

Weierstrass Institute for Applied Analysis and Stochastics



# Eigenvalue order statistics and mass concentration in the parabolic Anderson model

Joint work in progress with Marek Biskup (České Budějovice and Los Angeles)

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Cauchy problem for the heat equation with random coefficients and localised initial datum:

$$\frac{\partial}{\partial t}u(t,z) = \Delta u(t,z) + \xi(z)u(t,z), \quad \text{for } (t,z) \in (0,\infty) \times \mathbb{Z}^d, \quad (1)$$

$$u(0,z) = \mathbb{1}_0(z), \quad \text{for } z \in \mathbb{Z}^d. \quad (2)$$



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$$\xi = (\xi(z): z \in \mathbb{Z}^d)$$
 i.i.d. random potential,  $[-\infty, \infty)$ -valued.  
•  $\Delta f(z) = \sum_{y \sim z} [f(y) - f(z)]$  discrete Laplacian  
•  $\Delta + \xi$  Anderson Hamiltonian



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#### Interpretations / Motivations:

- Random mass transport through a random field of sinks and sources.
- Expected particle number in a branching random walk model in a field of random branching and killing rates.
- Anderson Hamiltonian  $\Delta + \xi$  describes conductance properties of alloys of metals, or optical properties of glasses with impurities. Many open questions about delocalised versus extended states.





# Feynman-Kac formula

$$u(t,z) = \mathbb{E}_0 \Big[ \exp \Big\{ \int_0^t \xi(X(s)) \,\mathrm{d}s \Big\} \mathbb{1} \{ X(t) = z \} \Big], \qquad z \in \mathbb{Z}^d, t > 0,$$

where  $(X(s))_{s \in [0,\infty)}$  is the simple random walk on  $\mathbb{Z}^d$  with generator  $\Delta$ , starting from z under  $\mathbb{P}_z$ .



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# **Eigenvalue expansion**

$$u(t,z) \approx \mathbb{E}_0 \Big[ \exp \Big\{ \int_0^t \xi(X(s)) \, \mathrm{d}s \Big\} \mathbb{1} \{ X(t) = z \} \mathbb{1} \{ X_{[0,t]} \subset B^{(2)}(t) \} \Big]$$
$$= \sum_k \mathrm{e}^{t\lambda_k(\xi, B^{(2)}(t))} \varphi_k(0) \varphi_k(z),$$

where  $(\lambda_k(\xi, B^{(2)}(t)), \varphi_k)_k$  is a sequence of eigenvalues  $\lambda_1 > \lambda_2 \ge \lambda_3 \ge \ldots$  and  $L^2$ -orthonormal eigenfunctions  $\varphi_1, \varphi_2, \varphi_3 \ldots$  of  $\Delta + \xi$  in some box  $B^{(2)}(t) = t \log^2 t \times [-1, 1]^d$  with zero boundary condition.



# Main objectives

- Mass concentration: The total mass U(t) = ∑<sub>z∈Z<sup>d</sup></sub> u(t, z) comes in probability from just one island (strong form of intermittency).
- Eigenvalue order statistics: The top eigenvalues and the concentration centres of the corresponding eigenfunctions (after rescaling and shifting) form a Poisson point process.



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# Explanation

- The top eigenvalues satisfy an order statistics in some box  $B^{(1)}(t) \subset B^{(2)}(t)$ . In particular, we have control on their differences, i.e., the spectral gaps.
- The corresponding eigenfunctions are exponentially localised in islands  $B_{r_t}(z_k)$  whose locations  $z_k$  form a Poisson point process.
- The main contribution to U(t) inside  $B_t^{(1)}$  comes from precisely that summand k which maximises  $e^{t\lambda_k(\xi, B_t^{(1)})} |\varphi_k(0)|$ .
- The contribution to U(t) from the outside of the *a priori* box  $B^{(2)}(t)$  is negligible.
- The contribution to U(t) from  $B^{(2)}(t) \setminus B^{(1)}(t)$  is negligible since the values of  $e^{t\lambda_k(\xi, B^{(2)}(t))} \varphi_k(0) \langle \varphi_k, 1 \rangle$  are substantially worse.



