

Localization and delocalization in Erdős-Rényi graphs

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Universality conjecture for disordered quantum systems

A disordered quantum system can be in one of two phases:

(1) **Localized** (insulator, strong disorder):

Eigenvectors are localized.

Local spectral statistics are Poisson.

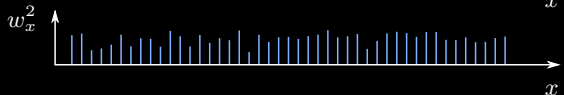
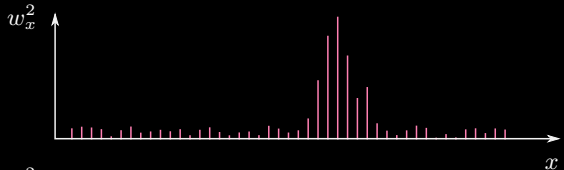
(2) **Delocalized** (metal, weak disorder):

Eigenvectors are delocalized.

Local spectral statistics follow random matrix theory (e.g. GOE).

ℓ^2 -normalized eigenvector

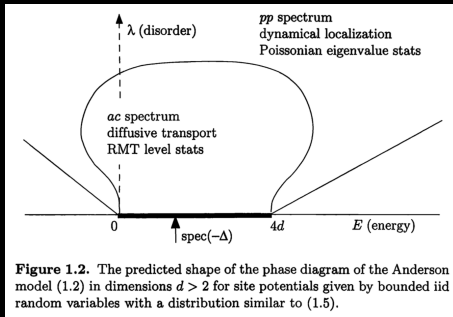
$\mathbf{w} = (w_x)$:



Conjectured phase diagram for Anderson model

Standard model of quantum disorder: Anderson model

$$-\Delta + \lambda V \quad \text{on } \Lambda \subset \mathbb{Z}^d, \quad V = (V_x)_{x \in \Lambda} \text{ i.i.d. } \mathcal{N}(0, 1).$$



(From M. Aizenman, S. Warzel, *Random Operators*, AMS.)

Localized phase very well understood ([Fröhlich, Spencer; 1983], [Aizenman, Molchanov; 1993], [Molchanov; 1981], [Minami; 1996], ...)

Delocalized phase wide open (extended states conjecture).

Random matrices

- Wigner matrices with light tails are in the delocalized phase [Erdős, Schlein, Yau, Yin; 2009–...], [Tao, Vu; 2009–...].
- Heavy-tailed Wigner matrices proposed as a simple model that exhibits a phase transition [Cizeau, Bouchaud; 1994], [Tarquini, Biroli, Tarzia; 2016].

For $1 < \alpha < 2$, any bounded interval lies in the delocalized phase [Bordenave, Guionnet; 2013], [Aggarwal, Lopatto, Yau; 2020].

For $0 < \alpha < 1$, delocalized phase in some neighbourhood of origin [Bordenave, Guionnet; 2017], [Aggarwal, Lopatto, Yau; 2020].

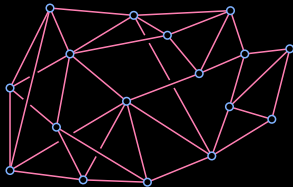
For $0 < \alpha < 2/3$, partially localized phase far away from origin [Bordenave, Guionnet; 2013].

- This talk: Sparse matrices.

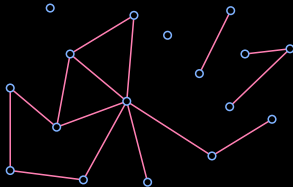
Erdős-Rényi graph and critical regime

Erdős-Rényi graph $\mathbb{G}(N, d/N)$

Critical regime: $d \approx \log N$, below which degrees do not concentrate.



