



Eigenvector distributions for certain random matrix models in the intermediate regime

with

Eugene Bogomolny (Université Paris-Sud)

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Martin Sieber



- Introduction
- Rosenzweig-Porter model
- Power law random-banded matrices and ultrametric matrices

The matrix ensembles in this talk are real Hermitian

1) Eugene Bogomolny & M.S., PRE **98**, 032139 (2018)

2) Eugene Bogomolny & M.S., PRE **98**, 042116 (2018)

Motivations for unusual random matrix ensembles

Random regular graphs with on-site energy disorder:

- Anderson Transition between localised and extended states on Bethe lattices (infinite regular trees) (Abou-Chacra et al, 1973)
- No consensus about extended states on random regular graphs (RRG)
 - Only one ergodic extended phase (Mirlin, Tikhonov, 2018; Biroli, Tarzia, 2018)
 - There is a second transition between ergodic and non-ergodic states with non-trivial fractal dimensions (Kravtsov et al, 2018)

Porter-Thomas distribution for eigenstates:

- The distribution of eigenvectors is universal for all standard invariant ensembles distribution. For real matrices: ($x = \sqrt{N}\psi_j$)

$$P(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad P(y = x^2) = \frac{e^{-y/2}}{\sqrt{2\pi y}}, \quad \langle \psi^2 \rangle = 1$$

- Recent experimental neutron resonance data are in contradiction with this distribution (Koehler et al, 2011 and 2013)

Rosenzweig-Porter model

- Each element is i.i.d. Gaussian variable (up to symmetry)

$$\langle H_{ij} \rangle = 0, \quad \langle H_{ij}^2 \rangle = 1, \quad \langle H_{ij}^2 \rangle_{i \neq j} = \frac{\epsilon^2}{N^\gamma}, \quad 1 \leq i, j \leq N$$

- Rule of thumb for the different regimes

$$S_1(N) = \frac{1}{N} \sum_{i,j=1}^N \langle |H_{ij}| \rangle, \quad S_2(N) = \frac{1}{N} \sum_{i,j=1}^N \langle |H_{ij}|^2 \rangle.$$

- If $\lim_{N \rightarrow \infty} S_1(N) < \infty \implies$ eigenvectors are localised and the spectral statistics is Poissonian
- If $\lim_{N \rightarrow \infty} S_2(N) = \infty \implies$ eigenvectors are fully delocalised and the spectral statistics is GOE
- $\gamma > 2 \implies$ localisation
- $\gamma < 1 \implies$ standard GOE

Intermediate region: $1 < \gamma < 2$ (Kravtsov et al, 2015)

Moments of eigenvectors ($q > 1/2$)

$$I_q = \langle \sum_j |\psi_j|^{2q} \rangle \xrightarrow{N \rightarrow \infty} C_q N^{-(q-1)D_q}$$

where D_q is the fractal dimension

- Localised regime ($\gamma > 2$): $D_q = 0$
- Ergodic regime ($\gamma < 1$): $D_q = 1$
- Intermediate regime ($1 < \gamma < 2$): $D_q = 2 - \gamma$

Recent rigorous proofs (von Soosten & Warzel, 2017)

In this talk: distribution of eigenvectors when $1 < \gamma < 2$ based on

- Breit-Wigner distribution of the variances $\langle |\psi_j(E)|^2 \rangle$
- Local Gaussian distribution for $\psi_j(E)$

Follows from rigorous results (Benigni, 2017)

Breit-Wigner distribution of eigenvector variances

$$\Sigma_j^2(E) \equiv \langle |\psi_j(E)|^2 \rangle \approx \frac{\Gamma(E)}{\pi \rho(E) N [(E - e_j)^2 + \Gamma^2(E)]}$$

Average is over off-diagonal elements, diagonal elements $e_j = H_{jj}$ are fixed.

- The spreading width $\Gamma(E)$ is given by the Fermi golden rule

$$\Gamma(E) = \frac{\pi \epsilon^2}{N^{\gamma-1}} \rho(E)$$

- The normalised level density $\rho(E)$ is given by

$$\rho(E) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{E^2}{2}\right)$$

the density of diagonal elements for $N \rightarrow \infty$ and $\gamma > 1$.

