

FUNCTIONAL CONVERGENCE OF
BERRY'S NODAL LENGTHS

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- **Berry's planar random wave**, written

$$B_E = \{B_E(x) : x \in \mathbb{R}^2\}, \quad E > 0$$

is the unique planar centred, isotropic Gaussian field such that

$$\Delta B_E + 4\pi^2 E \cdot B_E = 0 \quad \text{a.s.} \quad \text{where} \quad \Delta = \frac{\partial^2}{\partial x_1^2} + \frac{\partial^2}{\partial x_2^2}.$$

- Equivalently,

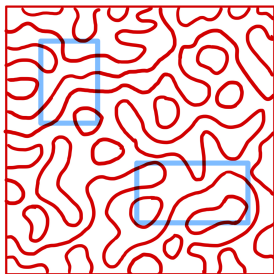
$$\mathbb{E}[B_E(x)B_E(y)] := J_0(2\pi\sqrt{E}\|x - y\|)$$

- Write $b = \{b(x) : x \in \mathbb{R}^2\}$ for $B_{(4\pi^2)^{-1}}$

- Think of b as a “canonical” Gaussian Laplace eigenfunction on \mathbb{R}^2 , emerging e.g. as a universal local scaling limit for arithmetic and monochromatic RWs, random spherical harmonics ...
See: *Marinucci & Rossi (2016), Canzani and Hanin (2021), Dierickx, Nourdin, Peccati & Rossi (2023)*.

NODAL LENGTHS

- ▶ $B_E^{-1}(0) := \{x \in \mathbb{R}^2 : B_E(x) = 0\}$ smooth curves
- ▶ $\mathcal{L}_E(Q) := \text{length}(B_E^{-1}(0) \cap Q)$ rectangle $Q \subset \mathbb{R}^2$



GOAL

To fix ideas, let \mathcal{Q} be the collection of all rectangles $Q \subset [0, 1]^2$.

- ▶ For every $E \geq 1$ and $Q \in \mathcal{Q}$, consider $\mathcal{L}_E(Q)$.
- ▶ **Task 1:** describe the **joint fluctuations**, as $E \rightarrow \infty$, of the random variables $\mathcal{L}_E(Q)$, $Q \in \mathcal{Q}$.
- ▶ **Task 2:** describe the **functional fluctuations**, as $E \rightarrow \infty$, of the random function

$$(s_1, s_2) \mapsto \mathcal{L}_E(s_1, s_2) := \mathcal{L}_E([0, s_1] \times [0, s_2]) \quad s_1, s_2 \in [0, 1]$$

- ▶ *Berry (J. Phys. A, 2002)* – as $E \rightarrow \infty$:

$$\mathbb{E}[\mathcal{L}_E(Q)] = \frac{\pi \text{area } Q}{\sqrt{2}} \sqrt{E} \quad \text{Var}(\mathcal{L}_E(Q)) \sim \frac{\text{area } Q}{512\pi} \log E$$

- ▶ Such an estimate follows from an analytical **cancellation** in Kac-Rice formulae: the natural guess for the order of the variance is \sqrt{E} .
- ▶ *Nourdin, Peccati & Rossi (CMP, 2019)*:

$$\sqrt{\frac{512\pi}{\log E}} (\mathcal{L}_E(Q) - \mathbb{E}\mathcal{L}_E(Q)) \xrightarrow{d} \mathcal{N}(0, \text{area}(Q)),$$

Define:

$$\widetilde{\mathcal{L}}_E(Q) := \sqrt{\frac{512\pi}{\log E}} \{ \mathcal{L}_E(Q) - \mathbb{E}[\mathcal{L}_E(Q)] \}, \quad t \geq 1,$$

and similarly $\left\{ \widetilde{\mathcal{L}}_E(s_1, s_2) := \mathcal{L}_E([0, s_1] \times [0, s_2]) : (s_1, s_2) \in [0, 1]^2 \right\}$.

