

Random tridiagonal matrices, β -ensembles and random Schrödinger operators

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Random eigenvalue problems

- ▶ Generalized β -ensembles
(GOE, GUE, Wishart. . .)

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Method: scaling limit of random **tridiagonal matrices**

Classical Random Matrix Ensembles

$x_{k,\ell}$: i.i.d. sequence of standard normals (real/complex/quaternion)

$X(n, m) = [x_{k,\ell}]_{1 \leq k \leq n, 1 \leq \ell \leq m}$ is an $n \times m$ random matrix

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Laguerre ensemble: BB^* $B \sim X(n, m)$
(Wishart)

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Joint density of the eigenvalues:

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$$\frac{1}{Z_{\beta,n}} \Delta(\lambda)^\beta \exp \left[-\frac{\beta}{4} \sum_{k=1}^n \lambda_k^2 \right]$$

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These are the general β -ensembles.

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tridiagonal matrix representation for the β ensembles

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- ▶ Eigenvalues of A : Hermite-ensemble
- ▶ Eigenvalues of BB^T : Laguerre ensemble

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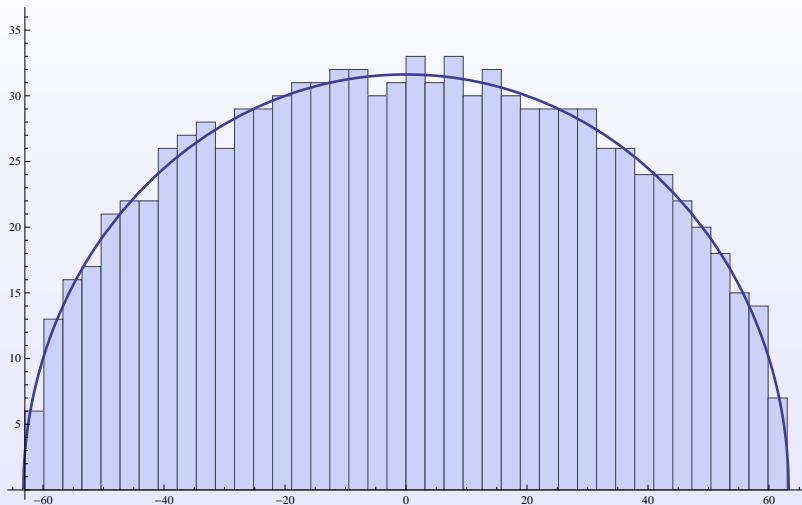
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Point process limit of the spectrum:

$$b_n(\Lambda_n - a_n) \Rightarrow ?$$

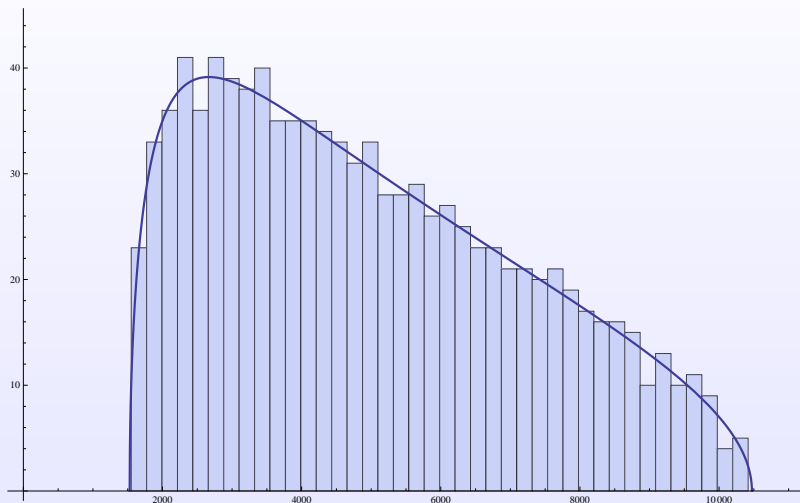
Limit of the empirical spectral measure

Histogram for a Hermite ensemble with $n = 1000$



Limit of the empirical spectral measure

Histogram for a Laguerre ensemble with $n = 1000$, $m = 5000$



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Thm: The properly normalized empirical spectral measure converges to a deterministic measure.

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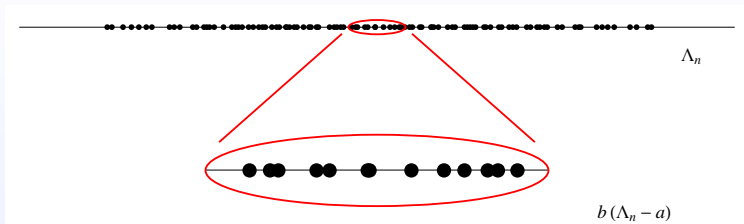
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Laguerre: $n/m \rightarrow y \in (0, 1]$, a, b are functions of y .

Local limit

We want to 'zoom' in to see a point process in the limit.

