

Peaks of the parabolic Anderson model with white noise potential

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Review

Consider the parabolic Anderson model (PAM) in $d \leq 3$:

$$\begin{cases} \partial_t u(t, x) = \Delta u(t, x) + \xi(x) u(t, x), & (t, x) \in \mathbf{R}_+ \times \mathbf{R}^d, \\ u(0, x) \equiv 1 & x \in \mathbf{R}^d, \end{cases}$$

where ξ is white noise potential, i.e. mean-zero Gaussian process with covariance $\mathbf{E}[\xi(x)\xi(y)] = \delta_0(x - y)$.

In particular, we are interested in

- ▶ **Spatial asymptotics** as $|x| \rightarrow \infty$
- ▶ **Macroscopic Hausdorff dimension** of tall peaks

Main results

For $d \leq 3$, define the random set of peaks

$$\mathcal{P}_t^d(\alpha) := \left\{ x \in \mathbf{R}^d : u(t, x) \geq \exp \left(\alpha t (\log |x|)^{2/(4-d)} \right) \right\},$$

for $\alpha, t > 0$.

Theorem (Ghosal-Y.'26 (AOP))

There exist $\mathfrak{c}_d > 0$ and $t_0 = t_0(\alpha, d) > 0$ such that for all $t \geq t_0$

$$\text{Dim}_H[\mathcal{P}_t^d(\alpha)] = (d - \alpha^{(4-d)/2} \mathfrak{c}_d) \vee 0, \quad a.s.$$

where Dim_H denotes the macroscopic Hausdorff dimension.

Moreover,

$$\limsup_{|x| \rightarrow \infty} \frac{\log u(t, x)}{(\log |x|)^{2/(4-d)}} = \left(\frac{d}{\mathfrak{c}_d} \right)^{\frac{2}{4-d}} t, \quad a.s.$$

We say that peaks of the PAM are *macroscopically multifractal*.

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Proof outline for the lower bound

- ▶ We show that $\mathcal{P}_t^d(\alpha)$ contains γ -thick set for all $\gamma > \mathfrak{c}_d \alpha^{(4-d)/2} / d$, which proves

$$\dim_{\mathrm{H}}[\mathcal{P}_t^d(\alpha)] \geq d(1 - \gamma) \rightarrow d - \alpha^{(4-d)/2} \mathfrak{c}_d.$$

- ▶ Indeed, for each n , we can find points $\{x_i\}_{i=1}^{m(n)}$ satisfying
 - ▶ $x_i \in B(x, e^{n\gamma})$ for $x \in (e^n, e^{n+1}]^d \cap e^{n\gamma} \mathbf{Z}$ and $i = 1, \dots, m(n)$,
 - ▶ $|x_i - x_j| \geq e^{n\theta}$ whenever $1 \leq i < j \leq m(n)$,
 - ▶ $m(n) \approx_d e^{nd(\gamma - \theta)}$.
- ▶ Then we show that

$$\sum_{n=0}^{\infty} \mathbf{P} \left(\min_{x \in (e^n, e^{n+1}]^d \cap e^{n\gamma} \mathbf{Z}} \max_{1 \leq i \leq m(n)} \frac{\log u(t, x_i)}{(\log |x_i|)^{2/(4-d)}} \leq \alpha t \right) < \infty.$$

Letting $\gamma \rightarrow \mathfrak{c}_d \alpha^{(4-d)/2} / d$ as $\theta \rightarrow 0$, the conclusion follows from lower bound of the tail estimate and the Borel-Cantelli lemma.

Proof outline for the upper bound

- We prove that for all $\rho \in (d - \mathfrak{c}_d \alpha^{(4-d)/2}, d)$,

$$\sum_{n=0}^{\infty} v_{\rho}^n(\mathcal{P}_t^d(\alpha)) < \infty \quad \text{a.s.},$$

which concludes the upper bound by taking $\rho \rightarrow d - \mathfrak{c}_d \alpha^{(4-d)/2}$.

- Recalling the definition of v_{ρ}^n , we show that

$$\sum_{n=0}^{\infty} e^{-n\rho} \sum_{\substack{y \in \mathbb{Z}^d \\ B(y,1) \subseteq \mathbb{S}_n}} \mathbf{P} \left(\sup_{x \in B(y,1)} \frac{\log u(t,x)}{(\log |x|)^{2/(4-d)}} \geq \alpha t \right) < \infty,$$

which implies $\mathbf{E}[\sum_{n=0}^{\infty} v_{\rho}^n(\mathcal{P}_t^d(\alpha))] < \infty$.