1. For all $Q_1, \dots, Q_d \in \mathcal{Q}$, as $E \rightarrow \infty$, $\left(\widetilde{\mathcal{L}}_E(Q_1), \dots, \widetilde{\mathcal{L}}_E(Q_d) \right)$ converges to a centered Gaussian vector with covariance function

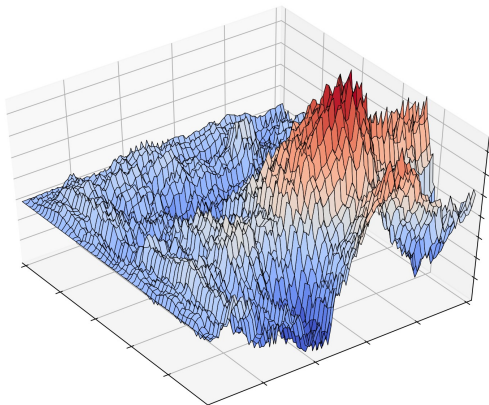
$$\Sigma(i, j) = \text{area}(Q_i \cap Q_j).$$

2. As $E \rightarrow \infty$, the random field $\left\{ \widetilde{\mathcal{L}}_E(s_1, s_2) : (s_1, s_2) \in [0, 1]^2 \right\}$ converges in the f.d.d.-sense to a standard **Wiener sheet**.

WIENER SHEET

A standard Wiener sheet $\{\mathbf{W}(s), s \in [0, 1]^2\}$ is a centred Gaussian process with covariance

$$\mathbb{E}[\mathbf{W}(s_1, s_2)\mathbf{W}(t_1, t_2)] = (s_1 \wedge t_1)(s_2 \wedge t_2)$$



A realization of a Wiener sheet, pic by George Lowther

Question: Does $\left\{ \widetilde{\mathcal{L}}_E(s_1, s_2) \right\}$ converge to a Wiener sheet *as a random function* (i.e. in \mathcal{D}_2 , the Skorohod space of cadlag mappings on $[0, 1]^2$)?

Lemma. $\{X, X_n : n \geq 1\} \subset \mathcal{D}_2$, $X_n = U_n + V_n + W_n$

(a) as $n \rightarrow \infty$, U_n converges weakly to X in \mathcal{D}_2 ,

(b) as $n \rightarrow \infty$, V_n converges weakly to zero in \mathcal{D}_2 ,

(c) for every $\varepsilon > 0$, $\lim_{n \rightarrow \infty} \mathbb{P} \left\{ \sup_{\mathbf{t} \in [0, 1]^2} |W_n(\mathbf{t})| > \varepsilon \right\} = 0$,

$\implies X_n$ converges weakly to X in \mathcal{D}_2 .

► $\widetilde{\mathcal{L}}_E[q] := \text{proj}(\widetilde{\mathcal{L}}_E|C_q)$, $C_q :=$ the q th Wiener chaos associated with b .

► Then,

$$\widetilde{\mathcal{L}}_E = \widetilde{\mathcal{L}}_E[2] + \widetilde{\mathcal{L}}_E[4] + R_E, \quad \text{where} \quad R_E := \sum_{q \geq 3} \widetilde{\mathcal{L}}_E[2q],$$

Strategy: applying the previous lemma to

$$(X_n, U_n, V_n, W_n) = (\widetilde{\mathcal{L}}_E, \widetilde{\mathcal{L}}_E[4], \widetilde{\mathcal{L}}_E[2], R_E)$$

- (I) $\widetilde{\mathcal{L}}_E[4]$ converges weakly to a standard Wiener sheet (OK – easy);
- (II) $\widetilde{\mathcal{L}}_E[2]$ converges weakly to zero – *total disorder* of $E^{1/4} \mathcal{L}_E[2]$;
- (III) the residual term R_E converges uniformly to zero in probability.

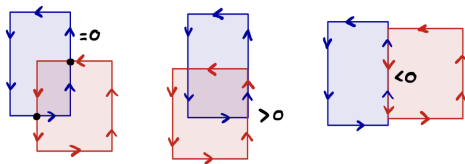
As $E \rightarrow \infty$,

$$\text{Cov}(\mathcal{L}_E[2](Q_1), \mathcal{L}_E[2](Q_2)) = \frac{\lambda(\partial Q_1, \partial Q_2)}{16\pi^2\sqrt{E}} + o\left(\frac{1}{\sqrt{E}}\right),$$

where

$$\lambda(\partial Q_1, \partial Q_2) = \int_{\partial Q_1 \cap \partial Q_2} \langle \mathbf{n}_1(x), \mathbf{n}_2(x) \rangle d\mathcal{H}^1(x).$$

indicates the **signed length** of $\partial Q_1 \cap \partial Q_2$.



