

# Fluctuations of the characteristic polynomial of random Jacobi matrices

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joint work with R. Butez & O. Zeitouni

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# Gaussian Ensembles

Let  $X$  a GOE or GUE matrix ( $\beta = 1, 2$ ),

$$X \propto e^{-\frac{n\beta}{4}\text{tr}H^2} d\ell_n^{(\beta)}(H),$$

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$$\text{(Wigner)} \quad \mu_n = \frac{1}{n} \sum_{i=1}^n \delta_{\lambda_i} \xrightarrow{n \rightarrow +\infty} \mu_\sigma,$$

where  $\lambda_1, \dots, \lambda_n$  are the eigenvalues of  $X$  and

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How does  $\mu_n$  fluctuate around  $\mu_\sigma$ ?

## Central limit theorems

The linear statistics associated to  $f : \mathbb{R} \rightarrow \mathbb{R}$  is the observable

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(Guionnet-Zeitouni): If  $f$  is a  $L$ -Lipschitz function, for any  $t > 0$ ,

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For Lipschitz functions,

$$n \int f d(\mu_n - \mathbb{E}\mu_n) \text{ is tight.} \quad (1)$$

In sharp contrast with the i.i.d  $(X_i)_i$  case where for

$\hat{\mu}_n = \frac{1}{n} \sum_{i=1}^n \delta_{X_i}$ , and  $f$  bounded,

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(1) is the manifestation of cancellations happening due to the rigidity of the eigenvalues.

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(Johansson, 98'): For a smooth function  $f$ ,

$$n \int f d(\mu_n - \mathbb{E}\mu_n) \xrightarrow[n \rightarrow +\infty]{} \mathcal{N}(0, \sigma(f)^2),$$

Variance structure:

$$\sigma(f)^2 = \frac{1}{4\pi^2} \int_{-2}^2 \frac{f(t)}{\sqrt{4-t^2}} \left( \int_{-2}^2 \frac{f'(s)\sqrt{4-s^2}}{t-s} ds \right) dt.$$

Expression of the centering:

$$n \int f d\mathbb{E}\mu_n = n \int f d\mu_\sigma + \left( \frac{2}{\beta} - 1 \right) \int f d\nu_\beta + o(1),$$

with  $\nu_\beta$  a signed measure supported on  $[-2, 2]$ .

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For  $z \in (-2, 2)$ ,  $f = \log |z - \cdot|$ , what are the fluctuations of

$$n \int f d\mu_n = \log |\det(zI_n - X)|, \quad X \text{ GUE matrix?}$$

## Variance of the log determinant: heuristics

As a matter of fact

$$\text{Var}(\log |\phi_n(z)|) \asymp \log n.$$



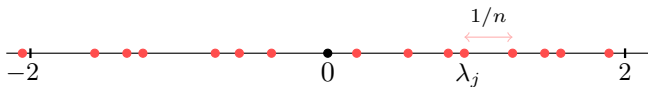
## Variance of the log determinant: heuristics

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$$\text{Var}(\log |\phi_n(\mathbf{0})|) \asymp \log n.$$

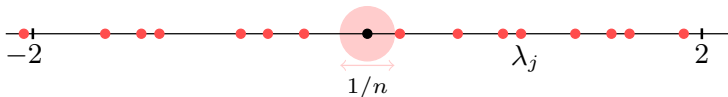
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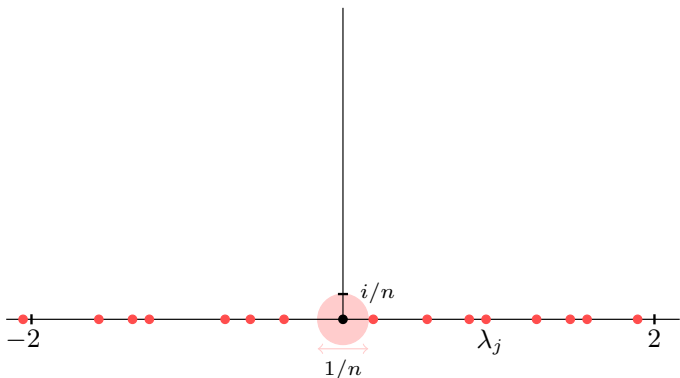
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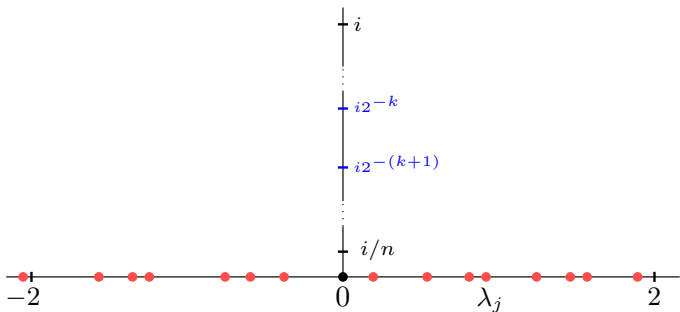
$$\log |\phi_n(0)| = \sum_{i=1}^n \log |\lambda_i| = \log |\phi_n(i/n)| + O(1) \text{ (variance-wise).}$$