# **Earlier results**

- [SZNITMAN 98] (Brownian motion among Poisson obstacles) and [GÄRTNER/K./MOLCHANOV 07] (double-exponential distribution): mass concentration a.s. in t<sup>o(1)</sup> islands.
- [K./LACOIN/MÖRTERS/SIDOROVA 09] (Pareto distribution): mass concentration in one site in probability, and in two sites a.s.
- [KILLIP/NAKANO 07], [GERMINET/KLOPP 10] (bounded distributions with smooth density): Poisson process convergence for rescaled eigenvalues and localisation centers of eigenfunctions in large boxes in the localised regime



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We are working here for  $\xi$  double-exponentially distributed, i.e., for some  $\varrho \in (0, \infty)$ ,

$$\operatorname{Prob}(\xi(0) > r) = \exp\left\{-e^{r/\varrho}\right\}, \quad r \in \mathbb{R}.$$

#### Earlier papers:

[GÄRTNER/MOLCHANOV 98], [GÄRTNER/DEN HOLLANDER 99], [GÄRTNER/K./MOLCHANOV 07]. The potential is unbounded to  $+\infty$ . The islands are of bounded size. The potential and of the solution approach (after shifting and normalization) certain shapes, which are given by a characteristic variational formula.



Abbreviate  $B_L = L \times \left[-\frac{1}{2}, \frac{1}{2}\right]^d$ .

#### Theorem 1

There is a number  $\chi = \chi_{\varrho} \in (0, 2d)$  and a sequence  $(a_L)_{L \in \mathbb{N}}$  with  $a_L = \rho \log \log |B_L| - \chi + o(1)$  as  $L \to \infty$  and, for any  $L \in \mathbb{N}$ , a sequence  $(X_k^{(L)})_k$  in  $B_L$  such that, in probability,

$$\lim_{L \to \infty} \sum_{\substack{z \colon |z - X_k^{(L)}| \le \log L}} \varphi_k(z)^2 = 1, \quad k \in \mathbb{N},$$

and the law of

$$\sum_{k \in \mathbb{N}} \delta_{\left(\frac{X_k^{(L)}}{L}, (\lambda_k(\xi, B_L) - a_L) \log L\right)}$$

converges weakly to a Poisson process on  $B_1 \times \mathbb{R}$  with intensity measure  $dx \otimes e^{-\lambda} d\lambda$ .





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converges weakly to a Poisson process on  $B_1 \times \mathbb{R}$  with intensity measure  $dx \otimes e^{-\lambda} d\lambda$ .

Hence, the top eigenvalues in  $B_L$  are of order  $\log \log L$ , leave gaps of order  $1/\log L$ , are in the max-domain of attraction of the Gumbel distribution, and the localisation centres are separated by a distance of order L and are uniformly distributed.



#### Theorem 2

Put  $r_L = L \log L \log \log \log L$  and

$$\Psi_{L,t}(z,\lambda) = \frac{t}{r_L}(\lambda - a_L)\log L - \frac{|z|}{L},$$

and pick k such that  $\Psi_{L,t}(X_k^{(L)}, \lambda_k(\xi, B_L))$  is maximal, and put  $Z_{L,t} = X_k^{(L)}/L$ . Then, with  $L_t$  defined by  $r_{L_t} = t$ ,

$$\lim_{t\to\infty} \frac{1}{U(t)} \sum_{z \colon |z-L_t Z_{L_t,t}| \leq R_t} u(t,z) = 1 \qquad \text{in probability},$$

for any  $R_t \gg \log t$ .

Hence, the total mass essentially comes from a single  $\gg \log t$ -island in the centred box  $B^{(1)}(t)$  with radius  $\approx t/(\log t\,\log\log\log t)$ 



# Ageing

A system is said to age if its significant changes come after longer and longer (or shorter and shorter) time lags, such that one can see from the frequency of changes how much time has elapsed.

Ageing properties of the PAM can now be studied in terms of the time lags between jumps of the concentration site.

These ones, in turn, may be described as follows.