- As $E \rightarrow \infty$,

$$4\pi E^{1/4} \left(\mathcal{L}_E[2](Q_1), \dots, \mathcal{L}_E[2](Q_d) \right) \xrightarrow{d} \mathcal{N}_d(0, \Sigma)$$

$$\Sigma(i, j) = \lambda(\partial Q_i, \partial Q_j)$$

- The centered Gaussian family

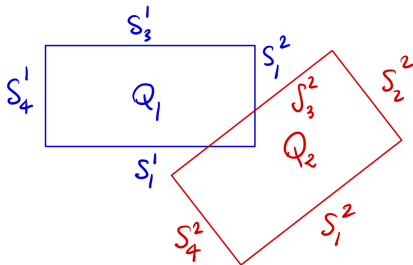
$$G = \{G(Q), Q \in \mathcal{Q}\} \quad \text{s.t.} \quad \mathbb{E}[G(Q_1)G(Q_2)] = \lambda(\partial Q_1, \partial Q_2)$$

is a **total disorder** random field: the linear span of G contains an uncountable collection of i.i.d. centered Gaussian random variables with unit variance.

STEP (II) – PROOF OF $\text{Cov}(\mathcal{L}_E[2](Q_1), \mathcal{L}_E[2](Q_2))$

► $\mathcal{L}_E(Q)[2] = \phi_E(\partial Q) = \frac{1}{8\pi\sqrt{2E}} \int_{\partial Q} B_E(x) \langle \nabla B_E(x), n_{\partial Q}(x) \rangle d\mathcal{H}^1(x)$

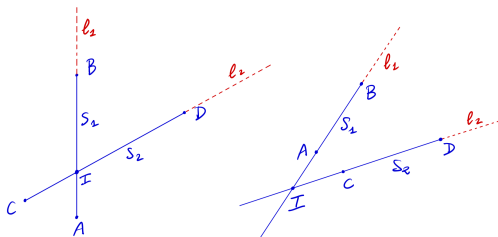
► $\text{Cov}(\mathcal{L}_E(Q_1)[2], \mathcal{L}_E(Q_2)[2]) = \text{Cov}\left(\sum_{k=1}^4 \phi_E(S_k^1), \sum_{k=1}^4 \phi_E(S_k^2)\right)$



► \implies we can reduce our study to **line segments**.

STEP (II) – PROOF

- $S_1 = [A, B]$ $S_2 = [C, D]$ $I = \ell_1 \cap \ell_2$



- $S_1 \cap S_2 = \{I\} \implies \text{Cov}(\phi_E(S_1), \phi_E(S_2)) =$
 $\text{Cov}(\phi_E([A, I]) + \phi_E([I, B]), \phi_E([C, I]) + \phi_E([I, D]))$
- $S_1 \cap S_2 = \emptyset \implies \text{Cov}(\phi_E(S_1), \phi_E(S_2)) =$
 $\text{Cov}(\phi_E([I, B]) - \phi_E([I, A]), \phi_E([I, D]) - \phi_E([I, C]))$
- **distributional invariance under rigid motion of $B_E \implies$ we consider line segments of the type $[(0, 0), P]$ (except when parallel and disjoint).**

STEP (II) – PROOF

- $\lambda_i := \text{length of } S_i$ e_i canonical basis vector of \mathbb{R}^2
 $\rho(\theta) = (\cos \theta, \sin \theta)$ $\theta \in [0, 2\pi)$

$$\gamma_1 : [0, \lambda_1] \longrightarrow S_1$$

$$t \longmapsto te_1$$

$$\gamma_2 : [0, \lambda_2] \longrightarrow S_2$$

$$t \longmapsto t\rho(\theta)$$

- $n_{S_1}(x) = e_2 \quad \forall x \in S_1$

$$n_{S_2}(x) = \rho(\theta)^\perp = (-\sin \theta, \cos \theta) \quad \forall x \in S_2$$

- $\text{Cov}(\phi_E(S_1), \phi_E(S_2)) = \frac{1}{64} \int_0^{\lambda_1} dt \int_0^{\lambda_2} ds$

$$\left[\psi_{2,2}^E(\gamma_1(t), \gamma_2(s)) \cos \theta - \psi_{2,1}^E(\gamma_1(t), \gamma_2(s)) \sin \theta \right]$$

$$\blacktriangleright \psi_{i,j}^E(x, y) := \mathbb{E} \left[B_E(x) B_E(y) \tilde{\partial}_i B_E(x) \tilde{\partial}_j B_E(y) \right], \quad i, j = 1, 2$$

$$\blacktriangleright \tau^E(t, s) := 2\pi\sqrt{E} \|\gamma_1(t) - \gamma_2(s)\| = 2\pi\sqrt{E} \sqrt{t^2 + s^2 - 2ts \cos \theta}$$

