

Directed Polymer in Random Environment with Correlation

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Basic notations

- $S = (S_k), k \geq 0$: a nearest-neighbor path starting from the origin in \mathbb{Z}^d
- $\omega = \{\omega(i, x), (i, x) \in \mathbb{N} \times \mathbb{Z}^d\}$: a family of real-valued random variables appearing as the environment
- $H_n^\omega(S) = \sum_{i=1}^n \omega(i, S_i)$: the n -step energy of a path S for a fixed environment ω
- $\mathbb{P}_n^\omega(S) = \frac{1}{Z_n^\omega(\beta)} e^{\beta H_n^\omega(S)} \mathbb{P}(S)$: the random polymer measure
- partition function:

$$Z_n^\omega(\beta) = \sum_S e^{\beta H_n^\omega(S)} \mathbb{P}(S) = \mathbb{E}_\mathbb{P} e^{\beta H_n^\omega(S)} \quad (1)$$

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Basic Results

- $W_n = Z_n^\omega(\beta) / \mathbb{E}(Z_n^\omega(\beta))$ is a supermartingale,
- $W_\infty = 0$: termed as strong disorder, $\beta > \beta_c$
- $W_\infty > 0$: termed as weak disorder, $\beta < \beta_c$
- $\beta_c = 0$, for $d = 1, 2$
- $0 < \beta_c < \infty$, for $d \geq 3$
- For more details see

[1] Hubert Lacoin(2010). New bounds for the free energy of directed polymers in dimension $1 + 1$ and $1 + 2$. *Comm. Math. Phys.*, 294(2):471-503.

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Recent results

An intermediate disorder regime: proposed in

[2]T. Alberts, K. Khanin and J. Quastel, THE INTERMEDIATE DISORDER REGIME FOR DIRECTED POLYMERS IN DIMENSION $1 + 1$, *The Annals of Probability* 2014, Vol. 42, No. 3, 1212-1256

[3]F. Caravenna, R. Sun, N. Zygouras, Polynomial chaos and scaling limits of disordered systems, *J. Eur. Math. Soc.* 19, 1-65,2017

Recent results

It says in [2]

Convergence of partition functions

- Under scaling $\beta_n = \beta n^{-\frac{1}{4}}$, the partition function

$$e^{-n\lambda(\beta n^{-\frac{1}{4}})} Z_n^\omega(\beta n^{-\frac{1}{4}}) \xrightarrow{D} Z_{\sqrt{2}\beta}.$$

$\lambda(\cdot)$: log Laplace of the environment variables.

$Z_{\sqrt{2}\beta}$: has explicit Wiener chaos decomposition.

Basic ideas

Using

$$e^x \approx 1 + x,$$

they have

modified function

$$\mathfrak{Z}_n^\omega(\beta n^{-\frac{1}{4}}) = \mathbb{E}_{\mathbb{P}} \left[\prod_{i=1}^n (1 + \beta n^{-\frac{1}{4}} \omega(i, \mathbf{S}_i)) \right]. \quad (2)$$

Then expanding it as,

Basic ideas

Expanding modified function

$$\begin{aligned}
 & \mathfrak{Z}_n^\omega(\beta n^{-\frac{1}{4}}) & (3) \\
 & = \mathbb{E}_{\mathbb{P}} \left[1 + \sum_{k=1}^n \beta^k n^{-\frac{k}{4}} \sum_{\mathbf{i} \in D_k^n} \prod_{j=1}^k \omega(\mathbf{i}_j, S_{\mathbf{i}_j}) \right] \\
 & = 1 + \sum_{k=1}^n \beta^k n^{-\frac{k}{4}} \sum_{\mathbf{i} \in D_k^n} \sum_{\mathbf{x} \in \mathbb{Z}^k} \omega(\mathbf{i}, \mathbf{x}) p_k(\mathbf{i}, \mathbf{x}).
 \end{aligned}$$

where

$$D_k^n = \{\mathbf{i} = (i_1, i_2, \dots, i_k) \in [n]^k : 1 \leq i_1 < i_2 < \dots < i_k \leq n\}.$$

Basic ideas

U-statistics

Let $\mathcal{S}_k^n(p_k^n) = \sum_{\mathbf{i} \in D_k^n} \sum_{\mathbf{x} \in \mathbb{Z}^k} \omega(\mathbf{i}, \mathbf{x}) p_k(\mathbf{i}, \mathbf{x})$. Then, in [2]

$$\begin{aligned} & \mathfrak{Z}_n^\omega(\beta n^{-\frac{1}{4}}) \\ &= 1 + \sum_{k=1}^n \beta^k n^{-\frac{3k}{4}} \mathcal{S}_k^n(n^{-\frac{k}{2}} p_k^n) \\ & \xrightarrow{D} \mathcal{Z}_{2\beta} = \sum_k \mathbf{I}_k(p_k). \end{aligned} \quad (4)$$

p_k^n, p_k : k -order transition probability of random walk and density of Brownian motion, respectively.