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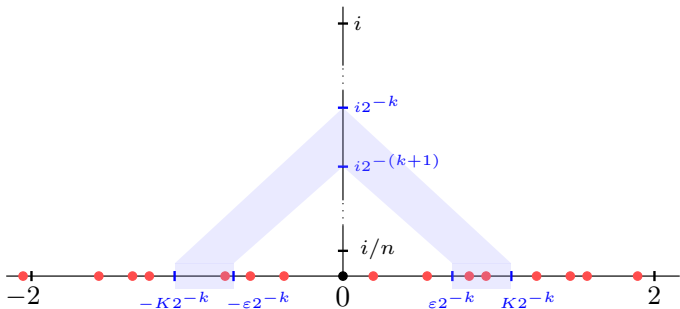
## Variance of the log determinant: heuristics

$$\log |\phi_n(i/n)| = \sum_{k=0}^N (\log |\phi_n(i2^{-(k+1)})| - \log |\phi_n(i2^{-k})|) + \log |\phi_n(i)|.$$



## Variance of the log determinant: heuristics

$$\log |\phi_n(i2^{-(k+1)})| - \log |\phi(i2^{-k})| = \int f_k d\mu_n, \quad \sigma(f_k)^2 \asymp 1.$$



## Variance of the log determinant: heuristics

With  $N \asymp \log n$ ,

$$\log |\phi_n(i/n)| = \log |\phi_n(i)| + \sum_{k=0}^{N-1} (\log |\phi_n(i2^{-(k+1)})| - \log |\phi_n(i2^{-k})|)$$

For a fixed  $k$ ,

$$\text{(Johansson)} \quad \text{Var} \left( \log \left| \frac{\phi_n(i2^{-(k+1)})}{\phi_n(i2^{-k})} \right| \right) \sim \sigma(f_k)^2 \asymp 1.$$

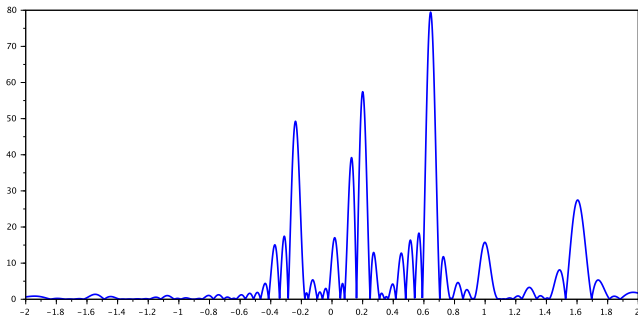
One expects the cross terms to vanish, which gives

$$\text{Var} \left( \log |\phi_n(i/n)| \right) \asymp \log n,$$

and thus  $\text{Var}(\log |\phi_n(0)|) \asymp \log n$ .

# The field of the characteristic polynomial

Let  $X$  a GUE matrix and  $\phi_n(z) = \det(zI_n - X)$ .

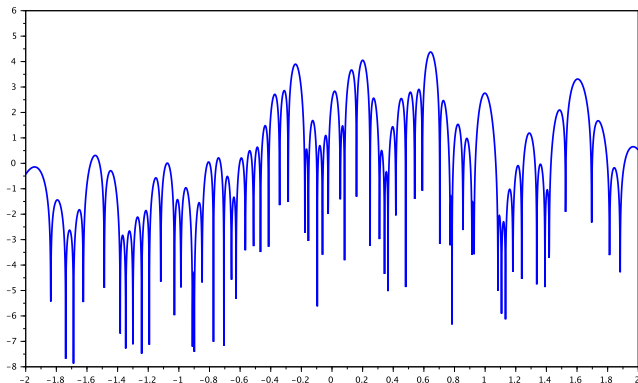


$$z \in [-2, 2] \mapsto |\phi_n(z)| e^{-\mathbb{E} \log |\phi_n(z)|}, n = 60$$



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# The maximum modulus of the characteristic polynomial

Conjecture (Fydororov- Simm, 15'):

$$\max_{z \in [-2, 2]} (\log |\phi_n(z)| - \mathbb{E} \log |\phi_n(z)|) - \log n + \frac{3}{4} \log \log n \xrightarrow{n \rightarrow +\infty} \xi,$$

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- The  $\frac{3}{4}$  is the manifestation of the correlation structure:  $\log |\phi_n(z)|$  is a log-correlated field:

If  $X_1, \dots, X_n$  are i.i.d  $\mathcal{N}(0, \frac{1}{2} \log n)$ , then w.h.p

$$\max_{1 \leq i \leq n} X_i = \log n - \frac{1}{4} \log \log n + O(1),$$

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- (Chahabi-Madaule-Najnudel, 18') proved the **tightness** for the Circular  $\beta$  Ensemble.