Proof outline

We first focus on the tail estimate $\mathbf{P}(u(t, x) > z)$.

Step 1. Approximate $u(t, x)$ by $u_L(t, x)$ where $L = L(t) = t^a$ using Feynman-Kac representation.

Step 2. Approximate $u_L(t, x)$ by $\exp(\lambda_L t)$ where λ_L is the principal eigenvalue of the Anderson hamiltonian on Q_L . Thus, we have

$$\mathbf{P}(u(t, x) > z) \approx \mathbf{P}(u_L(t, x) > z) \approx \mathbf{P}\left(\lambda_L > \frac{\log z}{t}\right).$$

Step 3. Apply upper and lower bounds for tail probability $\mathbf{P}(\lambda_L > \log z/t)$.

Tail probability of the eigenvalue

[Chouk-Zuijlen'21], [Hsu-Labbé'22]: Tail probability for the principal eigenvalue.

Theorem

Fix $\varepsilon \in (0, 1)$. There exist $c_2 > c_1 > 0$ and s_0 such that for all $L \geq 1$ and $s \geq s_0$

$$\mathbf{P}(\lambda_L \leq s) \leq \exp\left(-c_2 s^{d/2} e^{d \log L - (1+\varepsilon) c_d s^{(4-d)/2}}\right)$$

$$\mathbf{P}(\lambda_L \geq s) \leq c_1 s^{d/2} \exp\left(d \log L - (1-\varepsilon) c_d s^{(4-d)/2}\right).$$

Proof idea: Large deviation for the eigenvalue of the Anderson Hamiltonian. The constant c_d and the exponent $(4-d)/2$ comes from the corresponding variational problem.

Monotonicity of eigenvalues

For each $x \in Q_L$, we know that $\mathbf{P}(u(t, x) > z) \approx \mathbf{P}(\lambda_L > \log z / t)$. Then how do we derive $\mathbf{P}(\sup_{x \in B(y, 1)} u(t, x) > z)$?

Let $\lambda_L(x)$ be the principal eigenvalue on $Q_L^x = [x - L/2, x + L/2]$. For $L \geq r \geq 1$ and $x, y \in \mathbf{R}^d$ such that $Q_r^y \subseteq Q_L^x$

$$\lambda_L(y) \leq \lambda_L(x) .$$

Therefore, for $L \geq 1$

$$\mathbf{P}\left(\sup_{x \in B(y, 1)} u(t, x) > z\right) \leq \mathbf{P}\left(\lambda_L(y) \geq \frac{\log z}{t}\right) .$$

Independence of eigenvalues

For the lower bound, we need to bound

$$\mathbf{P}\left(\max_{1 \leq i \leq m} u(t, x_i) \leq z\right) \approx \mathbf{P}\left(\max_{1 \leq i \leq m} \lambda_L(x_i) \leq \frac{\log z}{t}\right),$$

for $|x_i - x_j| \geq e^{n\theta}$ with $\theta > 0$.

Note that if $|x_i - x_j| \geq 2L$, $\lambda_L(x_i)$ and $\lambda_L(x_j)$ are independent since the noise ξ restricted on $Q_L^{x_i}$ and $Q_L^{x_j}$ are independent. Therefore,

$$\mathbf{P}\left(\max_{1 \leq i \leq m} \lambda_L(x_i) \leq \frac{\log z}{t}\right) = \prod_{i=1}^m \mathbf{P}\left(\lambda_L(x_i) \leq \frac{\log z}{t}\right).$$

Step 1

By the Feynman-Kac formula, we have

$$u(t, x) - u_L(t, x) = \mathbf{E}_x \left[e^{\int_0^t (\xi_\varepsilon(B_s) - C_\varepsilon) \, ds} \mathbb{1}_{B[0, t] \not\subset Q_L} \right] .$$

We can easily get

$$\mathbf{P}(B[0, t] \not\subset Q_L) \approx \mathbf{P} \left(\sup_{s \in [0, t]} |B_t - x| > L \right) \lesssim e^{-cL^2/t} .$$

However, $\xi_\varepsilon - C_\varepsilon$ is not bounded in L^∞ as $\varepsilon \rightarrow 0$.

We need a change of measure or *Girsanov theorem*.