\mathbf{I}_k : k -multiple Wiener-Itô integral with respect to space-time white noise.

Basic ideas

Four-parameter fields

Let $\mathfrak{Z}^\omega(m, y; k, x; \beta) = \mathbb{P}[\prod_{i=m+1}^k (1 + \beta\omega(i, S_i)) \mathbf{1}_{S_k=x} | S_m = y]$. Similarly, we have $Z^\omega(m, y; k, x; \beta)$, $\mathfrak{Z}^\omega(k, x; \beta)$, $Z^\omega(k, x; \beta)$.

Theorem (AKQ 2014)

Assuming that the ω have six moments with mean zero and variance one, the fields for $0 \leq s < t \leq 1$, $x, y \in \mathbb{R}$

$$(s, y; t, x) \longrightarrow \frac{\sqrt{n}}{2} \mathfrak{Z}^\omega(ns, y\sqrt{n}; nt, x\sqrt{n}; \beta n^{-1/4})$$

converge weakly as $n \longrightarrow \infty$ to a random field $Z_{\sqrt{2}\beta}(s, y; t, x)$.

Basic ideas

Four-parameter fields

Let $\mathfrak{Z}^\omega(m, y; k, x; \beta) = \mathbb{P}[\prod_{i=m+1}^k (1 + \beta\omega(i, S_i)) 1_{S_k=x} | S_m = y]$.

Theorem (continued, AKQ 2014)

$Z_{\sqrt{2}\beta}(s, y; t, x)$ satisfies the following stochastic heat equation driven by white noise

$$\frac{\partial u(s, y; t, x)}{\partial t} = \frac{1}{2} \Delta u(s, y; t, x) + \sqrt{2}\beta u(s, y; t, x) \dot{W}(t, x) \quad (5)$$

$$u(s, y; s, y) = \delta(t - s, x - y) \quad (6)$$

Correlated environment

Environment

$\{\xi_{i,j} : i \in \mathbb{N}, j \in \mathbb{Z}\}$: i.i.d. with $\mathbb{E}_{\mathbb{Q}} \xi_{i,j} = 0$ and $\mathbb{E}_{\mathbb{Q}} \xi_{i,j}^2 = 1$ for any i, j .

$\omega = \{\omega(\mathbf{n}, \mathbf{x}) : \mathbf{n} \geq 0, \mathbf{x} \in \mathbb{Z}\}$: a stationary field by

$$\omega(\mathbf{n}, \mathbf{x}) = \sum_{y=-\infty}^{\infty} \psi_{y-x} \xi_{\mathbf{n}, y}, \quad (7)$$

with $\psi_j \sim \delta |j|^{-\alpha}$ and $1/2 < \alpha < 1$. Then, one has

$$\mathbb{E}(\omega(\mathbf{i}, \mathbf{x}) \omega(\mathbf{j}, \mathbf{y})) = \delta_{ij} \gamma(\mathbf{x} - \mathbf{y}),$$

where δ_{ij} is Kronecker and $\gamma(k) \sim \lambda |k|^{1-2\alpha}$ for large integer k and $\lambda = \delta^2 \frac{\Gamma(2\alpha-1)\Gamma(1-\alpha)}{\Gamma(\alpha)}$.

Correlated environment

Spectral measure of γ

Let $G(d\eta)$ be the spectral measure of the correlation function γ , i.e.,

$$\gamma(\mathbf{k}) = \int_{-\pi}^{\pi} e^{i\mathbf{k}\eta} G(d\eta), \quad \forall \mathbf{k} \in \mathbb{Z}. \quad (8)$$

For every $N \in \mathbb{N}$, we define a new measure G_N by

$$G_N(A) = N^{\alpha-1/2} G(N^{-1/2}A), \quad A \in \mathcal{B}(\mathbb{R}).$$

Then, $\lim_{N \rightarrow \infty} G_N = G_0$ (locally finite measure). Furthermore, G_0 has a spectral density $D^{-1}|\eta|^{1-2H}$ with $D = 2\Gamma(2-2H)\cos(1-H)\pi$, which is exactly the spectrum of fractional Brownian motion with Hurst parameter $H > 1/2$.

Correlated environment

A central limit theorem

Let ω be given by (7), S be the symmetrical nearest- random walk on \mathbb{Z} started at the origin under probability measure \mathbb{P} , and let $\varrho = H/2$. Then

$$n^{-\varrho} \beta \sum_{i=1}^n \sum_{x \in \mathbb{Z}} \omega(i, x) \mathbb{P}(S_i = x) \xrightarrow{\mathcal{D}} \mathbf{N}(0, \sigma^2) \quad (9)$$

with $\sigma^2 = \frac{4\beta^2 \Gamma(1-H/2)}{DH}$.