\blacktriangleright Feynmann's formula:

$$\begin{aligned} & \text{Cov}(\phi_E(S_1), \phi_E(S_2)) \\ &= \frac{2 \cos \theta}{64} \int_0^{\lambda_1} dt \int_0^{\lambda_2} ds \frac{J_0(\tau^E(t, s)) J_1(\tau^E(t, s))}{\tau^E(t, s)} \\ & \quad - \frac{2 \sin^2 \theta}{64} \int_0^{\lambda_1} dt \int_0^{\lambda_2} ds \frac{ts (J_0(\tau^E(t, s)) J_2(\tau^E(t, s)) + J_1(\tau^E(t, s)))}{t^2 + s^2 - 2ts \cos \theta} \\ &= A_E(\lambda_1, \lambda_2, \theta) + B_E(\lambda_1, \lambda_2, \theta) \end{aligned}$$

S_1 and S_2 are parallel:

► $\theta \in \{0, \pi\} \implies B_E(\lambda_1, \lambda_2, \theta) = 0$

► $0 \leq a < b, 0 \leq c < d$ and $L \geq 0$

$$\gamma_1 : t \in [a, b] \mapsto te_1$$

$$\gamma_2 : t \in [c, d] \mapsto te_1 + Le_2$$

► $L > 0 \iff S_1$ and S_2 are supported by parallel distinct lines.

► $L = 0 \iff S_1$ and S_2 are supported by the same line

► $\|\gamma_1(t) - \gamma_2(s)\|^2 = (t - s)^2 + L^2$

► change of variable:

$$\begin{aligned} \text{Cov}(\phi_E(S_1), \phi_E(S_2)) &= \frac{1}{32} \int_a^b du \frac{1}{2\pi\sqrt{E}} \int_{2\pi\sqrt{E}(u-d)}^{2\pi\sqrt{E}(u-c)} dv \\ &\quad \frac{J_0\left(\sqrt{v^2 + (2\pi^2\sqrt{E}L)^2}\right) J_1\left(\sqrt{v^2 + (2\pi\sqrt{E}L)^2}\right)}{\sqrt{v^2 + (2\pi\sqrt{E}L)^2}} \end{aligned}$$

$$\begin{aligned} K^E(u; L, c, d) &:= \frac{1}{2\pi\sqrt{E}} \int_{2\pi\sqrt{E}(u-d)}^{2\pi\sqrt{E}(u-c)} dv \\ &\quad \frac{J_0\left(\sqrt{v^2 + (2\pi^2\sqrt{E}L)^2}\right) J_1\left(\sqrt{v^2 + (2\pi\sqrt{E}L)^2}\right)}{\sqrt{v^2 + (2\pi\sqrt{E}L)^2}} \end{aligned}$$

$L > 0$:

- ▶ $|J_\nu(x)| = O(x^{-1/2})$ for $x > 0$ and $\nu = 0, 1, 2$
- ▶ \implies uniformly on u

$$\begin{aligned} \sqrt{E} \left| K^E(u; L, c, d) \right| &\leq \frac{O(1)}{2\pi} \int_{2\pi\sqrt{E}(u-d)}^{2\pi\sqrt{E}(u-c)} dv \frac{1}{v^2 + (2\pi\sqrt{E}L)^2} \frac{1}{E} \\ &= O(E^{-1/2}) \end{aligned}$$

- ▶ $\sqrt{E}K^E(u; L, c, d) \rightarrow 0$ uniformly on (a, b) and

$$\text{Cov}(\phi_E(S_1), \phi_E(S_2)) = O\left(\frac{1}{E}\right) \quad E \rightarrow \infty$$

$L = 0$:

$$\blacktriangleright K^E(u; 0, c, d) = \frac{1}{2\pi\sqrt{E}} \int_{2\pi\sqrt{E}(u-d)}^{2\pi\sqrt{E}(u-c)} dv \frac{J_0(|v|)J_1(|v|)}{|v|}$$

$\blacktriangleright [a, b] \cap [c, d] = \emptyset \implies \text{Cov}(\phi_E(S_1), \phi_E(S_2)) = O(E^{-1})$ (as before)

$\blacktriangleright [a, b] \cap [c, d] \neq \emptyset$: we start with the case $(a, b) = (c, d)$.