## Gaussian $\beta$ -Ensembles

Let  $\beta > 0$ . The probability measure  $\mathbb{P}_{n,\beta}$  on  $\mathbb{R}^n$ ,

$$d\mathbb{P}_{n,\beta} = Z_{n,\beta}^{-1} \prod_{i < j} |\lambda_i - \lambda_j|^\beta e^{-\frac{n\beta}{4} \sum_{i=1}^n \lambda_i^2} \prod_{i=1}^n d\lambda_i.$$

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is the eigenvalues distribution of the

- GOE ( $\beta = 1$ ), GUE ( $\beta = 2$ ), GSE ( $\beta = 4$ ).

















# Fluctuations of the characteristic polynomial

(A.-Butez-Zeitouni, 20):

Let  $z \in (-2, 2) \setminus \{0\}$ . Define  $\phi_n = \log |\det(zI_n - J_n/\sqrt{n})|$ .

$$\frac{\log |\phi_n(z)| - \log C_n(z)}{\sqrt{v \log n/2}} \xrightarrow{n \rightarrow +\infty} \mathcal{N}(0, 1).$$

where

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- The case  $z = 0$  exhibits special symmetries: see (Tao, Vu 11').
- (Bourgade-Mody-Pain): Multi-dimensional CLT for  $\beta$ -ensembles with generic potential.
- (Lambert-Paquette): CLT at the edge for Gaussian  $\beta$ -ensembles.

## The three-terms recursion

The sequence  $p_k = \det(z\sqrt{n}I_k - J_k)$  satisfies the recursion

$$\forall k \in \{1, \dots, n\}, p_{k+1} = (z\sqrt{n} - b_{k+1})p_k - \underbrace{a_k^2}_{\simeq k} p_{k-1},$$

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Equivalently, if  $X_k = (\psi_k, \psi_{k-1})$ ,

$$X_{k+1} = T_k X_k, \quad T_k = \begin{pmatrix} z\sqrt{\frac{n}{k}} - \frac{b_k}{\sqrt{k}} & -\frac{a_{k-1}^2}{k} \\ 1 & 0 \end{pmatrix}.$$

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The dynamics will depend on the **spectrum** of the matrices  $T_k$ .



## Spectrum of the transfer matrix

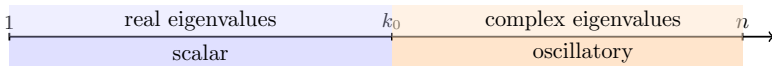
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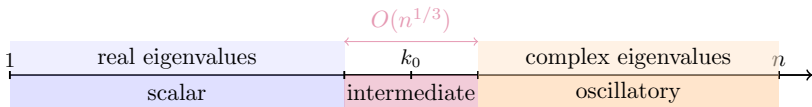
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## Scalar regime

For  $k \leq k_0$ , the recursion is

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We take the **normalization** such that  $\psi \equiv 1$  is solution of the **deterministic recursion**:

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- ▶ We show that  $\tilde{\psi}$  remains close to the direction  $(1, 1)$

## Scalar regime

For  $k \leq k_0$ , the recursion is

$$\forall k \in \{1, \dots, n\}, \psi_{k+1} = \left( z \sqrt{\frac{n}{k}} - \frac{b_{k+1}}{\sqrt{k}} \right) \psi_k - \frac{a_k^2}{k} \psi_{k-1}.$$

We take the **normalization** such that  $\psi \equiv 1$  is solution of the **deterministic recursion**:

$$\tilde{\psi}_k = \left( \prod_{i=1}^k \lambda_i^{-1} \right) \psi_k,$$

where  $\lambda_k$  is the **top eigenvalue** of  $\mathbb{E}T_k$ .

This has the effect of changing the **top eigenvalue** of  $\mathbb{E}T_k$  to 1 and **top eigenvector** to  $(1, 1)$ .

- ▶ We show that  $\tilde{\psi}$  remains close to the direction  $(1, 1)$
- ▶ We **linearize** the equation around the **stable solution**  $\tilde{\psi} \equiv 1$ , leading to a **recursion of order 1**.