The proof consists of the computation of the variance and verification of Lindeberg's condition.

Remark

In the case of iid, the CLT holds with $\varrho = 1/4$, $\sigma^2 = 2\beta^2 / \sqrt{\pi}$.

Fractional Gaussian fields

- A time-space fractional Brownian random field $W = \{W(t, x) : t \geq 0, x \in \mathbb{R}\}$ defined on some probability space $(\Omega_H, \mathcal{F}_H, \mathbb{P}_H)$ is a mean zero Gaussian field with covariance

$$\mathbb{E}_H(W(t, x)W(s, y)) = \frac{1}{2}(s \wedge t)(|x|^{2H} + |y|^{2H} - |x - y|^{2H}),$$

$H \in (0, 1)$: Hurst parameter.

- Introduce the following Hilbert space:

$$\mathcal{L}_H = \{f : \|f\|_H^2 = \int_0^1 \int_{\mathbb{R}} \int_{\mathbb{R}} f(s, u)K(u, v)f(s, v)dsdudv < \infty\},$$

where $K(u, v) = H(2H - 1)|u - v|^{2H-2}$.

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Stochastic integral

- For $f \in \mathcal{L}_H$, the stochastic integral $\int_0^1 \int_{\mathbb{R}} f(t, x) W(dt dx) := W(f)$ is defined as usual with

$$\mathbb{E}_H \left[\int_0^1 \int_{\mathbb{R}} f(t, x) W(dt dx) \right]^2 = \int_0^1 \int_{\mathbb{R}^2} f(s, u) K(u, v) f(s, v) ds du dv.$$

- Symmetric tensor product of \mathcal{L}_H .

$$\begin{aligned} \mathcal{L}_H^{\otimes k} = \{ & f : ([0, 1] \times \mathbb{R})^k \rightarrow \mathbb{R}; \\ & \int_{[0, 1]^k} \int_{\mathbb{R}^{2k}} f(t_1, x_1, t_2, x_2, \dots, t_k, x_k) \\ & \prod_{i=1}^k K(x_i, y_i) f(t_1, y_1, t_2, y_2, \dots, t_k, y_k) dt dx dy < \infty \} \end{aligned}$$

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Stochastic integral

- Multiple Ito integral $I_k(f^{\otimes k}) = H_k(W(f))$ for $f \in \mathcal{L}_H$, H_k , k -order Hermite polynomial. Then define $I_k(f) = \int_{([0,1] \times \mathbb{R})^k} f(\mathbf{t}, \mathbf{x}) W^{\otimes k}(d\mathbf{t}d\mathbf{x})$ for general $f \in \mathcal{L}_H^{\otimes k}$ by density argument with

$$\mathbb{E}(I_k(f)I_k(g)) = k! \langle f, g \rangle_H .$$

- r -order contraction of two symmetry functions f and g by

$$\begin{aligned} & f \otimes_r g(t_1, x_1; \dots; t_{m+n-2r}, x_{m+n-2r}) \\ = & \text{Sym} \left\{ \int_{[0,1]^r} \int_{\mathbb{R}^{2r}} f(t_1, x_1; \dots; t_{n-r}, x_{n-r}; s_1, u_1; \dots; s_r, u_r) \right. \\ & \times g(t_1, x_1; \dots; t_{m-r}, x_{m-r}; s_1, u_1; \dots; s_r, u_r) \prod_{i=1}^r K(u_i, v_i) \\ & f(t_1, x_1; \dots; t_{n-r}, x_{n-r}; \tau_1, v_1; \dots; \tau_r, v_r) \\ & \left. \times g(t_1, x_1; \dots; t_{m-r}, x_{m-r}; \tau_1, v_1; \dots; \tau_r, v_r) ds d\tau dudv \right\}, \end{aligned}$$

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Recursive identities



$$I_n(f)I_m(g) = \sum_{r=0}^{m \wedge n} r! \binom{n}{r} \binom{m}{r} I_{n+m-2r}(f \otimes_r g) \quad (10)$$

for $f \in \mathcal{L}_H^{\otimes m}$, $g \in \mathcal{L}_H^{\otimes n}$. Especially, when $m = 1$, it is reduced to

$$I_n(f)I_1(g) = I_{n+1}(f \otimes g) + nI_{n-1}(f \otimes_1 g). \quad (11)$$



$$\begin{aligned} & (f_1^{\otimes m} \otimes f_2^{\otimes(n-1)}) \otimes_1 f_2 \\ &= \frac{m}{m+n-1} f_1^{\otimes(m-1)} \otimes f_2^{\otimes(n-1)} \langle f_1, f_2 \rangle_H \\ & \quad + \frac{n-1}{m+n-1} f_1^{\otimes m} \otimes f_2^{\otimes(n-2)} \|f_2\|_H^2. \end{aligned}$$