- Standard normalisation is assumed

$$\sum_j |\psi_j(E_\alpha)|^2 = 1 \quad \text{or} \quad \sum_\alpha |\psi_j(E_\alpha)|^2 = 1$$

Derivation of the Breit-Wigner formula

Recursive relation for the Green function $G = (E - i\eta - H)^{-1}$

$$G_{ii}(E - i\eta) = \left(E - i\eta - H_{ii} - \sum_{j,k \neq i} H_{ij} G_{jk}^{(i)}(E - i\eta) H_{ki} \right)^{-1}$$

where $G(E)^{(i)}$ is the Green function after removing the row and column i from H . (Schur complement formula, also Feshbach's projection method)

For large N

$$\sum_{j,k \neq i} H_{ij} G_{jk}^{(i)} H_{ki} \approx \frac{\epsilon^2}{N^\gamma} \sum_{j \neq i} G_{jj}^{(i)} \xrightarrow{N \rightarrow \infty} \frac{\epsilon^2}{N^\gamma} \int \frac{N \rho(e) de}{E - i\eta - e}$$

The variance $\langle |\Psi_i(E)|^2 \rangle$ follows from

$$\text{Im } G_{ii}(E - i\eta) \xrightarrow{\eta \rightarrow 0} \pi \langle |\Psi_i(E)|^2 \rangle \rho(E)$$

Full eigenvector distribution

The second ingredient is a local Gaussian distribution of $\Psi_j(E)$ (for fixed e_j)

$$P(\Psi_j(E)) = \frac{1}{\sqrt{2\pi\Sigma_j^2(E)}} \exp\left(-\frac{|\Psi_j(E)|^2}{2\Sigma_j^2(E)}\right)$$

Integrating over the diagonal element e_j gives $[x = \Psi_j(E)]$

$$P(x)_E = \int \frac{\rho(E)}{\sqrt{2\pi\Sigma_j^2(E)}} \exp\left(-\frac{x^2}{2\Sigma_j^2(E)}\right) d e_j$$

Result for the distribution in a small window around $E = 0$

$$P(x)_{E=0} = \frac{\delta^2}{4\pi\sqrt{a}} [K_0(\zeta) + K_1(\zeta)] e^{-\zeta + \frac{\delta^2}{2}}$$

where

$$a = \frac{C^2 \epsilon^2}{N^\gamma}, \quad \delta = \Gamma(0) = \frac{\sqrt{\pi} \epsilon^2}{\sqrt{2} N^{\gamma-1}}, \quad \zeta = \frac{\delta^2}{4a} (x^2 + a).$$

Distribution in the bulk and in the tail

In the bulk, x has values of the order of \sqrt{a} . It is convenient to scale

$$y = N^{\gamma/2} \psi_j(E)$$

As $N \rightarrow \infty$

$$P_{\text{bulk}}(y) \approx \frac{\epsilon}{\pi(y^2 + \epsilon^2)}$$

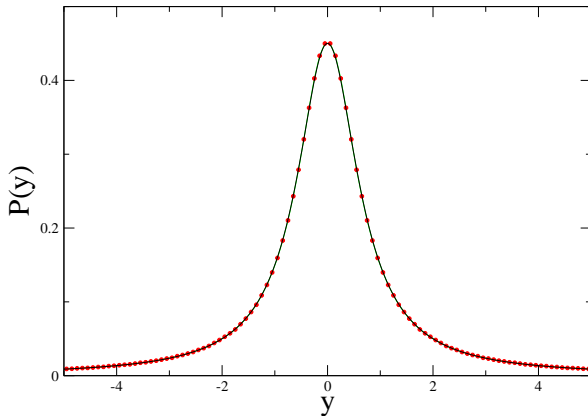
In the tail (small δ , finite ζ) it is convenient to rescale

