x; \sigma) = x, \quad H_2(x; \sigma) = x^2 - \sigma$$

$$H_3(x; \sigma) = x^3 - 3\sigma x, \quad H_4(x; \sigma) = x^4 - 6\sigma x^2 + 3\sigma^2, \dots$$

• Set $H_k(x) = H_k(x; 1)$.

orthogonality: $\int H_k(x) H_m(x) \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx = k! \delta_{km}$.

FACT: $\left\{ \frac{H_k(x)}{\sqrt{k!}} \right\}_{k \in \mathbb{Z}_{\geq 0}}$ forms an O.N.B. of $L^2(\mathbb{R}, \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx)$.

Note: $\{x^k\}_{k \in \mathbb{Z}_{\geq 0}} \xrightarrow{\text{Gram-Schmidt in } L^2(\mathbb{R}; \text{Gauss})} \{H_k(x)\}_{k \in \mathbb{Z}_{\geq 0}}$

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We also have the following orthogonality:

f, g , Gaussian r.v.'s with var σ_f and σ_g
(mean 0)

$$\mathbb{E} [H_k(f; \sigma_f) H_m(g; \sigma_g)] = \delta_{km} k! \left\{ \mathbb{E}[fg] \right\}^k$$

⌊ \Leftarrow Take the product of the generating functions
for (f, σ_f) and (g, σ_g) .
and express in two different series
and compare the coeffs of t^k .

- In the complex-valued setting, we instead use
the (generalized) Laguerre polynomials: See Oh-Thomann
SPDE '18.

f, g .
jointly
Gaussian

(*)

Wick renormalization

Wick power, renormalized power

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$$: U_N^k(x) : = H_k(U_N(x); \sigma_N) \quad \text{pointwise}$$

$$\text{where } \sigma_N = \mathbb{E}[U_N^2(x)] = \sum_{|m| \leq N} \frac{1}{\langle m \rangle^2} \sim \log N \rightarrow \infty.$$

(On a mfd, σ_N depends on $x \in \mathcal{U}$.

- $k=2$: $: U_N^2 : = U_N^2 - \sigma_N$
- $k=3$: $: U_N^3 : = U_N^3 - 3 \sigma_N U_N$
- Wick renormalization = orthogonal projection onto the homog Wiener chaos of deg k

• (H, B, μ) , abstract Wiener space

$$d\mu = Z^{-1} e^{-\frac{1}{2}\|u\|_H^2} du$$

(Think of $H = H^1(\mathbb{T}^2)$
 $B = H^{-\epsilon}(\mathbb{T}^2)$.)

• $\{e_j\}_{j \in \mathbb{N}} \subset B^*$ complete orthonormal system of $H^* = H$.

• Consider $\prod_{j=1}^{\infty} H_{k_j}(\langle x, e_j \rangle)$, $k_j \neq 0$ only for finitely many j 's

product of finitely many Hermite poly in different directions.

$\langle \cdot, \cdot \rangle = B - B^*$ dual pairing.

\mathcal{H}_k = homog Wiener chaoses of deg k
= span $\{ \prod_{j=1}^{\infty} H_{k_j}(\langle x, e_j \rangle) : \sum_j k_j = k \}$ $\|\cdot\|_{L^2(B, \mu)}$

Wiener-Ito decomposition:

$$L^2(B, \mu) = \bigoplus_{k=0}^{\infty} \mathcal{H}_k$$

We set $\mathcal{H}_{\leq k} = \bigoplus_{j=0}^k \mathcal{H}_j$

Lemma (Wiener chaos estimate)

Let $k \in \mathbb{N}$. Then, we have

$$\|X\|_{L^p(\Omega)} \leq (p-1)^{k/2} \|X\|_{L^2(\Omega)}$$

for any $p \geq 2$ and any $X \in \mathcal{H}_{\leq k}$

Nelson '65. \Leftarrow follows from the hypercontractivity of the OU semigroup.
 $e^{tL} \quad L = \Delta - x \cdot \nabla \quad (B = \mathbb{R}^d)$
elements in $\mathcal{H}_k =$ eigenfunction of L with e-val $-k$.

Prop: $\{ : u_N^k : \}_{N \in \mathbb{N}}$ forms a Cauchy seq

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in $L^p(\Omega; W^{s,r}(\mathbb{T}^2))$, $s < 0$, $r \leq \infty$.