Step 1

Consider a semilinear elliptic equation for Y with a large $\eta > 0$

$$(\eta - \frac{1}{2}\Delta)Y = \frac{1}{2}|\nabla X|^2 - C + \nabla Y \cdot \nabla X + \frac{1}{2}|\nabla Y|^2.$$

Then since $X - \frac{1}{2}\Delta X = \xi$ (recall $X = (1 - \frac{1}{2}\Delta)^{-1}\xi$), we have

$$\xi - C = X + \eta Y - \frac{1}{2}\Delta(X + Y) - \frac{1}{2}|\nabla(X + Y)|^2,$$

or equivalently,

$$\frac{1}{2}\Delta(X + Y) = -(\xi - C) + X + \eta Y - \frac{1}{2}|\nabla(X + Y)|^2.$$

On the other hand, by Itô's formula, we obtain

$$(X + Y)(B_t) = (X + Y)(B_0) + \frac{1}{2} \int_0^t \Delta(X + Y)(B_s) \, ds + \int_0^t \nabla(X + Y)(B_s) \, dB_s.$$

Step 1

Therefore, we have

$$\exp\left(\int_0^t (\xi(B_s) - C) \, ds\right) = \mathcal{N}_t \cdot \mathcal{D}_{0,t},$$

where

$$\mathcal{N}_t = \exp\left(\int_0^t \nabla(X + Y)(B_s) \, dB_s - \frac{1}{2} \int_0^t |\nabla(X + Y)(B_s)|^2 \, ds\right)$$

and

$$\mathcal{D}_{s,t} = \exp\left(\int_s^t (X + \eta Y)(B_r) \, dr + (X + Y)(B_s) - (X + Y)(B_t)\right).$$

Note that $X \in \mathcal{C}^{2-d/2-\kappa}$ and $Y \in \mathcal{C}^{3-d/2-\kappa}$. In particular, they are bounded and continuous if $d \leq 3 \implies \|\mathcal{D}_{0,t}\|_{L^\infty} < \infty$ for all $t > 0$.

Step 1

By the Girsanov theorem,

$$u(t, x) - u_L(t, x) = \mathbf{E}_x \left[e^{\int_0^t (\xi(B_s) - C) ds} \mathbb{1}_{B[0, t] \not\subseteq Q_L} \right] = \bar{\mathbf{E}}_x[\mathcal{D}_{0, t} \mathbb{1}_{Z[0, t] \not\subseteq Q_L}]$$

where $\bar{\mathbf{E}}_x$ is taken w.r.t $\bar{\mathbf{P}}_x$ under which the coordinate process Z_t satisfies

$$Z_t = x + \int_0^t \nabla(X + Y)(Z_s) ds + B'_s,$$

for a Brownian motion B' under $\bar{\mathbf{P}}_x$.

To estimate $\bar{\mathbf{P}}_x(Z[0, t] \not\subseteq Q_L)$, we will obtain the **transition density** Γ_t of Z_t . Γ is the **heat kernel** of the Cauchy problem

$$\partial_t w = \frac{1}{2} \Delta w + \mu \cdot \nabla w, \quad w(T, \cdot) = \delta_y,$$

where $\mu = \nabla(X + Y) \in \mathcal{C}^{1-d/2-\kappa}$ is distribution valued if $d \geq 2$.

Step 1

Note that for $d = 3$, $\mu = \nabla(X + Y) \in \mathcal{C}^{-1/2-\kappa}$, then the solution w to

$$\partial_t w = \frac{1}{2} \Delta w + \mu \cdot \nabla w, \quad w(T, \cdot) = \delta_y$$

belong to at most $w \in \mathcal{C}^{3/2-\kappa}$. Thus, $\nabla w \in \mathcal{C}^{1/2-\kappa}$, which implies that $\mu \cdot \nabla w$ is not well-defined since $(-1/2 - \kappa) + (1/2 - \kappa) < 0$.

Therefore, the above equation is also singular. We introduce a pair (w, w') (*paracontrolled distribution*) such that

$$w^\sharp(t) := w(t) - \nabla w(t) < \mathcal{I}(\mu)(t) \in \mathcal{C}^{2-2\kappa}.$$

Here, $<$ is a praproduct and $\mathcal{I}(\mu)(t) := \int_t^T P_{s-t} \mu(s) \, ds$.

We then solve a fixed point problem for $(w, \nabla w)$.