$$z = N^{1-\gamma/2}$$

Then

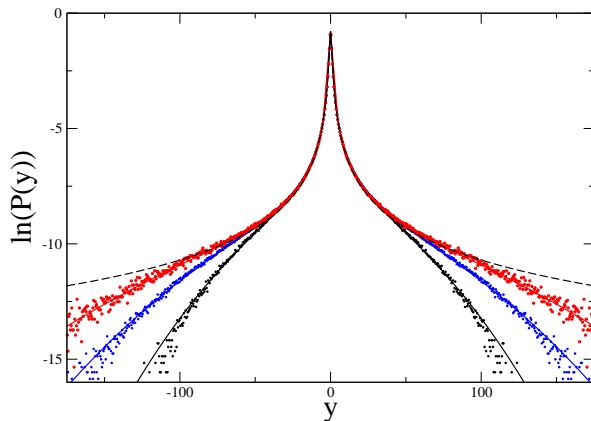
$$P_{\text{tail}}(z) = \frac{2\sqrt{2}b^3}{\pi\sqrt{\pi}N^{\gamma-1}}(K_0(b^2z^2) + K_1(b^2z^2))e^{-b^2z^2}, \quad b = \frac{\sqrt{\pi}\epsilon}{2\sqrt{2}}$$

Distribution of eigenvector components in the bulk



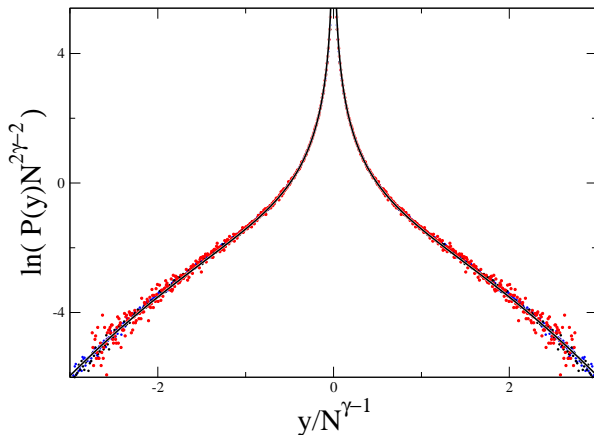
Distribution of $y = N^{\gamma/2} \psi_j(E)$ for the RP model with $\gamma = 1.5$ and $\epsilon = \frac{1}{\sqrt{2}}$ for $N = 4096, 2048, 1024$.

Distribution in logarithmic scale



Distribution of $y = N^{\gamma/2} \psi_j(E)$ for the RP model with $\gamma = 1.5$ and $\epsilon = \frac{1}{\sqrt{2}}$ for $N = 4096$ (red), $N = 2048$ (blue) and $N = 1024$ (black).

Rescaled distribution of eigenvector components in the tail



Distribution of $z = N^{1-\gamma/2} \psi_j(E)$ for the RP model with $\gamma = 1.5$ and $\epsilon = \frac{1}{\sqrt{2}}$ for $N = 4096$ (red), $N = 2048$ (blue) and $N = 1024$ (black).

Moments of the eigenvectors

Results for the centre of the spectrum

$$I_q \equiv \left\langle \sum_{j=1}^N |\Psi_j(E)|^{2q} \right\rangle = \frac{2^{q-1/2} a^q \Gamma(q+1/2)}{\sqrt{\pi} \delta^{2q-1}} \Psi\left(\frac{1}{2}, \frac{3}{2} - q; \frac{\delta^2}{2}\right)$$

where $\Psi(\alpha, \beta; z)$ is the Tricomi confluent hypergeometric function

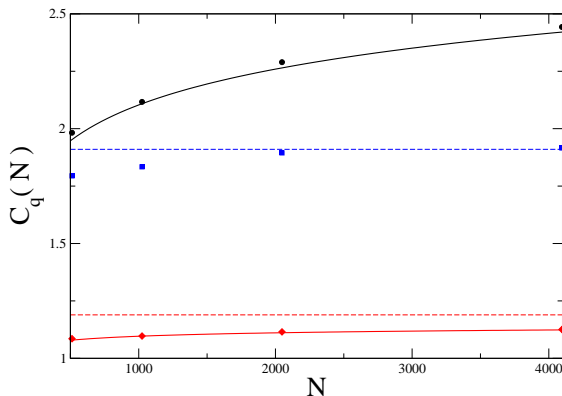
In the limit $\delta \rightarrow 0$

$$\begin{aligned} I_{q > \frac{1}{2}} &= N^{-(q-1)(2-\gamma)} C_{q > \frac{1}{2}}, & C_{q > \frac{1}{2}} &= \frac{\Gamma(q-1/2)\Gamma(q+1/2)}{\pi b^{2q-2} 2^{q-2} \Gamma(q)} \\ I_{q = \frac{1}{2}} &= N^{1-\gamma/2} C_{\frac{1}{2}}, & C_{\frac{1}{2}} &= \frac{\epsilon}{\pi} \left[2(\gamma-1) \ln N - \ln\left(\frac{\pi \epsilon^4}{16}\right) - \gamma \right] \\ I_{q < \frac{1}{2}} &= N^{-\gamma q + 1} C_{q < \frac{1}{2}}, & C_{q < \frac{1}{2}} &= \frac{\epsilon^{2q}}{\pi} \Gamma(q+1/2) \Gamma(1/2-q) c_{\text{cor}}(q) \end{aligned}$$