Denoting the limit by $: u^k :$, we have

$: u^k : \in W^{s,r}(\mathbb{T}^2)$, a.s. $s < 0$, $r \leq \infty$

Pf: $\mathbb{E} [: u_N^k(x) : : u_N^k(y) :]$

$$= k! \left\{ \mathbb{E} [u_N(x) u_N(y)] \right\}^k$$

$$= k! \sum_{\substack{|m_j| \leq N \\ j=1, \dots, k}} \frac{1}{\langle m_1 \rangle^2} e_{n_1}(x-y) \cdots \frac{1}{\langle m_k \rangle^2} e_{n_k}(x-y)$$

$$= k! \sum_{|m_j| \leq N} \prod_{j=1}^k \frac{1}{\langle m_j \rangle^2} \underline{e_{n_1 + \dots + n_k}(x-y)}$$

Insert $\langle \nabla_x \rangle^\varepsilon$ and $\langle \nabla_y \rangle^\varepsilon$ and set $x=y$.

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$$\begin{aligned} \Rightarrow \mathbb{E} \left[\left| \langle \nabla \rangle^\varepsilon : u_N^k(x) : \right|^2 \right] \\ = k! \sum_{|n_j| \in \mathbb{N}} \prod_{j=1}^k \frac{1}{\langle m_j \rangle^2} \cdot \frac{1}{\langle n_1 + \dots + n_k \rangle^{2\varepsilon}} \\ \lesssim_{\varepsilon, k} 1, \quad \text{unif in } N \in \mathbb{N}. \end{aligned}$$

$$\begin{aligned} \Rightarrow_{r < \infty} \left\| \left\| : u_N^k : \right\|_{W_x^{s,r}} \right\|_{L^p(\Omega)} \stackrel{\text{Mink}}{\leq}_{p \geq r} \left\| \left\| \langle \nabla \rangle^s : u_N^k(x) : \right\|_{L^p(\Omega)} \right\|_{L_x^r} \\ \lesssim p^{k/2} \cdot \underline{\text{for } s < 0} \leq p^{k/2} \|\dots\|_{L^2(\Omega)} \\ \uparrow \\ \text{Wiener chaos estimate.} \end{aligned}$$

• A similar computation holds

for $: u_N^k : - : u_M^k :$, $N \geq M \geq 1$.

In this case, we use $\max(|m_j|) \geq M \Rightarrow \frac{1}{\langle m_j \rangle^2} \leq \frac{1}{M^2} \frac{1}{\langle m_j \rangle^{2-\delta}}$

□

Note:

$$P(\|:u^k:\|_{W^{s,r}} > \lambda) \leq c e^{-c \lambda^{2/k}} \text{ for any } \lambda > 0$$

(\Leftarrow Chebyshev)

$$\Downarrow \int e^{c \|:u^k:\|_{W^{s,r}}^{2/k}} d\mu_1 < \infty$$

Tzvetkov
Lem 4.5
in PTRF '10

Lemma: Let $\{X_N\}_{N \in \mathbb{N}}$ and X be
spatially homogeneous stoch processes: $\mathbb{R} \rightarrow \mathcal{D}'(\mathbb{T}^d)$

i.e. for any $x_0 \in \mathbb{T}^d$,

$$\{X(t, \cdot)\}_{t \in \mathbb{R}} \text{ and } \{X(t, \cdot + x_0)\}_{t \in \mathbb{R}}$$

have the same law.

• Suppose that $X_N(t), X(t) \in \mathcal{H}^s \leq \mathbb{R}, \forall t \in \mathbb{R}$

(i) Fix $t \in \mathbb{R}$. If $\exists s_0 \in \mathbb{R}$ s.t.

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(**)
$$\mathbb{E} [|\hat{X}(t, n)|^2] \lesssim \langle n \rangle^{-d-2s_0}, \quad \forall n \in \mathbb{Z}^d.$$

then, $X(t) \in W^{s, \infty}(\mathbb{T}^d)$, $\underline{s < s_0}$, a.s.

Moreover, if $\exists \theta > 0$ s.t.

$$\mathbb{E} [|\hat{X}_N(t, n) - \hat{X}(t, n)|^2] \lesssim N^{-\theta} \langle n \rangle^{-d-2s_0}, \quad \forall n \in \mathbb{Z}^d$$

then, $X_N \rightarrow X$ in $W^{s, \infty}(\mathbb{T}^d)$, $s < s_0$, a.s.

(ii) Fix $T > 0$. Suppose (i) holds on $[-T, T]$.

• If $\exists \sigma \in (0, 1)$ s.t.

$$\mathbb{E} [|d_h \hat{X}(t, n)|^2] \lesssim \langle n \rangle^{-d-2s_0+\sigma} |h|^\sigma \quad \forall n \in \mathbb{Z}^d$$

then $X \in C([-T, T]; W^{s, \infty}(\mathbb{T}^d))$

$\forall |h| \leq 1$
 $\forall t \in [-T, T]$

for $\underline{s < s_0 - \sigma/2}$, a.s.