$d \gg \log N$



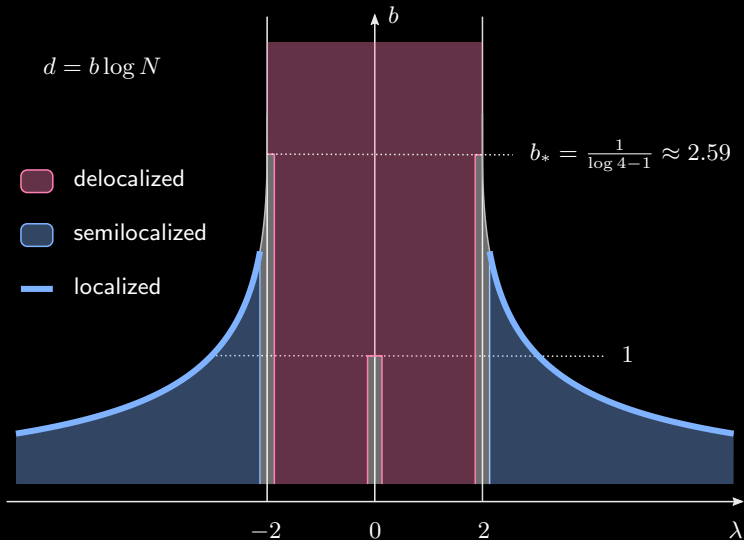
$d \ll \log N$

Supercritical $d \gg \log N$: **homogeneous**.

Subcritical $d \ll \log N$: **inhomogeneous** (hubs, leaves, isolated vertices, ...).

Consider the adjacency matrix $A = (A_{xy}) \in \{0, 1\}^{N \times N}$.

Phase diagram for $H := d^{-1/2}A$

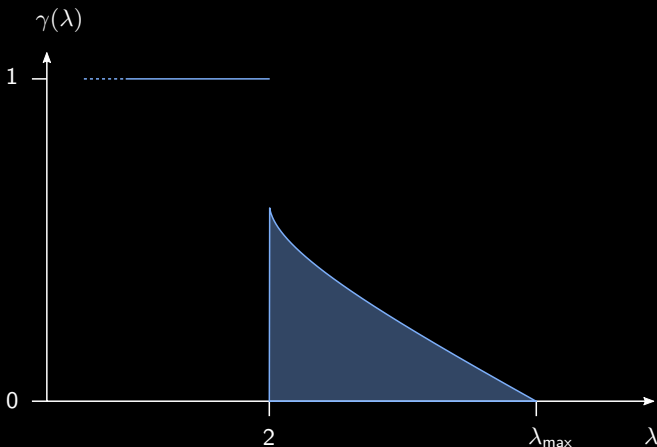


Behaviour of localization exponent

Eigenvalue λ with eigenvector \mathbf{w} has localization exponent $\gamma(\lambda) \in [0, 1]$:

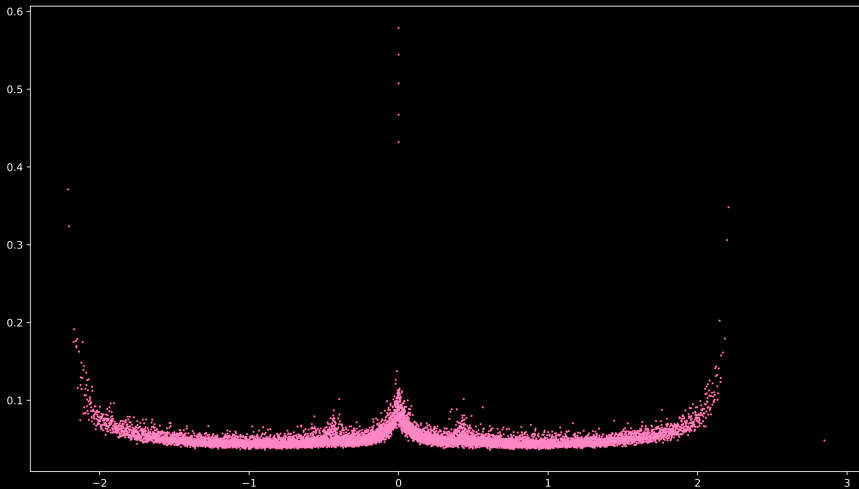
$$\|\mathbf{w}\|_{\infty}^2 =: N^{-\gamma(\lambda)}.$$