Corrective factor for $q < 1/2$

$$c_{\text{cor}}(q) = 1 + \frac{\pi^{1-q} \epsilon^{2-4q} \Gamma(q-1/2)}{2^{1-2q} \Gamma(q) \Gamma(1/2-q)} N^{-(\gamma-1)(1-2q)}$$

Eigenvector moments for $N = 512, 1024, 2048, 4096$



Eigenvector moments for $q = \frac{1}{8}$ (red), $q = 2$ (blue) and $q = \frac{1}{2}$ (black).

Here $C_{\frac{1}{8}} = 1.19$ with $c_{\text{cor}} = (1 - .44/N^{1/4})$, and $C_2 = 1.19$.

Summary of first part

- The statistical distribution for eigenvectors of the Rosenzweig-Porter model has been obtained in the regime $1 < \gamma < 2$.
- The derivation is based on two physical assumptions (which are exact for the considered model).
- The first states that the mean square modulus of eigenvectors is given by a Breit-Wigner formula with a spreading width in agreement with the Fermi golden rule.
- The second states that the eigenvectors have a local Gaussian distribution with variance given by the above formula.
- This approach leads to explicit formulas that agree extremely well with numerical calculations.

Power-law random banded matrices and ultrametric matrices

Each matrix element is i.i.d. Gaussian variable (up to symmetry)

$$\langle H_{ij} \rangle = 0, \quad \langle H_{ii}^2 \rangle = 2, \quad \langle H_{ij}^2 \rangle_{i \neq j} = a^2(i, j)$$

Power-law random banded matrices (Mirlin et al, 1996)

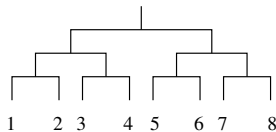
$a(r)$ with $r = |i - j|$ decreases as a power of the distance $a(r) \xrightarrow[r \rightarrow \infty]{} \epsilon r^{-s}$

A translation-invariant choice to avoid boundary effects is

$$a(r) = \epsilon \left[1 + \left(\frac{N}{\pi} \sin\left(\frac{\pi r}{N}\right) \right)^2 \right]^{-s/2}.$$

Ultrametric random matrices (Fyodorov et al, 2009)

$2^n \times 2^n$ matrices with $a(i, j) = \epsilon 2^{-s \text{dist}(i, j)}$



$\text{dist}(i, j)$ is the ultrametric distance on a binary tree.

For example, $\text{dist}(1, 2) = 1$, $\text{dist}(1, 3) = 2$ and $\text{dist}(1, 5) = 3$.

Intermediate region $\frac{1}{2} < s < 1$

The rule of thumb for the two moments $S_1(N)$ and $S_2(N)$ predicts for both ensembles

- $s > 1 \implies$ eigenvectors are localised and the spectral statistics is Poissonian
- $s < \frac{1}{2} \implies$ eigenvectors are fully delocalised and the spectral statistics is GOE

Intermediate region

$$\frac{1}{2} < s < 1$$

Due to the absence of a small or large parameter standard analytical approaches to random matrices are not applicable.