Step 1

By the relation between transition semigroup and the generator, for any $\phi \in C_c(\mathbf{R}^d)$ we have

$$\bar{\mathbf{E}}_x[\phi(Z_t)] = \int_{\mathbf{R}^d} \phi(y) \Gamma_t(x, y) \, dy,$$

where the transition density is defined by

$$\Gamma_t(x, y) = w(t, x).$$

The heat kernel estimate for Γ follows from the classical theory by Strook.

Step 1: Strook's theory

Let $\mu = \nabla U$ where $U \in L^\infty(\mathbf{R}^d)$ ($U = X + Y$ in our case). Then, there exist $C_1, C_2, C_3, C_4 > 0$ depending on d such that

$$\Gamma_t(x, y) \leq \frac{1}{t^{d/2}} \exp \left(C_1 \|U\|_{L^\infty} - C_2 \frac{|y-x|^2}{t} \right),$$
$$\Gamma_t(x, y) \geq \frac{1}{t^{d/2}} \exp \left(-C_3 \|U\|_{L^\infty} - C_4 \|U\|_{L^\infty} \frac{|x-y|^2}{t} \right).$$

Step 1

Let us summarise what we have estimated: Recall $X = (1 - 2^{-1} \Delta)^{-1} \xi$.

$$\|X\|_{\mathcal{C}^{1/2-\kappa}} + \|Y\|_{\mathcal{C}^{1-2\kappa}} \lesssim \mathfrak{a}(\log L)^2,$$

where $\mathfrak{a}$ is a random finite constant, which implies that

$$\exp(-\mathfrak{a}^{N_1}(t-s)(\log L)^{N_2}) \mathbb{1}_{Z[s,t] \subseteq Q_L} \leq \mathcal{D}_{s,t} \mathbb{1}_{Z[s,t] \subseteq Q_L} \leq \exp(\mathfrak{a}^{N_1}(t-s)(\log L)^{N_2}).$$

By Strook's theory, we have

$$t^{-d/2} \exp(-\mathfrak{a}(\log L)^2(1+t^{-1}|x-y|^2)) \leq \Gamma_t(x, y) \leq p_t(x, y) \exp(\mathfrak{a}(\log L)^2),$$

where p_t is transition density of the Brownian motion. As a corollary, for $r \leq L$ and $|x| < r/4$

$$\bar{\mathbb{P}}_x(Z[0, t] \not\subseteq Q_r) \lesssim \exp\left(C\mathfrak{a}(\log L)^2 t - \frac{r^2}{Ct}\right).$$

Step 1

Recall that

$$u(t, x) - u_L(t, x) = \mathbf{E}_x \left[e^{\int_0^t (\xi(B_s) - C) \, ds} \mathbb{1}_{B[0, t] \not\subseteq Q_L} \right] = \bar{\mathbf{E}}_x[\mathcal{D}_{0, t} \mathbb{1}_{Z[0, t] \not\subseteq Q_L}] .$$

Taking $L = r = t^2$, for sufficiently large t , we can make

$$u(t, x) - u_L(t, x) \lesssim \exp \left(\mathfrak{a}^{N_1} (2 \log t)^{N_2} t + \mathfrak{a} (2 \log t)^2 t - \frac{t^4}{Ct} \right) ,$$

arbitrarily small. In other words, we have approximated

$$u(t, x) \approx u_L(t, x) .$$

Step 2: Upper bound

Recall the spectral representation with an initial profile $\psi \in L^2(Q_L)$

$$u_L^\psi(t, x) = \sum_{k \in \mathbf{Z}_+} \exp(\lambda_L^{(k)} t) \langle \varphi_L^{(k)}, \psi \rangle \varphi_L^{(k)}(x).$$

Therefore,

$$\int_{Q_L} u_L^\psi(t, y) \, dy = \sum_{k \in \mathbf{Z}_+} \exp(\lambda_L^{(k)} t) \langle \varphi_L^{(k)}, \psi \rangle \langle \varphi_L^{(k)}, \mathbb{1}_{Q_L} \rangle \leq \exp(\lambda_L^{(1)} t) \|\psi\|_{L^2} \|\mathbb{1}_{Q_L}\|_{L^2}$$