Furthermore, if $\exists \theta > 0$ s.t.

$$\mathbb{E} \left[|f_h \hat{X}_N(t, n) - f_h \hat{X}(t, n)|^2 \right] \lesssim N^{-\theta} \langle n \rangle^{-d-2s_0+\sigma} |h|^\sigma$$

then,

$$\forall n \in \mathbb{Z}^d, |h| \leq 1, t \in [-T, T]$$

X_N converges to X in $C([-T, T]; W^{s, \infty}(\mathbb{T}^d))$

for $s < s_0 - \frac{\sigma}{2}$, a.s.

Note: $W^{s, \infty}(\mathbb{T}^d)$ can be replaced by $C^s(\mathbb{T}^d) = B_{\infty, \infty}^s(\mathbb{T}^d)$

Mourrat-Weber-Xu '17. Oh-Okamoto-Tzvetkov

2-d SNLW:

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$$\Psi(t, x) = \int_0^t \frac{\sin(t-t') \langle \nabla \rangle}{\langle \nabla \rangle} dW(t') \in C_T W_x^{s, \infty}(\mathbb{T}^2), \text{ a.s.}$$

$s < 0.$

$$\boxed{\Psi_N = P_{\leq N} \Psi}$$

We set

$$\Psi_N^k(t, x) := H_{\frac{k}{N}}(\Psi_N(t, x); \sigma_N(t))$$

where

$$\sigma_N(t) = \mathbb{E}[|\Psi_N(t, x)|^2] \sim \log N$$

Prop: $\{\Psi_N^k\}_{N \in \mathbb{N}}$ forms a Cauchy seq

in $L^p(\Omega; C_T W_x^{-\varepsilon, \infty})$, $p < \infty$.

and also almost surely.

Gubinelli-Koch-Obermann
TAMS '18

Pf: We only verify ~~(*)~~ in Lemma
regularity lemma

$$\mathbb{E} \left[\left| \widehat{\Psi_N^k}(t, n) \right|^2 \right]$$

$$= \iint_{\mathbb{T}_x^2 \times \mathbb{T}_y^2} \mathbb{E} \left[\underbrace{:\Psi_N^k(t, x): : \Psi_N^k(t, y):}_{\text{Lemma}} \right] e_n(y-x) dx dy$$

$$= k! \left\{ \mathbb{E} \left[\Psi_N(t, x) \Psi_N(t, y) \right] \right\}^k$$

$$= \sum_{|m| \leq N} \int_0^t \left(\frac{\sin(t-t') \langle m \rangle}{\langle m \rangle} \right)^2 dt' e_m(x-y)$$

$$\lesssim \sum_{n_1 + \dots + n_k = n} \prod_{j=1}^k \frac{1}{\langle m_j \rangle^2} \lesssim \langle n \rangle^{-2+\varepsilon} \quad \parallel -250.$$

$\Rightarrow \Psi_N(t) \in W^{s, \infty}(\mathbb{T}^d), s < 0, a.s.$

□

A word on the summation:

$$\sum_{\substack{n_1 + \dots + n_k = n \\ n_j \in \mathbb{Z}^2}} \prod_{j=1}^k \frac{1}{\langle m_j \rangle^2}$$

k-1 summations

• Divide the sum into regions I_ℓ
 $I_\ell = \{ |m_\ell| \geq \max_{j=1, \dots, k} |m_j| \}$

• Under $n_1 + \dots + n_k = n$, we have $\max_{j=1, \dots, k} |m_j| \geq |n|$

$$\sum_{I_\ell} \prod_{j=1}^k \frac{1}{\langle m_j \rangle^2} \lesssim \frac{1}{\langle n \rangle^{2-\varepsilon}} \sum_{\substack{I_\ell \\ j \neq \ell}} \prod_{j=1}^k \frac{1}{\langle m_j \rangle^{2+\frac{\varepsilon}{k-1}}} \lesssim \frac{1}{\langle n \rangle^{2-\varepsilon}}$$

• One can also use the following lemma (Part(i)) iteratively.

Lemma:

(i) $\alpha, \beta \in \mathbb{R}$ s.t. $\alpha + \beta > d$ and $\alpha, \beta < d$

Then,
$$\sum_{\substack{n=n_1+n_2 \\ n_j \in \mathbb{Z}^d}} \frac{1}{\langle m_1 \rangle^\alpha} \frac{1}{\langle m_2 \rangle^\beta} \lesssim \langle n \rangle^{d-\alpha-\beta}$$

resonant
case

(ii) $\alpha, \beta \in \mathbb{R}$ s.t. $\alpha + \beta > d$.

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Then,

$$\sum_{\substack{n = n_1 + n_2 \\ |n_1| \sim |n_2| \\ n_j \in \mathbb{Z}^d}} \frac{1}{\langle n_1 \rangle^\alpha \langle n_2 \rangle^\beta} \approx \langle n \rangle^{d - \alpha - \beta}.$$

This lemma follows from elementary computation.

See Moumat - Weber - Xu '17 (Also, see Ginibre - Tautsumi - Velo

JFA '97
Lemma 4.2.