\implies Numerical investigation of the two ensembles

Main numerical results

- No indication of non-trivial fractal dimensions when $\frac{1}{2} < s < 1$.
Distribution of $x = \sqrt{N}\psi_j$ becomes quickly independent of N
- Eigenvector distribution is extremely well approximated by the **generalised hyperbolic distribution** (GHD)(symmetric case)

$$P_{\text{GHD}}(x) = \frac{\sqrt{\alpha}}{\sqrt{2\pi}\delta^\lambda K_\lambda(\alpha\delta)} (x^2 + \delta^2)^{(\lambda-1/2)/2} K_{\lambda-1/2}(\alpha\sqrt{x^2 + \delta^2})$$

GHD is a variance mixture of the normal distribution with variance distributed according to the **generalised inverse Gaussian distribution** (GIG)
(normal variance-mean mixture)

$$P_{\text{GHD}}(x) = \int_0^\infty P_{\text{GIG}}(y) \frac{e^{-x^2/2y}}{\sqrt{2\pi y}} dy$$

where

$$P_{\text{GIG}}(x) = \frac{\alpha^\lambda}{2\delta^\lambda K_\lambda(\alpha\delta)} x^{\lambda-1} e^{-\frac{1}{2}(\alpha^2 x + \delta^2 x^{-1})}$$

Parameters and moments

GHD and GIG depend on three parameters α , δ and λ .

The moments are known analytically

$$C_q \equiv \langle x^{2q} \rangle_{\text{GHD}} = C_{\text{GOE}}(q) \langle x^q \rangle_{\text{GIG}}$$

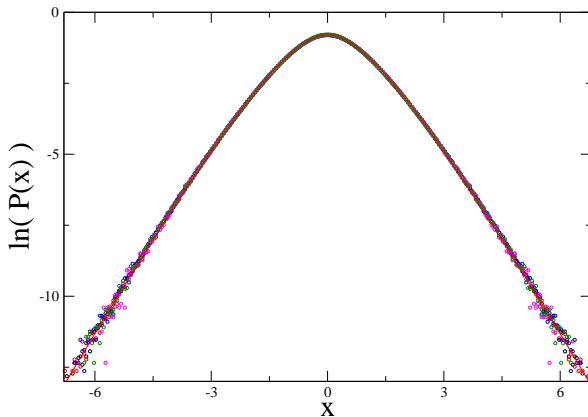
$$C_{\text{GOE}}(q) = \frac{2^q \Gamma(q + \frac{1}{2})}{\sqrt{\pi}}, \quad \langle x^q \rangle_{\text{GIG}} = \left(\frac{\delta}{\alpha}\right)^q \frac{K_{\lambda+q}(\alpha\delta)}{K_{\lambda}(\alpha\delta)}$$

The variance of the GHD is fixed to one by the normalisation. We set

$$\alpha = \sqrt{\frac{\xi K_{\lambda+1}(\xi)}{K_{\lambda}(\xi)}}, \quad \delta = \frac{\xi}{\alpha}, \quad \xi = \alpha\delta$$

With this choice the distributions depend on two parameters: λ and ξ .

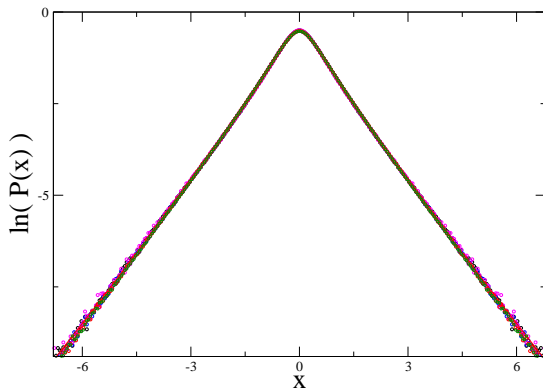
Eigenvector distribution for PLBM with $s = 0.7$ and $\epsilon = 1$



Distribution of $x = \sqrt{N}\psi_j$ for $N = 8192$ (black), $N = 4096$ (red), $N = 2048$ (blue), $N = 1024$ (green) and $N = 512$ (magenta).

Compared to GHD with $\alpha = 2.6154$, $\lambda = 3.3615$, $\delta = 0.2903$ (red line)

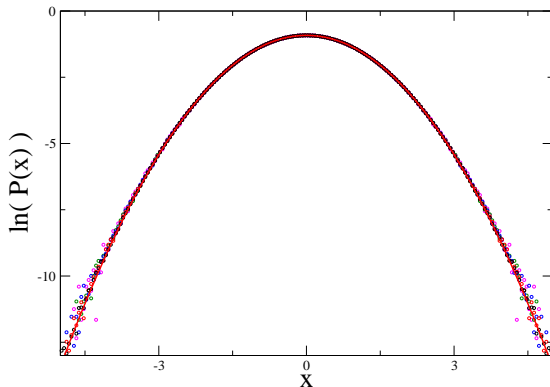
Eigenvector distribution for UMM with $s = 0.7$ and $\epsilon = 1$



Distribution of $x = \sqrt{N}\psi_j$ for $N = 8192$ (black), $N = 4096$ (red), $N = 2048$ (blue), $N = 1024$ (green) and $N = 512$ (magenta).