Asymptotically allowed region for $\gamma(\lambda)$ (plotted for $b = 1$):



Simulation of eigenvectors

Scatter plot of (eigenvalue, $\|\text{eigenvector}\|_\infty$). ($N = 10'000$, $b = 0.6$)



Localized Phase I: Poisson eigenvalue statistics

Theorem [Alt, Ducatez, K; 2021]. Suppose that

$$(\log \log N)^4 \leq d \leq (1 - o(1)) b_* \log N .$$

There exist deterministic u, σ, τ, θ (which are explicit functions of d and N) such that the rescaled eigenvalue process

$$\Phi := \sum_i \delta_{d\tau(\lambda_i(H) - \sigma)}$$

is asymptotically close to a Poisson point process Ψ on \mathbb{R} on intervals $[-\kappa, \infty)$ containing at most $\mathcal{K} \gg 1$ points.

Corollary. Asymptotic equality in law of $k = O(\mathcal{K})$ largest points.

Intensity of Ψ

The intensity of Ψ is

$$\rho(ds) := \sum_{\ell \in \mathbb{Z}} u^{\langle du \rangle + \ell} g(s + \theta(\langle du \rangle + \ell)) ds,$$

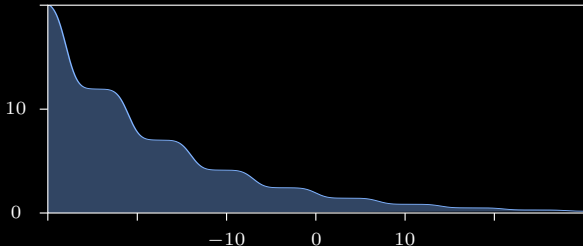
where $\langle \cdot \rangle$ is the periodic representative in $[-1/2, 1/2)$, and $g(s) := \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}s^2}$.

Scaling laws

$$u \asymp \tau \asymp \sigma^2 \asymp \theta^2 \asymp \frac{t}{\log(t \vee 2)}, \quad t := \frac{\log N}{d}.$$

Distribution of ρ in
critical regime $t \asymp 1$:

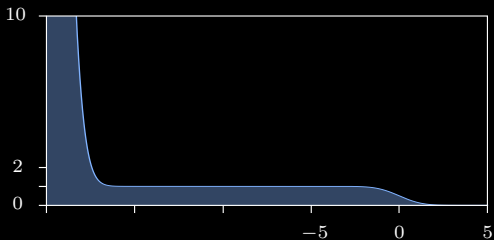
Top eigenvalue not
governed by Gumbel
law.



Distribution of ρ in **subcritical regime** $t \gg 1$:

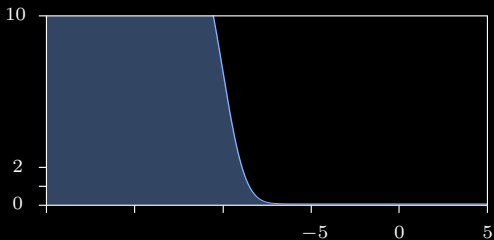
Resonance $\langle du \rangle = 0$:

Top eigenvalue not governed by Gumbel law.



Off-resonance $|\langle du \rangle| \geq c$:

Top eigenvalue governed by Gumbel law.



Localized Phase II: eigenvector localization

Let x be a vertex whose normalized degree $\alpha_x := |S_1(x)|/d$ is greater than 2. Define

$$u_1(x) = \frac{\sqrt{\alpha_x}}{\sqrt{\alpha_x - 1}} u_0(x), \quad u_{i+1}(x) = \frac{1}{\sqrt{\alpha_x - 1}} u_i(x) \quad (i \geq 1),$$

and the exponentially decaying radial vector (with $r \gg 1$)

$$\mathbf{v}(x) := \sum_{i=0}^r u_i(x) \frac{\mathbf{1}_{S_i(x)}}{\|\mathbf{1}_{S_i(x)}\|}.$$

Theorem [Alt, Ducatez, K; 2021]. Let $\mathbf{w} = (w_x)$ be an eigenvector associated with one of the top (or bottom) \mathcal{K} eigenvalues. Then with high probability there exists a vertex x with $\alpha_x > 2$ such that $\|\mathbf{w} - \mathbf{v}(x)\| = o(1)$.