$\blacktriangleright \sqrt{E} \text{Cov}(\phi(S_1), \phi(S_2)) = \frac{1}{32} \int_a^b du \sqrt{E} K^E(u; 0, a, b)$ with

$$\sqrt{E} K^E(u; 0, a, b) = \frac{1}{2\pi} \int_{2\pi\sqrt{E}(u-b)}^{2\pi\sqrt{E}(u-a)} dv \frac{J_0(v)J_1(v)}{v}$$

since J_0 is even and J_1 is odd.

STEP (II) – PROOF

$$\blacktriangleright \frac{d}{dv} [v(J_0(v)^2 + J_1(v)^2) - J_0(v)J_1(v)] = J_0(v)J_1(v)/v$$

$$\begin{aligned} \blacktriangleright \sqrt{E} \operatorname{Cov}(\phi(S_1), \phi(S_2)) &= \underbrace{\frac{1}{64\pi} \int_a^b du [v(J_0(v)^2 + J_1(v)^2)]}_{(\star)} \frac{2\pi\sqrt{E}(u-a)}{2\pi\sqrt{E}(u-b)} \\ &\quad - \underbrace{\frac{1}{64\pi} \int_a^b du [J_0(v)J_1(v)]}_{o(1)} \frac{2\pi\sqrt{E}(u-a)}{2\pi\sqrt{E}(u-b)}. \end{aligned}$$

$$\blacktriangleright f(v) := v(J_0(v)^2 + J_1(v)^2) \implies f(-v) = -f(v)$$

$$\blacktriangleright \lim_{E \rightarrow \infty} \frac{[f(v)]^{2\pi\sqrt{E}(u-a)}}{2\pi\sqrt{E}(u-b)} = \lim_{y \rightarrow \infty} 2f(y) = \frac{4}{\pi}$$

$$\blacktriangleright \text{dominated convergence: } (\star) \longrightarrow \frac{1}{64\pi} \int_a^b \frac{4}{\pi} du = \frac{1}{16\pi^2} (b-a)$$

- ▶ we just proved that: $\text{Cov}(\phi(S_1), \phi(S_2)) = \frac{1}{16\pi^2} \frac{b-a}{\sqrt{E}} + o\left(\frac{1}{\sqrt{E}}\right)$
- ▶ $S_1 \neq S_2$ but $S_1 \cap S_2 \neq \emptyset$:
- ▶ w.l.o.g. $0 < a < c \leq b < d \implies S_1 \cap S_2 = [c, b] \times \{0\}$
- ▶ $\phi_E(S_1) = \phi_E([a, c] \times \{0\}) + \phi_E([c, b] \times \{0\})$
 $\phi_E(S_2) = \phi_E([c, b] \times \{0\}) + \phi_E([b, d] \times \{0\})$
- ▶ $\text{Cov}(\phi_E(S_1), \phi_E(S_2))$
 $= \text{Cov}(\phi_E([c, b] \times \{0\}), \phi_E([c, b] \times \{0\})) + o\left(\frac{1}{\sqrt{E}}\right)$
 $= \frac{\lambda(S_1, S_2)}{16\pi^2\sqrt{E}} + o\left(\frac{1}{\sqrt{E}}\right)$

S_1 and S_2 are not parallel:

$$\blacktriangleright \text{Cov}(\phi_E(S_1), \phi_E(S_2)) = A_E(\lambda_1, \lambda_2, \theta) + B_E(\lambda_1, \lambda_2, \theta)$$

$$\blacktriangleright A_E(\lambda_1, \lambda_2, \theta) = O\left(\frac{1}{E}\right)$$

$$\blacktriangleright B_E(\lambda_1, \lambda_2, \theta) = O\left(\frac{\log E}{E}\right)$$

$$\blacktriangleright \implies \text{Cov}(\phi_E(S_1), \phi_E(S_2)) = o\left(\frac{1}{\sqrt{E}}\right)$$

- **Gaussian entire function:** $\{\zeta_n\}$ i.i.d. complex std Gaussian

$$z \mapsto f(z) = \sum_{n=0}^{\infty} \zeta_n \frac{z^n}{n!} \quad z \in \mathbb{C}$$

- $f_R(z) := f(Rz) \quad n_R(Q) := \# \{f_R^{-1}(0) \cap Q\}$

- $\text{Var}(n_R(Q)) = c_0 R \cdot \text{length}(\partial Q) + o(R) \quad R \rightarrow \infty$