Compared to GHD with $\alpha = 1.1673$, $\lambda = 0.3880$, $\delta = 0.4409$ (red line)

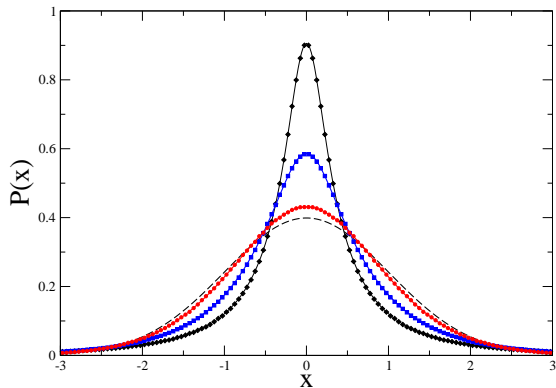
Eigenvector distribution for PLBM with $s = 0.3$ and $\epsilon = 1$ (GOE)



Distribution of $x = \sqrt{N}\psi_j$ for $N = 8192$ (black), $N = 4096$ (red), $N = 2048$ (blue), $N = 1024$ (green) and $N = 512$ (magenta).

Compared to Gaussian with zero mean and unit variance (red line)

PLBM with $s = 0.7$ and different ϵ (N=2048)



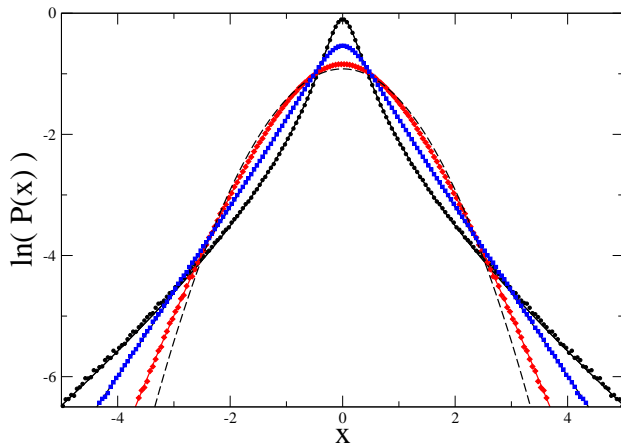
$\epsilon = 0.3$ (black circles), $\epsilon = 0.5$ (blue squares) and $\epsilon = 1.5$ (red diamond)

GHD for $\epsilon = 0.3$: $\alpha = 0.6506$, $\lambda = -0.1067$, $\delta = 0.2805$

GHD for $\epsilon = 0.5$: $\alpha = 1.2754$, $\lambda = 0.5862$, $\delta = 0.3945$

GHD for $\epsilon = 1.5$: $\alpha = 2.9341$, $\lambda = 3.6392$, $\delta = 1.0377$

PLBM with $s = 0.7$ and different ϵ in logarithmic scale ($N = 2048$)



$\epsilon = 0.3$ (black circles), $\epsilon = 0.5$ (blue squares) and $\epsilon = 1.5$ (red diamond)

Local eigenvector variance

- Choose interval $I = [E - \delta E/2, E + \delta E/2]$ with M_I consecutive levels
- Calculate local variance

$$x = \frac{1}{M_I} \sum_{E_\alpha \in I} N |\psi_j(E_\alpha)|^2$$

- Calculate the distribution $P(x)$ of x for the ensemble
- If $\psi_j(E_\alpha)$ are independent (GOE) then $P(x)$ is χ^2 -distribution with $\nu = M_I$