## Oscillatory regime

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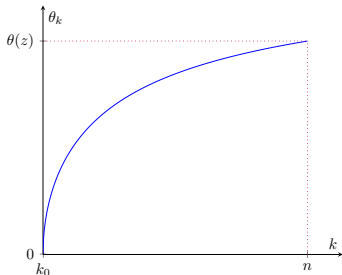
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$$\begin{aligned} Y_{k+1} &= (R_{\theta_{k+1}} + W_{k+1}) Q_{k+1} Q_k^{-1} Y_k \\ &\simeq (R_{\theta_{k+1}} + \underbrace{W_{k+1}}_{\text{noise}} + \underbrace{\Delta_{k+1}}_{\text{drift}}) Y_k. \end{aligned}$$

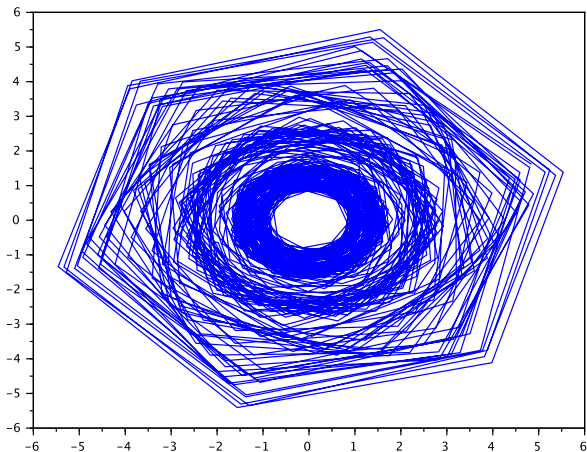


$$\mathbb{E}\|W_k\|^2 \asymp \frac{1}{k - k_0}$$

$$\|\Delta_k\| \asymp \frac{1}{k - k_0}$$

$R_{\theta_k}$  faster and faster

## Oscillatory regime



A trajectory of  $Y_k$ ,  $k_0 \leq k \leq n$ ,  $n = 2000$

## Analysis of the recursion on a section

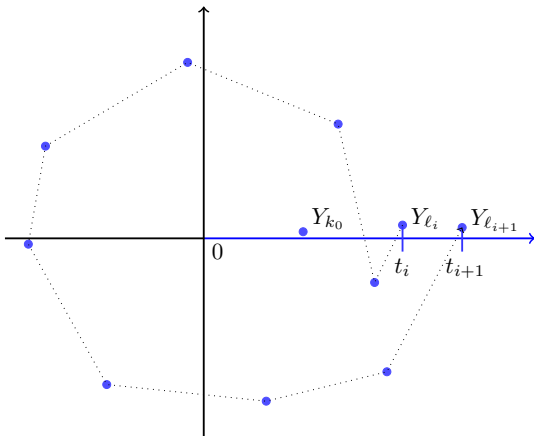
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## Analysis of the recursion on a section

Output from the scalar regime: Initial condition of the oscillatory regime  $Y_{k_0}$  is close to the direction  $(1, 0)$ .

We define “return times” of the recursion in the direction  $(1, 0)$

$$k_0 \leq l_1 \leq \dots \leq l_m.$$





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This allows us to represent  $\log \|Y_n\|$  as a drifted martingale, and the CLT follows.

# Fluctuations of the characteristic polynomial

(A.-Butez-Zeitouni, 20):

Let  $z \in (-2, 2) \setminus \{0\}$ . Define  $\phi_n = \log |\det(zI_n - J_n/\sqrt{n})|$ .

$$\frac{\log |\phi_n(z)| - \log C_n(z)}{\sqrt{v \log n/2}} \xrightarrow{n \rightarrow +\infty} \mathcal{N}(0, 1).$$

where

$$\log C_n(z) = n \underbrace{\left( \frac{z^2}{4} - \frac{1}{2} \right)}_{\int \log(z-x) d\mu_\sigma(x)} + a_v \log n + O(1).$$

- The case  $z = 0$  exhibits special symmetries: see (Tao, Vu 11').
- (Bourgade-Mody-Pain): Multi-dimensional CLT for  $\beta$ -ensembles with generic potential.
- (Lambert-Paquette): CLT at the edge for Gaussian  $\beta$ -ensembles.

## Further questions

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- Multi-dimensional CLT (work in progress)



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- Multi-dimensional CLT (work in progress)
- Can one strengthen the CLT to get a control on exponential moments? → key to obtaining estimate on the maximum

$$\max_{z \in [-2, 2]} (\log |\phi_n(z)| - \mathbb{E} \log |\phi_n(z)|).$$

Thank you !