### **Theorem 3: Scaling limit of concentration location**

As  $L \to \infty$ , the process  $(Z_{L,tr_L})_{t \in [0,\infty)}$  converges in distribution to the process of maximizers of  $z \mapsto t\lambda - |z|$  over the points  $(z, \lambda)$  of a Poisson process on  $[-\frac{1}{2}, \frac{1}{2}]^d \times \mathbb{R}$  with intensity measure  $dx \otimes e^{-\lambda} d\lambda$ .



The rescaled and shifted eigenvalues  $(\lambda_k(\xi, B_L) - a_L) \log L$  are asymptotically independent and lie in the max-domain of attraction of the Gumbel distribution.



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- The control from Poisson process convergence holds only in the box  $B^{(1)}(t)$  of radius  $\approx t/\log t \log \log \log t$ , and the contribution from the outside of the box  $B^{(2)}(t)$  of radius  $t \log^2 t$  is easily seen to be negligible. The treatment of the region  $B^{(2)}(t) \setminus B^{(1)}(t)$  is delicate and requires a comparison of the two eigenvalue expansions.



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- The two terms in the optimized functional  $\Psi_{L,t}$  come from the eigenvalue and the probabilistic cost for the random walk in the Feynman-Kac formula to reach the island. The latter term can also be seen as describing the decay of the eigenfunction term  $\varphi_k(0)$ .



# Some elements of the proof of Theorem 1 (I)

The top eigenvalues in  $B=B_L$  remain the top eigenvalues after discarding potential values significantly less than the eigenvalues. Put  $\varepsilon_R=2d\Big(1+\frac{A}{2d}\Big)^{1-2R}$ .

# **Proposition 1**

Fix 
$$A > 0$$
 and  $R \in \mathbb{N}$  and put  $U = \bigcup_{z \in B : |\xi(z) \ge \lambda_1(\xi, B)} B_R(z)$ . Then

 $\lambda_k(\xi, B) \ge \lambda_1(\xi, B) - A/2 \implies |\lambda_k(\xi, B) - \lambda_k(\xi, U)| \le \varepsilon_R.$ 

- Any ℓ<sup>2</sup>-normalized eigenvector v = v<sub>k,ξ</sub> with eigenvalue λ = λ<sub>k</sub>(ξ, B) ≥ λ<sub>1</sub> − A/2 decays rapidly away from U.
- Proof uses the martingale  $(v(Y_n) \prod_{k=0}^{n-1} \frac{2d}{2d+\lambda-\xi(Y_k)})_{n \in \mathbb{N}}$  (with  $(Y_n)_n$  an SRW).



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 $\lambda_k(\xi, B) \ge \lambda_1(\xi, B) - A/2 \implies |\lambda_k(\xi, B) - \lambda_k(\xi, U)| \le \varepsilon_R.$ 

- Any  $\ell^2$ -normalized eigenvector  $v = v_{k,\xi}$  with eigenvalue  $\lambda = \lambda_k(\xi, B) \ge \lambda_1 A/2$  decays rapidly away from U.
- Proof uses the martingale  $(v(Y_n) \prod_{k=0}^{n-1} \frac{2d}{2d+\lambda-\xi(Y_k)})_{n \in \mathbb{N}}$  (with  $(Y_n)_n$  an SRW).
- Furthermore, we use that  $\partial_{\xi(z)}\lambda_k(\xi,B) = v(z)^2$ .
- Introduce  $\xi_s = \xi s \mathbbm{1}_{B \setminus U}$  for  $s \in [0,\infty].$  Then

$$|\partial_s \lambda_k(\xi_s, B)| = \sum_{z \in B \setminus U} v_{k,\xi_s}(z)^2,$$

which is very small. Integrating over  $s\in[0,\infty]$  gives the estimate.



A bit more precisely, with the help of the variational characterisation of the asymptotics of the PAM [GÄRTNER/K./MOLCHANOV 07], one proves the following.

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- For any component C, if  $\lambda_1(\xi, C)$  is close to  $a_L \approx \rho \log \log L$ , then  $\lambda_1(\xi, C)$  is bounded away from  $\lambda_2(\xi, C)$ .