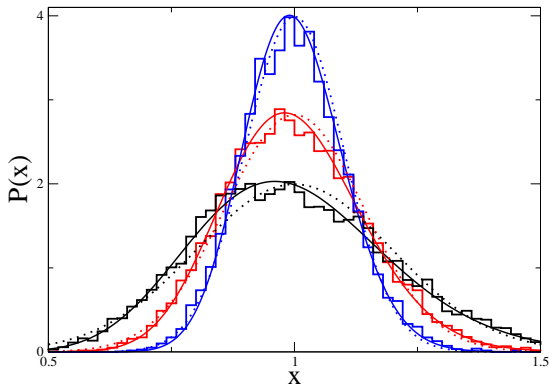
$$P_{\chi^2}(x, \nu) = \frac{\nu^\nu x^{\nu/2-1}}{2^{\nu/2} \Gamma(\nu/2)} e^{-\nu x/2}, \quad \langle x \rangle_{\chi^2} = 1.$$

- Asymptotic formula for $M_I \rightarrow \infty$ (central limit theorem)

$$P(x)_{\text{GOE}} \xrightarrow{M_I \rightarrow \infty} \sqrt{\frac{M_I}{4\pi}} e^{-M_I(x-1)^2/4}$$

- Deviation from this distribution $\longrightarrow \psi_j(E_\alpha)$ are not independent

Local eigenvector variance for PLBM with $s = 0.3$ (GOE)

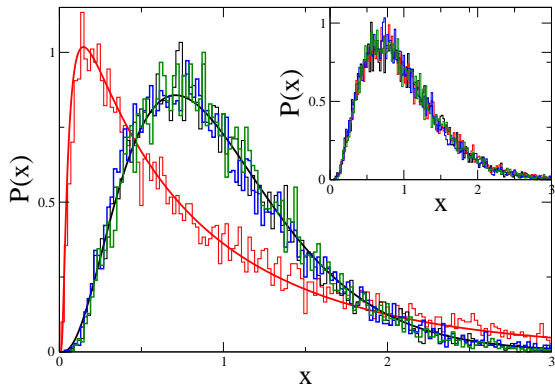


Results for $\epsilon = 1$, $N = 4096$, different M_l :

Staircase lines: $M_l = 200$ (blue), $M_l = 100$ (red), $M_l = 50$ (black)

Solid lines: χ^2 -distribution with $\nu = M_l$, and their asymptotic form (dotted)

Local eigenvector variance for PLBM with $s = 0.7$



Red staircase: $\epsilon = 0.5$, $N = 4096$, $M_l = 100$

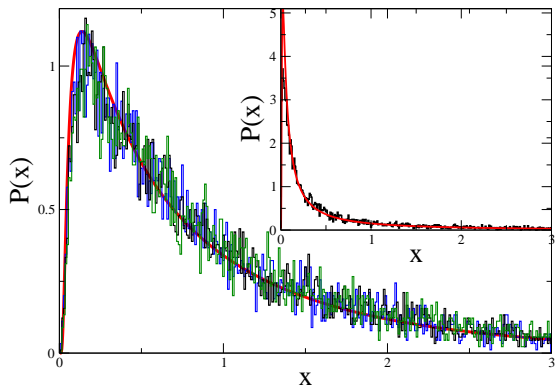
Other staircases: $\epsilon = 1$, $N = 4096$, different M_l

(blue: $M_l = 50$, black: $M_l = 100$, green: $M_l = 200$)

compared to GIG distribution with previous parameters (solid lines)

Insert: $\epsilon = 1$, $M_l = 100$ with $N = 1024, 2048, 4096, 8192$

Local eigenvector variance for UMM with $s = 0.7$



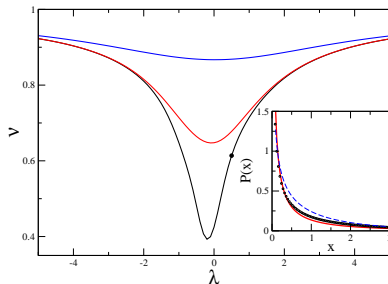
Staircases: $\epsilon = 1$, $N = 4096$, different M_l
(blue: $M_l = 50$, black: $M_l = 100$, green: $M_l = 200$)
compared to GIG distribution with previous parameters (solid red line)
Insert: $\epsilon = 0.5$, $M_l = 100$, $N = 4096$