Overview of proof

Basic intuition: one-to-one correspondence between eigenvalues and vertices of large degree.

Main steps of proof:

- Step 1.** Characterize the fluctuations of an eigenvalue associated with a vertex of large degree.
- Step 2.** Establish a one-to-one relation between such eigenvalues and the eigenvalues of H near the edge.

Step 1

Consider neighbourhood of vertex in

$$\mathcal{U} := \{x : \alpha_x \geq 2 + o(1)\}.$$

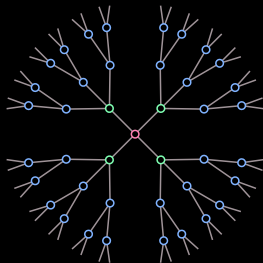
Use the **tridiagonal representation** of $H := d^{-1/2}A$ around x : write H in the basis $\mathbf{h}_0, \mathbf{h}_1, \mathbf{h}_2, \dots$ obtained by orthogonalizing $\mathbf{1}_x, H\mathbf{1}_x, H^2\mathbf{1}_x, \dots$.

Apply transfer matrix (or orthogonal polynomial) analysis.

Problem: Fluctuations of transfer matrices very hard to control precisely, because \mathbf{h}_i is unwieldy.

Toy model: in a **rooted regular tree**, the degree depends only on the distance to the root.

Exercise: if $\mathbb{G}|_{B_r(x)}$ is a rooted regular tree, then $\mathbf{h}_i = \mathbf{1}_{S_i(x)}$ for $i \leq r$.



- Naive attempt: write H in basis $(\mathbf{1}_{S_i(x)})$ instead of (\mathbf{h}_i) , to get an almost tridiagonal matrix.

Problem: off-tridiagonal matrix is too large.

- More refined attempt: If $\mathbb{G}|_{B_r(x)}$ is a tree, the vector $H^i \mathbf{1}_x$ can be decomposed as a sum over simple walks in \mathbb{N} of length i .

jump left / right \iff terms decreasing / increasing distance from root

- Basis (\mathbf{h}_i) : all walks
- Basis $(\mathbf{1}_{S_i(x)})$: only steps to the right

Define basis (\mathbf{f}_i) using walks with at most one step to the left.

For instance,

$$\mathbf{f}_3 = \mathbf{1}_{S_3(x)} + \sum_{y \in S_1(x)} (d\alpha_y - F) \mathbf{1}_y, \quad F \in \mathbb{R}.$$

$Z_{\mathfrak{d}}(\alpha_x, \beta_x)$ has a unique eigenvalue $\Lambda_{\mathfrak{d}}(\alpha_x, \beta_x) > 2 + o(1)$, with exponentially decaying eigenvector $(u_i)_{i=0}^r$.

Back to graph \mathbb{G} with

$$\mathbf{y}(x) := \sum_{i=0}^r u_i \frac{\mathbf{f}_i}{\|\mathbf{f}_i\|}.$$

It is possible to show that

$$\|(H - \Lambda_{\mathfrak{d}}(\alpha_x, \beta_x))\mathbf{y}(x)\| \leq d^{-1-c}. \quad (1)$$

Step 1 is concluded by analysing the fluctuations of $\Lambda_{\mathfrak{d}}(\alpha_x, \beta_x)$ (of order d^{-1}).