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Recursive identities

- Furthermore, according to (11), we have

$$\begin{aligned}
 & \mathbf{I}_{m+n}(\mathbf{f}_1^{\otimes m} \otimes \mathbf{f}_2^{\otimes n}) \\
 = & \mathbf{I}_{m+n-1}(\mathbf{f}_1^{\otimes m} \otimes \mathbf{f}_2^{\otimes(n-1)})\mathbf{I}_1(\mathbf{f}_2) \\
 & - m \langle \mathbf{f}_1, \mathbf{f}_2 \rangle_{\mathbf{H}} \mathbf{I}_{m+n-2}(\mathbf{f}_1^{\otimes(m-1)} \otimes \mathbf{f}_2^{\otimes(n-1)}) \quad (12) \\
 & - (n-1) \|\mathbf{f}_2\|_{\mathbf{H}}^2 \mathbf{I}_{m+n-2}(\mathbf{f}_1^{\otimes m} \otimes \mathbf{f}_2^{\otimes(n-2)})
 \end{aligned}$$

Chaos expansion

Proposition

Let W be the gaussian random field above with spatial parameter $1/2 < H < 1$. Let $(\Omega_H, \mathcal{F}_H, P_H)$ be the canonical probability space corresponding to W . Then for any $F \in L^2(\Omega_H)$, it admits the following chaos expansion:

$$F = \sum_{k=0}^{\infty} I_k(f_k),$$

where $f_k \in \mathcal{L}_H^{\otimes k}$, $k = 0, 1, \dots$, and the series converges in $L^2(\Omega_H, \mathcal{F}_H, P_H)$. Moreover,

$$\mathbb{E}_H[F^2] = \sum_{k=0}^{\infty} k! \|f_k\|_H^2.$$

Stochastic heat equations

Mild solution

We turn to stochastic heat equations (5) with multiplicative noise with initial value $u(s, x) = u(x)$, $0 \leq s \leq t \leq 1, x \in \mathbb{R}$. Its solution is formulated in the mild form, i.e.,

$$u(t, x; s) = P_{t-s}u(x) + \beta \int_s^t \int_{\mathbb{R}} P_{t-r}(x - z)u(r, z)W(dr, dz), \quad (13)$$

where $P_t(x) = \frac{1}{\sqrt{2\pi t}}e^{-\frac{x^2}{2t}}$ and $P_t f(x) = \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi t}}e^{-\frac{(x-y)^2}{2t}}f(y)dy$. If, furthermore, let $u(x) = \delta(x - y)$, we get a four-parameter field $u(t, x; s, y)$ by

$$u(t, x; s, y) = P_{t-s}(x - y) + \beta \int_s^t \int_{\mathbb{R}} P_{t-r}(x - z)u(r, z; s, y)W(drdz).$$

Chaos expansion for solution

$$\begin{aligned}
 & u(t, \mathbf{x}; s, y) \\
 &= P_{t-s}(\mathbf{x} - y) \\
 &+ \sum_{k=1}^{\infty} \beta^k \int_{\Delta(s, t]^k} \int_{\mathbb{R}^k} \prod_{i=1}^k P_{t_i - t_{i-1}}(\mathbf{x}_i - \mathbf{x}_{i-1}) P_{t-t_k}(\mathbf{x} - \mathbf{x}_k) W(dt_i d\mathbf{x}_i) \\
 &= P_{t-s}(\mathbf{x} - y) + \sum_{k=1}^{\infty} \beta^k I_k(\widetilde{P}_k(t, \mathbf{x}; s, y))
 \end{aligned}$$

with $\Delta(s, t]^k = \{s < t_1 < \dots < t_k < t\}$, $\mathbf{x}_0 = y$, and

$$\begin{aligned}
 P_k(t, \mathbf{x}; s, y; t_1, \dots, t_k; \mathbf{x}_1, \dots, \mathbf{x}_k) &= \prod_{i=1}^k P_{t_i - t_{i-1}}(\mathbf{x}_i - \mathbf{x}_{i-1}) P_{t-t_k}(\mathbf{x} - \mathbf{x}_k) \\
 &\stackrel{\Delta}{=} P_k(t, \mathbf{x}; s, y; \tau; \mathbf{x})
 \end{aligned}$$

The convergence of partition functions

Theorem 1

Let $\{u(t, x), (t, x) \in [0, 1] \times \mathbb{R}\}$ be the solution to (5) with parameter 2β , initial data $u(x) = \delta(x)$. And let Z_n^ω be the partition function (1) of random polymer in the random environment $\{\omega(n, x) : n \geq 0, x \in \mathbb{Z}\}$ with the representation (7). Then

$$e^{-n\lambda(\beta n^{-e})} Z_n^\omega(\beta n^{-e}, tn, x\sqrt{n}) \longrightarrow u(t, x) \quad n \longrightarrow \infty,$$

in the sense of fdd.