- $\frac{1}{\sqrt{c_0 R}} ((n_R(Q_1) - \mathbb{E}[n_R(Q_1)]), \dots, (n_R(Q_d) - \mathbb{E}[n_R(Q_d)])) \xrightarrow{d} \mathcal{N}_d(0, \Sigma)$

- ▶ *hyperuniformity*: a variance that scales as the length of the boundary of $R \cdot Q$, rather than as $\text{area}(R \cdot Q) \asymp R^2$

$$\text{Var}(\mathcal{L}(b; R \cdot Q)[2]) = \frac{R \cdot \text{length}(\partial Q)}{8\pi} + o(R) \quad R = 2\pi\sqrt{E}$$

$$\text{Var}(n_R(Q)) = c_0 R \cdot \text{length}(\partial Q) + o(R) \quad R \rightarrow \infty$$

- ▶ *total disorder process*:
 - ▶ $\mathcal{L}_E(Q)[2]$ and $n_R(Q)$ exhibit the same limit in the f.d.d.-sense, which is a total disorder field.
 - ▶ physics, random matrix theory.
- ▶ As $E \rightarrow \infty$, the field $\{\widetilde{\mathcal{L}}_E(\mathbf{s})[2] : \mathbf{s} \in [0, 1]^2\}$ weakly converges to zero in \mathcal{D}_2 (tightness + estimates for sup of stationary Gaussian fields).

Fix $K \geq 1$, consider the partition Π_K of $[0, 1]^2$ formed by the collection of squares of side length 2^{-K} :

- For every $i = (i_1, i_2) \in \{0, \dots, 2^K\}^2$, we define the *partition points* $\mathbf{p}_i(K, K) := (p_{i_1}(K), p_{i_2}(K)) \in [0, 1]^2$ by

$$p_{i_1}(K) := \frac{i_1}{2^K}, \quad p_{i_2}(K) := \frac{i_2}{2^K}, \quad i_1, i_2 = 0, 1, \dots, 2^K.$$

- For $\mathbf{s} = (s_1, s_2) \in [0, 1]^2$, we write $i_{K,K}(\mathbf{s}) = (i_{1,K}(s_1), i_{2,K}(s_2))$ for the vector verifying

$$p_{i_{1,K}(s_1)} \leq s_1 < p_{i_{1,K}(s_1)+1} \quad p_{i_{2,K}(s_2)} \leq s_2 < p_{i_{2,K}(s_2)+1}$$

that is, the vector $i_{K,K}(\mathbf{s})$ is such that $\mathbf{p}_{i_{K,K}(\mathbf{s})}(K, K)$ is the closest partition point to \mathbf{s} on the left.

Discretized nodal length:

$$\mathcal{L}_E^K(s_1, s_2) := \mathcal{L}_E \left([0, p_{i_1, K}(s_1)(K)] \times [0, p_{i_2, K}(s_2)(K)] \right)$$

Take $\{K(E) : E > 0\}$ numerical sequence s.t. $K(E) \rightarrow \infty$ and $K(E) = o((\log E)^{1/10})$ as $E \rightarrow \infty$. Then,

1. for every $\varepsilon > 0$, $\mathbb{P} \left\{ \sup_{\mathbf{s} \in [0,1]^2} |R_E^{K(E)}(\mathbf{s})| > \varepsilon \right\} \rightarrow 0$
2. the normalized process $\left\{ \widetilde{\mathcal{L}}_E^{K(E)}(\mathbf{s}) \right\}$ converges weakly to a standard Wiener sheet \mathbf{W} on $[0, 1]^2$ in the Skorohod space \mathcal{D}_2

Main difficulty for directly dealing with the residual term R_E :

- ▶ The **expectation** of $\mathcal{L}_E(\mathbf{s})$ (order \sqrt{E}) grows much faster than the normalizing factor $\log E$.
- ▶ Planar chaining argument with R_E requires

$$|\mathbb{E}[\mathcal{L}_E(\mathbf{t})] - \mathbb{E}[\mathcal{L}_E(\mathbf{p}_{i_{K,K}}(\mathbf{t})(K, K))]| \approx \frac{\sqrt{E}}{\sqrt{\log E}} \frac{1}{2^K}$$

to be bounded.

- ▶ This requirement is **incompatible** with the constrained choice $K(E) = o((\log E)^{1/10})$, as is needed in the above statements.

Thank you!