Comparison with experimental results

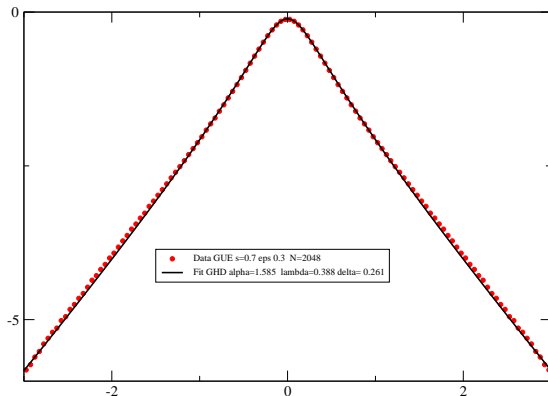
- Experimental results for neutron widths were fitted with a χ^2 -distribution

$$P_{\chi^2}(x, \nu) = \frac{\nu^{\nu/2} x^{\nu/2-1}}{2^{\nu/2} \Gamma(\nu/2)} e^{-\nu x/2}, \quad \langle x \rangle_{\chi^2} = 1.$$

- ^{192}Pt : $\nu = 0.57 \pm 0.16$, ^{194}Pt : $\nu = 0.47 \pm 0.19$, ^{196}Pt : $\nu = 0.60 \pm 0.28$
- The normalised GHD depends on 2 parameters λ and ξ
- We fixed $\xi = 0.02, 0.2, 2$ (black, red, blue) and fitted ν for different λ



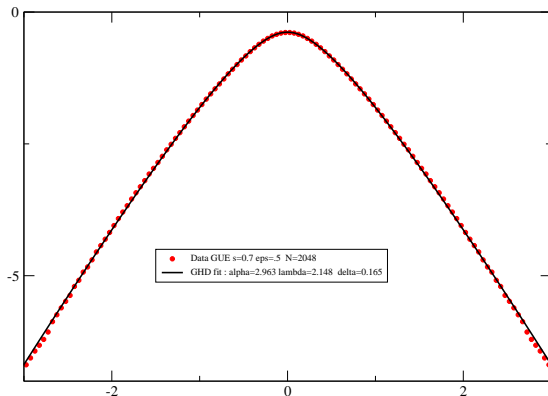
Eigenvector distribution for complex PLBM with $s = 0.7$ and $\epsilon = 0.3$



Distribution of $x = \sqrt{N}\psi_j$ for $N = 2048$ (red dots).

Compared to GHD with $\alpha = 1.585$, $\lambda = 0.388$, $\delta = 0.261$ (black line).

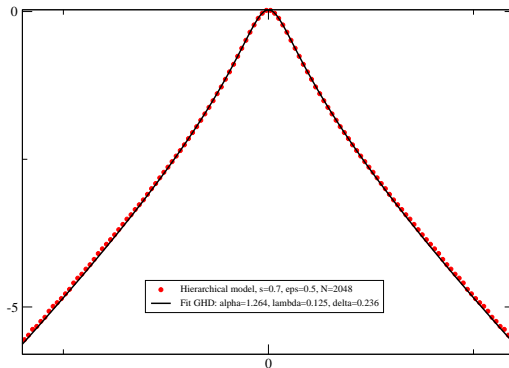
Eigenvector distribution for complex PLBM with $s = 0.7$ and $\epsilon = 0.5$



Distribution of $x = \sqrt{N}\psi_j$ for $N = 2048$ (red dots).

Compared to GHD with $\alpha = 2.963$, $\lambda = 2.148$, $\delta = 0.165$ (black line).

Eigenvector distribution for complex UMM with $s = 0.7$ and $\epsilon = 0.5$



Distribution of $x = \sqrt{N}\psi_j$ for $N = 2048$ (red dots).

Compared to GHD with $\alpha = 1.264$, $\lambda = 0.125$, $\delta = 0.236$ (black line).

Summary of second part

- Power-law banded and ultrametric matrices are representatives of random matrix ensembles with varying strength of interaction
- We numerically investigated the intermediate region $\frac{1}{2} < s < 1$ between the fully delocalised regime ($s < 1/2$) and the localised regime ($s > 1$)
- No non-trivial fractal dimensions were observed. After rescaling by \sqrt{N} the eigenvector distributions become N -independent
- Main result: the eigenvector distributions can be extremely accurately fitted by the **generalised hyperbolic distribution**
- The investigation of the PLBM and UMM in the intermediate regime is of importance as they constitute a new class of random matrices potentially important for different applications
- One possible application is the explanation of deviations of recent experimental data of neutron widths from the Porter-Thomas distribution