Convergence of modified partition functions

Theorem 2

Let $\{u(t, x), (t, x) \in [0, 1] \times \mathbb{R}\}$ be the solution to (5) with parameter $\sqrt{2}\beta$, initial data $u(x) = \delta(x)$. Then

$$\mathfrak{Z}_n^\omega(\beta n^{-e}, tn, x\sqrt{n}) \longrightarrow u(t, x) \quad \text{as } n \longrightarrow \infty.$$

Tightness

For $0 \leq t \leq 1$, $x \in \mathbb{R}$, define two-parameter fields by

$$z_n(t, x) := \sqrt{n} \mathfrak{Z}_n^\omega(nt, \sqrt{nx}; \beta n^{-e}).$$

Then, we have

$$z_n(t, x) = p_n(t, x) + n^{-\frac{1}{2}} \beta \sum_{\substack{s \in [0, t] \cap n^{-1} \mathbb{Z} \\ y \in n^{-1/2} \mathbb{Z}}} p_n(t - s, x - y) \bar{z}_n(s, y) \omega_n(s, y),$$

where $\omega_n(s, y) = n^{-e} \omega(ns, \sqrt{ny})$. $\bar{z}_n(s, y)$ corresponding to $\bar{\mathfrak{Z}}_n^\omega(k, x; \beta)$, and $\bar{\mathfrak{Z}}_n^\omega(k, x; \beta) = \frac{1}{2} [\mathfrak{Z}_n^\omega(k + 1, x; \beta) + \mathfrak{Z}_n^\omega(k - 1, x; \beta)]$.

Tightness

Theorem 3

Let $\epsilon > 0$ be small enough. For any $n \in \mathbb{N}$, $t, s \in [\epsilon, 1]$ and $x, y \in \mathbb{R}$, for some $q > 1$, there exist constant $C_\epsilon > 0$, $0 < \iota < H$, such that

$$\mathbb{E}|z_n(t, x) - z_n(s, y)|^{2q} \leq C_\epsilon (|t - s|^{Hq} + |x - y|^{\iota q}). \quad (14)$$

Moreover, if $2q$ -order moment of ω is finite for $q > \frac{2}{H}$, then the family of process $\{z_n\}_{n=1}^\infty$ is tight in $C([\epsilon, 1], \mathbb{R})$.

Convergence of U -statistics in correlated case

weighted U -statistics \mathcal{S}_k^n by

$$\mathcal{S}_k^n(f) = 2^{k/2} \sum_{\mathbf{i} \in \mathbb{E}_k^n} \sum_{\mathbf{x} \in \mathbb{Z}^k} \bar{f}_n\left(\frac{\mathbf{i}}{n}, \frac{\mathbf{x}}{\sqrt{n}}\right) \omega(\mathbf{i}, \mathbf{x}) 1_{\{\mathbf{i} \leftrightarrow \mathbf{x}\}}. \quad (15)$$

where \bar{f}_n is the conditional expectation of $f \in L^2([0, 1]^k \times \mathbb{R}^k)$ with respect to the sigma algebra generated by

$$\mathfrak{R}_k^n \triangleq \left\{ \left(\frac{\mathbf{i}-1}{n}, \frac{\mathbf{i}}{n} \right] \times \left(\frac{\mathbf{x}-1}{\sqrt{n}}, \frac{\mathbf{x}+1}{\sqrt{n}} \right] : \mathbf{i} \in \mathbb{D}_k^n, \mathbf{i} \leftrightarrow \mathbf{x} \right\}.$$

$$\bar{f}_n(\mathbf{t}, \mathbf{x}) = \frac{1}{|\mathbb{R}|} \int_{\mathbb{R}} f d\mathbf{t} d\mathbf{x}, \quad (\mathbf{t}, \mathbf{x}) \in \mathbb{R} \in \mathfrak{R}_k^n,$$

Convergence of U -statistics in correlated case

Theorem 4

Let $f \in \mathcal{H}^k$. Then, as $n \rightarrow \infty$,

$$n^{\frac{-(H+1)k}{2}} \mathcal{S}_k^n(f) \xrightarrow{D} \int_{[0,1]^k} \int_{\mathbb{R}^k} f(\mathbf{t}, \mathbf{x}) W^{\otimes k}(d\mathbf{t}d\mathbf{x}) = I_k(f), \quad (16)$$

where $W(d\mathbf{t}d\mathbf{x})$ is fractional Gaussian noise (??) and $I_k, k = 1, 2, \dots$, is k -multiple integral.

The proof is more complicated than that of the case i.i.d. environment since the correlation exists between two disjoint spatial areas.