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If  $\lambda$  is an eigenvalue of  $\Delta + \xi$  larger than  $\lambda_1(\xi, B_L) - A/2$  and v a corresponding  $\ell^2$ -normalised eigenfunction such that the distance of  $\lambda$  to the nearest eigenvalue (spectral gap) is larger than  $3\varepsilon_R$ , then v decays exponentially away from one of the components of U.



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- The proof uses that the path  $[0, \infty] \ni s \mapsto \lambda_k(\xi_s, B_L)$  (with  $\xi_s = \xi s \mathbb{1}_{B_L \setminus U}$ ) does not cross other eigenvalues and therefore admits a continuous choice of corresponding eigenfunctions. The one for  $s = \infty$  puts all its mass in one component, and the one for s = 0 is uniformly close.



The scale  $a_L$  satisfies  $\operatorname{Prob}(\lambda_1(\xi, B_R) > a_L) = 1/|B_L|$ , hence we may expect finitely many sites in  $B_L$  where the local eigenvalue is  $\approx a_L$ . Then the random variable  $\lambda_1(\xi, B_R)$  lies in the max-domain of a Gumbel random variable:

# **Proposition 2**

As  $L \to \infty$ , for any  $s \in \mathbb{R}$ ,

$$\operatorname{Prob}(\lambda_1(\xi, B_R) > a_L + s/\log L) = e^{-s} \frac{1}{|B_L|} (1 + o(1)).$$

The event  $\{\lambda_1(\xi, B_R) > a\}$  is more or less the same as the event that some shift of the potential  $\xi(\cdot)$  is larger than  $a + \chi + \psi(\cdot)$  for some well-chosen function  $\psi$ .



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- Shifting  $\xi$  by an amount of  $s/\log L$  yields an additional factor of  $e^{-s}$ , using properties of  $\psi$  and of the distribution of  $\xi$  and some information from the variational characterisation.



# Some elements of the proof of Theorem 2 (I)

Consider boxes  $B^{(1)}(t)=B_{L_t^{(1)}}$  and  $B^{(2)}(t)=B_{L_t^{(2)}}$  with

$$L_t^{(1)} = \mathrm{const.} \times \frac{t}{\log t \, \log \log \log t} \qquad \text{and} \qquad L_t^{(2)} = t \log^2 t.$$

Inside B<sup>(1)</sup>(t), we have the Poisson process convergence.
 Outside B<sup>(2)</sup>(t), the contribution is negligible.

Why is the contribution from  $B^{(2)}(t) \setminus B^{(1)}(t)$  negligible?

#### Our Strategy:

- Consider the eigenvalue expansion in  $B^{(2)}(t)$ . A version of Minami's estimate gives that each spectral gap close to the top is  $\geq \varepsilon_R$ , with high probability.
- This enables us to prove exponential localisation of the top eigenfunctions in  $B^{(2)}(t)$ . This makes the top eigenvalues in  $B^{(2)}(t)$  essentially independent.



Our Strategy (continued):

- The top eigenvalues of  $B^{(1)}(t)$  are also top eigenvalues in  $B^{(2)}(t)$ . But the  $B^{(2)}(t)$ -eigenfunctions are located much further away (if 'const' is large). Hence their contributions in the eigenvalue expansion are negligible w.r.t. the optimizer of  $\Psi_{L_t^{(1)},t}$  in  $B^{(1)}(t)$ .
- For N large enough, the eigenvalues  $\lambda_k(\xi, B^{(2)}(t))$  for k > N are negligible w.r.t. the optimizer of  $\Psi_{L_t^{(1)}, t}$  in  $B^{(1)}(t)$  and hence their contribution to the eigenvalue expansion.
  - The remaining N eigenvalues can be ordered with gaps  $\approx 1/\log L_t^{(1)} \approx 1/\log t$  between them, and the optimizer is among them.



#### **Future directions**

- Presumably, the mass concentration property of the PAM also holds in almost sure sense, but the assertion must be adapted.
- Control on the process of localisation centres,  $(Z_{L_t,tr_L_t})_{t \in [0,\infty)}$ , opens up the possibility to study the time-evolution of the PAM, e.g. in terms of ageing properties.
- Replacing double-exponential distribution by bounded distributions will lead to the same max-domains of attractions for the top eigenvalues, but other rescalings of the gaps.