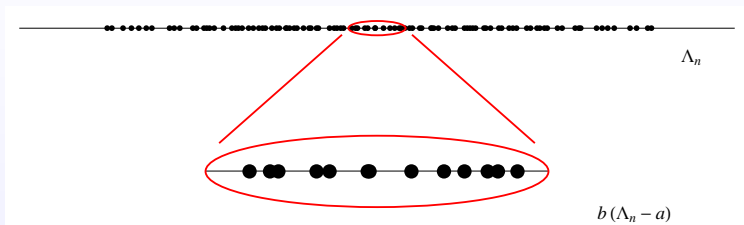


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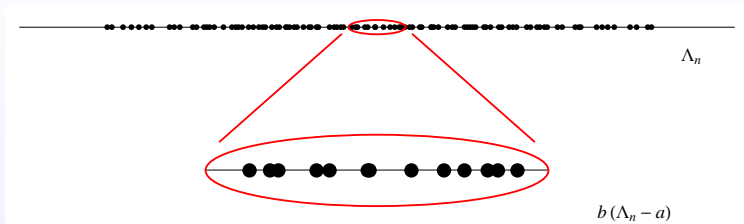
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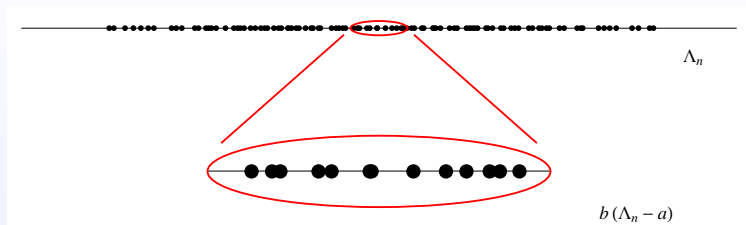
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Bulk/soft edge/hard edge: different scalings

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$\beta = 1, 2, 4$: known

Random Schrödinger operator

1D Schrödinger operator: $\mathcal{H} = \partial_x^2 + V$

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If we have a finite interval with Dirichlet boundary conditions \rightsquigarrow

$$M_n = \begin{bmatrix} V_1 & 1 & & & & \\ 1 & V_2 & 1 & & & \\ & \ddots & \ddots & & & \\ & & & 1 & V_{n-1} & 1 \\ & & & & 1 & V_n \end{bmatrix}$$

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$$f_{k-1} + V_k f_k + f_{k+1} = \lambda f_k, \quad k = 1, \dots, n$$

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$$\text{spectrum} = \Lambda_n = \{2 \cos(2\pi k / (n + 1)) : k = 1 \dots n\}$$

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Local limit (with noise): $n(\Lambda_n - \mu) \Rightarrow ?$

Eigenvalue equations for tridiagonal matrices

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Setting $r_k = u_{k+1}/u_k$:

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(Discrete version of Sturm-Liouville theory)

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If we can analyze the **SDE** then we can also understand the limiting point process.

Results

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Also: large deviation estimate for large gaps

The limiting process

The limit is described by the SDE

$$d\alpha_\lambda = \lambda \frac{\beta}{4} e^{-\beta/4 t} dt + \operatorname{Re} [(e^{-i\alpha_\lambda} - 1) d(B_1 + iB_2)], \quad \alpha_\lambda(0) = 0$$

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Killip-Stoiciu '06: similar equation for the limit of circular ensembles

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Thm(Kritchevski-V-Virág, '10):

Local limit of the spectrum (along an appropriate subsequence).

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- ▶ Eigenvalue repulsion ($\gg \beta$ -ensembles)
- ▶ CLT for number of points in a large interval

Outline of the proof

$$f_{k-1} + f_{k+1} - \frac{\sigma \xi_k}{\sqrt{n}} f_k = \left(\mu + \frac{\lambda}{n} \right) f_k$$

$$1 \leq k \leq n, \quad \xi_k \sim N(0, 1), \quad \mu \in (-2, 2) \setminus \{0\}$$

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$$T_k \text{ is close to } T = \begin{bmatrix} \mu & -1 \\ 1 & 0 \end{bmatrix} \Rightarrow \text{We should follow } Y_k = T^{-k} M_k!$$

Outline of the proof

$$M_k = T_k T_{k-1} \dots T_1$$

Evolution of $Y_k = T^{-k} M_k$:

$$Y_k - Y_{k-1} = T^{-k} T_k T^{k-1} Y_{k-1} - Y_{k-1}$$

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T can be diagonalized:

$$T = Z \begin{bmatrix} z & 0 \\ 0 & \bar{z} \end{bmatrix} Z^{-1}, \quad Z = \begin{bmatrix} z & \bar{z} \\ 1 & 1 \end{bmatrix}$$

$$z = \mu/2 + i\sqrt{1 - (\mu/2)^2}$$

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Explicit computation:

$$T^{-k} \begin{bmatrix} 0 & \frac{\lambda}{n} + \frac{\sigma \xi_k}{\sqrt{n}} \\ 0 & 0 \end{bmatrix} T^k = Z \frac{1}{i\sqrt{4 - \mu^2}} \left(\frac{\lambda}{n} + \frac{\sigma \xi_k}{\sqrt{n}} \right) \begin{bmatrix} 1 & z^{-2k} \\ -z^{2k} & -1 \end{bmatrix} Z^{-1}$$

Outline of the proof

SDE limit for $X_k = Z^{-1}Y_k$:

$$dX = i(\lambda dt - \sigma dB_1) \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} X + \frac{i\sigma}{\sqrt{2}} \begin{bmatrix} 1 & dB_2 + idB_3 \\ -dB_2 + idB_3 & 1 \end{bmatrix} X$$

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scaling limit of the counting function

Final comments

- ▶ V.-Virág: multidimensional version (long boxes)

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- ▶ V.-Virág: multidimensional version (long boxes)
- ▶ Proof for the β -ensembles is a bit more tricky (several regions)

THANK YOU!