The way to Theorem 4

- $k = 1$: Show it is true for $f = 1_{\{t_0 \leq t \leq t_1, x_0 \leq x \leq x_1\}}$ for some $0 \leq t_0 < t_1 \leq 1, x_0 < x_1$. In this case,

$$\mathcal{S}_1^n(f) = 2^{1/2} \sum_{\{i \in E_k^n, n t_0 \leq i \leq n t_1\}} \sum_{\{x \in \mathbb{Z}, \sqrt{nx_0} \leq x \leq \sqrt{nx_1}\}} \omega(i, x) 1_{\{i \leftrightarrow x\}}$$

We take the method used in the proof of CLT in this talk.

- $k > 1$: We show

$$n^{-\frac{(H+1)k}{2}} \mathcal{S}_k^n(g^{\otimes k}) \longrightarrow I_k(g^{\otimes k}) \quad \text{as } n \longrightarrow \infty.$$

with

$$g(t, x) = 1_{\{t_0 \leq t \leq t_1, x_0 \leq x \leq x_1\}}(t, x) \in \mathcal{L}_H$$

for $\forall t_0 \leq t_1, x_0 \leq x_1$ by induction

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The way to Theorem 4

- We show it holds for $f \in \mathcal{L}_H^k$ of the form $f = g_1^{\otimes k_1} \otimes \cdots \otimes g_s^{\otimes k_s}$ with g_1, \dots, g_s being indicators as $k = 1$ of some disjoint rectangles, $k_1 + \cdots + k_s = k, k_1 > 0, \dots, k_s > 0, s = 2, \dots$
- We need to show for $s = 2$

$$\begin{aligned}
 & n^{-\frac{(H+1)(m+l)}{2}} \mathcal{S}_{m+l}^n(f^{\otimes m} \otimes g^{\otimes l}) \\
 & \longrightarrow I_{m+l-1}(f^{\otimes m} \otimes g^{\otimes(l-1)}) I_1(g) \\
 & \quad - m I_{m+l-2}(f^{\otimes(m-1)} \otimes g^{\otimes(l-1)}) \langle f, g \rangle_H \\
 & \quad - (l-1) I_{m+l-2}(f^{\otimes m} \otimes g^{\otimes(l-2)}) \|g\|_H^2.
 \end{aligned}$$

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The way to Theorem 4

Lemma

For all fixed k, n , $\mathcal{S}_k^n(f)$ is linear in h with probability one, and, by the definition of ω , for $k_1 \neq k_2$, $\mathbb{E}_{\mathbb{Q}}(\mathcal{S}_{k_1}^n(f_1)\mathcal{S}_{k_2}^n(f_2)) = 0$ for $f_i \in \mathcal{L}_H^{\otimes k_i}$, $i = 1, 2$. Furthermore, for $k_1 = k_2 = k$, we have

$$\mathbb{E}_{\mathbb{Q}}[(\mathcal{S}_k^n(f))^2] \leq C\lambda^k n^{(1+H)k} \|f\|_{\mathcal{H}^k}^2$$

for some generic positive constant C .

- Based on the lemma and previous discussion for general $f \in \mathcal{L}_H^k$

$$n^{-\frac{(H+1)k}{2}} \mathcal{S}_k^n(f) \xrightarrow{D} \int_{[0,1]^k} \int_{\mathbb{R}^k} f(\mathbf{t}, \mathbf{x}) W^{\otimes k}(d\mathbf{t}d\mathbf{x}) = I_k(f),$$

is true.

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- Based on the lemma and previous discussion for general $f \in \mathcal{L}_H^k$

$$n^{\frac{-(H+1)k}{2}} \mathcal{S}_k^n(f) \xrightarrow{D} \int_{[0,1]^k} \int_{\mathbb{R}^k} f(\mathbf{t}, \mathbf{x}) W^{\otimes k}(d\mathbf{t}d\mathbf{x}) = I_k(f),$$

is true.

Convergence of $\mathfrak{Z}_n^\omega(\beta n^{-\varrho})$

It is easy to get

$$\begin{aligned}
 \mathfrak{Z}_n^\omega(\beta n^{-\varrho}, tn, x\sqrt{n}) &= 1 + \sum_{k=1}^n \beta^k n^{-k\varrho} \sum_{\mathbf{i} \in D_k^n} \sum_{\mathbf{x} \in \mathbb{Z}^k} \left[\prod_{j=1}^k \omega(i_j, x_j) p_k(\mathbf{i}, \mathbf{x}) \right] \\
 &= 1 + \sum_{k=1}^n 2^k \beta^k n^{-\frac{k(H+1)}{2}} \mathcal{S}_k^n(n^{\frac{k}{2}} p_k^n) \\
 &\longrightarrow 1 + \sum_{k=1}^{\infty} 2^k \beta^k I_k(p_k(t, x, \cdot)),
 \end{aligned}$$

where $p(t, x)$ is Brownian motion density and the last expression is just the chaos expansion of mild solution $u(t, x)$ to stochastic heat equation (5) driven by fractional Gaussian noise.