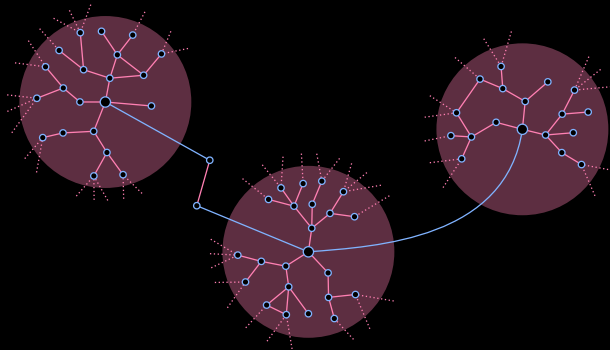
Step 2

Need to ensure:

- (a) $(\mathbf{y}(x) : x \in \mathcal{U})$ are orthogonal (i.e. $(B_r(x) : x \in \mathcal{U})$ are disjoint).
- (b) The high probability bounds hold simultaneously for all $x \in \mathcal{U}$.
- (c) The remaining eigenvalues cannot “pollute” the edge of the spectrum.

All of these present significant complications. In fact, (a) and (b) are wrong.

(a) $(B_r(x) : x \in \mathcal{U})$ are disjoint only if either (i) \mathcal{U} is small or (ii) we **prune** the graph by removing edges to disconnect balls.

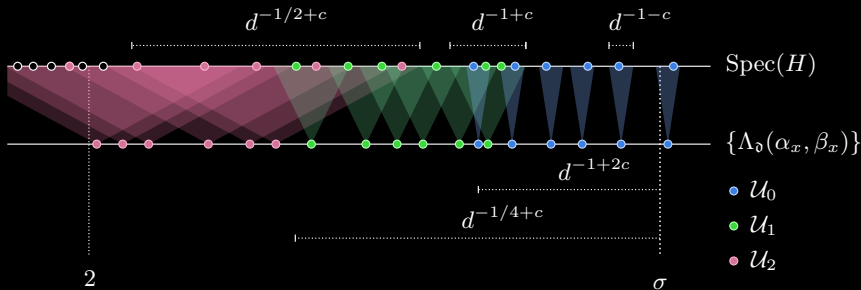


The pruning is potentially deadly, since in general removing even a single edge perturbs an eigenvalue by $O(1/\sqrt{d})$.

We have to prune in places that have a small impact on the extreme eigenvalues: prune only in the neighbourhoods of vertices x whose α_x is far from the top degree.

(b) The estimate (1) is not true simultaneously for all $x \in \mathcal{U}$. Solution: **three-scale rigidity argument** with the partition $\mathcal{U} = \mathcal{U}_0 \sqcup \mathcal{U}_1 \sqcup \mathcal{U}_2$, where $\alpha_x > \alpha_y$ for $x \in \mathcal{U}_i$ and $y \in \mathcal{U}_{i+1}$.

The sets $\mathcal{U}_0, \mathcal{U}_1, \mathcal{U}_2$ are increasing in size, but the accuracy of the estimate (1) deteriorates as i increases.



Block diagonal representation

$$O^{-1}HO = \begin{pmatrix} \mathcal{D}_0 & 0 & 0 & E_0^* \\ 0 & \mathcal{D}_1 & 0 & E_1^* \\ 0 & 0 & \mathcal{D}_2 + \mathcal{E}_2 & E_2^* \\ E_0 & E_1 & E_2 & X \end{pmatrix}$$

where

$$\mathcal{D}_i = \text{diag}(\Lambda_{\mathfrak{d}}(\alpha_x, \beta_x) + O(\kappa_i) : x \in \mathcal{U}_i)$$

$$\kappa_i + \|E_i\| = \begin{cases} d^{-1-c} & \text{if } i = 0 \\ d^{-1+c} & \text{if } i = 1 \\ d^{-1/2+c} & \text{if } i = 2 \end{cases} \quad \leftarrow \text{main estimates}$$

$$\|\mathcal{E}_2\| = O(d^{-1/2+c}) \quad \leftarrow \text{pruning}$$

$$\|X\| \leq 2 + o(1) \quad \leftarrow \text{(c)}$$

(c) Estimate of $\|X\|$ relies on:

- analysis of nonbacktracking version of A
- approximate Ihara-Bass identities
- local delocalization bounds from a radial Combes-Thomas argument