Since $\delta_y \notin L^2$, we cannot directly use

$$u_L(t, x) = \int_{Q_L} u_L^{\delta_x}(t, y) \, dy.$$

However, by the Chapman-Kolmogorov equation with $\psi = u_L^{\delta_x}(t_0, \cdot)$,

$$u_L(t, x) \leq \exp(\lambda_L^{(1)} t) \|u_L^{\delta_x}(t_0, \cdot)\|_{L^2} \|\mathbb{1}_{Q_L}\|_{L^2} \lesssim \exp(\lambda_L^{(1)} t).$$

Step 2 : Lower bound

Now we estimate the lower bound. Recall that we have

$$u_L(t, x) = \bar{E}_x \left[\mathcal{D}_{0,t} \mathbb{1}_{Z[0,t] \subseteq Q_L} \right], \quad x \in Q_L.$$

Split into time intervals $[0, \delta]$ and $[\delta, t]$.

$$u_L(t, x) = \bar{E}_x \left[\mathcal{D}_{0,\delta} \mathbb{1}_{Z[0,\delta] \subseteq Q_L} \cdot \mathcal{D}_{\delta,t} \mathbb{1}_{Z[\delta,t] \subseteq Q_L} \right]$$

Using the Markov property at time δ , we have

$$u_L(t, x) \geq \exp(-C\alpha^{N_1} (\log L)^{N_2} \delta) \bar{E}_x \left[\mathbb{1}_{Z[0,\delta] \subseteq Q_L} u_L(t - \delta, Z_\delta) \right].$$

For any $r < L$, we can estimate as

$$\bar{E}_x \left[\mathbb{1}_{Z[0,\delta] \subseteq Q_L} u_L(t - \delta, Z_\delta) \right] \gtrsim \bar{E}_x \left[\mathbb{1}_{Z_\delta \in Q_r} u_r(t - \delta, Z_\delta) \right],$$

since $\{Z_\delta \in Q_r\} \subseteq \{Z[0,\delta] \subseteq Q_L\} \cup \{Z[0,\delta] \not\subseteq Q_L, Z_\delta \in Q_r\}$.

Step 2: Lower bound

Now it remain to bound

$$u_L(t, x) \gtrsim \bar{\mathbf{E}}_x \left[\mathbb{1}_{Z_\delta \in Q_r} u_r(t - \delta, Z_\delta) \right] .$$

Using the transition density,

$$\bar{\mathbf{E}}_x \left[\mathbb{1}_{Z_\delta \in Q_r} u_r(t - \delta, Z_\delta) \right] = \int_{Q_r} \Gamma_\delta(x, z) u_r(t - \delta, z) \, dz \gtrsim \int_{Q_r} u_r(t - \delta, z) \, dz .$$

By the Chapman-Kolmogorov equation and the Markov property,

$$u_r^{\delta_z}(t, z) \lesssim u_r(t - \delta, z) \cdot \mathcal{D}_{0, \delta} \mathbb{1}_{Z[0, \delta] \in Q_r} \lesssim u_r(t - \delta, z) \cdot \exp \left(C \alpha^{N_1} (\log r)^{N_2} \delta \right) .$$

Therefore, we have

$$\int_{Q_r} u_r(t - \delta, z) \, dz \gtrsim \int_{Q_r} u_r^{\delta_z}(t, z) \, dz = \sum_{k \in \mathbb{Z}_+} e^{\lambda_r^{(k)} t} \int_{Q_r} (\varphi_r^{(k)}(z))^2 \, dz \geq e^{\lambda_r^{(1)} t} .$$

Summary

$$\begin{cases} \partial_t u(t, x) = \Delta u(t, x) + \xi(x) u(t, x), & (t, x) \in \mathbf{R}_+ \times \mathbf{R}^d, \\ u(0, x) \equiv 1 & x \in \mathbf{R}^d. \end{cases}$$

- ▶ We showed spatial asymptotics and macroscopic multifractality of tall peaks of the solution to the PAM for $d \leq 3$.
- ▶ We employed
 - ▶ Paracontrolled distribution for singular SPDE theory.
 - ▶ Feynman-Kac representation.
 - ▶ Spectral representation.
 - ▶ Heat kernel estimates, Girsanov's theorem, etc.

Further questions

Open questions:

- ▶ Limit theorems for spatial integrals for the PAM in 2D and 3D.
- ▶ Concentration: Total mass of the solution is localised on small regions / several points.
- ▶ Solution theory and properties for the 4D PAM. The 4D is the critical dimension since there are infinitely many renormalisations.

Thank you

Thank you!