Hermite expansion

Let

$$\tilde{\omega}_n(i, x) = \frac{e^{\beta n^{-e} \omega(i, x) - \lambda(\beta n^{-e})} - 1}{\beta n^{-e}} \triangleq F^{(n)}(\omega(i, x)), \quad (17)$$

where $\lambda(\cdot)$ is the Log-Laplace of $\omega(i, x)$. Thus, we get a mean zero stationary field $\tilde{\omega}_n(i, x)$ (n -dependent), which is a non-linear functional of $\omega(i, x)$. The covariance of $\tilde{\omega}_n(i, x)$ and $\tilde{\omega}_n(i, y)$ is given by

$$\begin{aligned} \mathbb{E}_{\mathbb{Q}}(\tilde{\omega}_n(i, x)\tilde{\omega}_n(i, y)) &= \frac{1}{\beta^2 n^{-2e}} \mathbb{E}_{\mathbb{Q}}\{e^{\beta n^{-e}(\omega_n(i, x) + \omega_n(i, y)) - 2\lambda(\beta n^{-e})} - 1\} \\ &= \gamma(x - y)(1 + o(1)) := \tilde{\gamma}_n(x - y). \end{aligned}$$

Hermite expansion

We can expand $F^{(n)}(z), z \in \mathbb{R}$, by

$$F^{(n)}(z) = \frac{1}{\beta n^{-\varrho}} \sum_{k=1}^{\infty} (\beta n^{-\varrho})^k A_k(z),$$

where $A_k(z), k \in \mathbb{N}$, is the system of Appell polynomials related to the distribution of ω with $A_0 = 1$. Let $c_k, k \in \mathbb{N}$, be the expansion coefficients of F^n with respect to Appell system $A_k, k \in \mathbb{N}$. We remark that the Appell rank, which is the least index k such that $c_k \neq 0$, of $F^{(n)}$ is 1. Now by (17), we have

$$e^{-n\lambda(\beta n^{-\varrho})} \mathcal{Z}_n^\omega = e^{-n\lambda(\beta n^{-\varrho})} \mathbb{E}_{\mathbb{Q}} e^{\beta n^{-\varrho} \sum_{i=1}^n \omega(i, S_i)} = \mathbb{E}_{\mathbb{Q}} \prod_{i=1}^n (1 + \beta n^{-\varrho} \tilde{\omega}_n(i, S_i)). \quad (18)$$

Modified partition functions for $\tilde{\omega}_n(i, \mathbf{x})$

$$\begin{aligned}
 \mathfrak{Z}_n^{\tilde{\omega}_n}(\beta n^{-\varrho}) &= \mathbb{E}_{\mathbb{P}} \left[\prod_{i=1}^n (1 + \beta n^{-\varrho} \tilde{\omega}_n(i, S_i)) \right] \\
 &= \mathbb{E}_{\mathbb{P}} \left[1 + \sum_{k=1}^n \beta^k n^{-k\varrho} \sum_{\mathbf{i} \in D_k^n} \prod_{j=1}^k \tilde{\omega}_n(i_j, S_{i_j}) \right] \\
 &= 1 + \sum_{k=1}^n \beta^k n^{-k\varrho} \sum_{\mathbf{i} \in D_k^n} \sum_{\mathbf{x} \in \mathbb{Z}^k} \left[\prod_{j=1}^k \tilde{\omega}_n(i_j, x_j) p_k(\mathbf{i}, \mathbf{x}) \right],
 \end{aligned}$$

and the corresponding weighted U -statistics \mathcal{S}_k^n by

$$\mathcal{S}_k^n(f, \tilde{\omega}_n) = 2^{k/2} \sum_{\mathbf{i} \in E_k^n} \sum_{\mathbf{x} \in \mathbb{Z}^k} \bar{f}_n\left(\frac{\mathbf{i}}{n}, \frac{\mathbf{x}}{\sqrt{n}}\right) \tilde{\omega}_n(\mathbf{i}, \mathbf{x}) \mathbf{1}_{\{\mathbf{i} \leftrightarrow \mathbf{x}\}}.$$

Modified partition functions for $\tilde{\omega}_n(i, x)$

Theorem 5

For $f \in \mathcal{L}_H^{\otimes k}$, $k \in \mathbb{N}$, $\tilde{\omega}_n$ is defined by (17). Then

$$n^{-\frac{k(H+1)}{2}} \mathcal{S}_k^n(f, \tilde{\omega}_n) \xrightarrow{D} I_k(f), \quad \text{as } n \rightarrow \infty.$$

Sketch of the proof: We only show it holding for $k = 1$ and f of the form $f(t, x) = 1_{\{t_0 \leq t \leq t_1, x_0 \leq x \leq x_1\}}$ for some $0 \leq t_0 \leq t_1, x_0 \leq x_1 \in \mathbb{R}$ as before.

$$\mathcal{S}_1^n(f, \tilde{\omega}_n) = 2^{1/2} \sum_{i \in E_k^n, nt_0 \leq i \leq nt_1} \sum_{x \in \mathbb{Z}, \sqrt{nx_0} \leq x \leq \sqrt{nx_1}} \tilde{\omega}_n(i, x) 1_{\{i \leftrightarrow x\}}.$$

It is easy to show that

$$n^{-(H+1)} \mathbb{E}_{\mathbb{Q}}[(\mathcal{S}_1^n(f, \tilde{\omega}_n))^2] \rightarrow \frac{\lambda(t_1 - t_0)(x_1 - x_0)^{2H}}{H(2H - 1)} = \mathbb{E}_H \left(\int_0^1 \int_R f(t, x) W(dt dx) \right)^2$$

Modified partition functions for $\tilde{\omega}_n(i, x)$

Sketch of the proof: By following Dobrushin's lines (see[?]) to show the normal asymptotics.

$$c_1^{(n)} = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} H_1(x) \frac{e^{\beta n^{-e} x - \beta^2 n^{-2e} / 2} - 1}{\beta n^{-e}} e^{-\frac{x^2}{2}} dx = 1 \neq 0.$$

We split $n^{\frac{-(H+1)}{2}} \mathcal{S}_1^n(f, \tilde{\omega}_n)$ as the sum of $n^{\frac{-(H+1)}{2}} \mathcal{S}_1^n(f)$ and $\mathcal{R}_n^{\tilde{\omega}_n}$.

$$\mathcal{R}_n^{\tilde{\omega}_n} = 2^{1/2} \sum_{i \in E_k^n, nt_0 \leq i \leq nt_1} \sum_{x \in \mathbb{Z}, \sqrt{nx_0} \leq x \leq \sqrt{nx_1}} \sum_{j=2}^{\infty} c_j^{(n)} H_j(\omega(i, x)) 1_{\{i \leftrightarrow x\}}.$$

Then, by gaussian property,

$$\begin{aligned}
 \mathbb{E}_{\mathbb{Q}}[\mathcal{R}_n^{\tilde{\omega}_n}]^2 &= 2 \sum_{i \in E_k^n, nt_0 \leq i \leq nt_1} \mathbb{E}_{\mathbb{Q}} \left[\sum_{x \in \mathbb{Z}, \sqrt{nx_0} \leq x \leq \sqrt{nx_1}} \sum_{j=2}^{\infty} c_j^{(n)} H_j(\omega(i, x)) 1_{\{i \leftrightarrow x\}} \right]^2 \\
 &= 2 \sum_{i \in E_k^n, nt_0 \leq i \leq nt_1} \sum_{j=2}^{\infty} (c_j^{(n)})^2 \mathbb{E}_{\mathbb{Q}} \left[\sum_{x \in \mathbb{Z}, \sqrt{nx_0} \leq x \leq \sqrt{nx_1}} H_j(\omega(i, x)) 1_{\{i \leftrightarrow x\}} \right]^2 \\
 &= \sum_{j=2}^{\infty} (c_j^{(n)})^2 j! M \left[N \gamma^j(0) + \sum_{l=1}^{N-1} (N-l) \gamma^j(l) \right]
 \end{aligned}$$

with $M = \lfloor n(t_1 - t_0) \rfloor$. The rhs of the above identity is bounded by


$$\begin{aligned} MN \sum_{j=2}^{\infty} (c_j^{(n)})^2 j! \left[\gamma^j(0) + \sum_{l=1}^{N-1} \gamma^j(l) \right] &\sim MN \sum_{j=2}^{\infty} (c_j^{(n)})^2 j! \left[\gamma^j(0) + \sum_{l=1}^{N-1} \gamma^j(l) \right] \\ &\leq CMN \sum_{j=2}^{\infty} (c_j^{(n)})^2 j! N^{1+j(1-2\alpha)}. \end{aligned}$$

Whence we have

$$n^{-(H+1)} \mathbb{E}_{\mathbb{Q}} [\mathcal{R}_n^{\tilde{\omega}_n}]^2 \sim C(t_1 - t_0)(x_1 - x_0)^2 \sum_{j=2}^{\infty} (c_j^{(n)})^2 j! n^{(j-1)(H-1)} (x_1 - x_0)^{2j(H-1)}$$

and

$$n^{-\frac{H+1}{2}} \mathcal{S}_1^n(h, \tilde{\omega}_n) \xrightarrow{D} I_1(h)$$

for $h = 1_{\{t_0 \leq t \leq t_1, x_0 \leq x \leq x_1\}}$ as $n \rightarrow \infty$ by Slutsky's theorem again. 

Thank You!