

# Law of fractional logarithm for random matrices

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Log-gases in Caeli Australi

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## Classical laws for iid sum

Consider a sequence of iid r.v.s  $X_1, X_2, \dots$  with mean 0 and variance 1. Let

$$S_n = \sum_{i=1}^n X_i$$

**SLLN**

$$\frac{S_n}{n} \xrightarrow{\text{a.s.}} 0$$

**CLT**

$$\frac{S_n}{\sqrt{n}} \Rightarrow N(0, 1)$$

**LIL**

$$\limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = 1, \quad \liminf_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = -1, \quad \text{a.s.}$$

## Laws for extreme random matrix eigenvalue

Consider a **doubly infinite array** of independent (up to symmetry  $x_{ij} = \bar{x}_{ji}$ ) real ( $\beta = 1$ ) or complex ( $\beta = 2$ ) r.v.s  $X = (x_{ij})_{i,j \in \mathbb{N}}$ : (i) mean 0, (ii)  $\mathbb{E}x_{ij}^2 = 1 + \delta_{ij}$  ( $\beta = 1$ ) and  $\mathbb{E}|x_{ij}|^2 = 1, \mathbb{E}x_{ij}^2 = \delta_{ij}$  ( $\beta = 2$ ), (iii) existence of all moments.

**Wigner Minor Process** Let  $X^{(N)}$  be the  **$N \times N$  upper left corner** of  $X$ . Denote

$$H^{(N)} = \frac{1}{\sqrt{N}} X^{(N)}, \quad N = 1, 2, \dots$$

and let  $\tilde{\lambda}_i^{(N)}$  be the  $i$ -th largest eigenvalue of  $H^{(N)}$ .

**Bai-Yin Law** [Bai-Yin '88]

$$\tilde{\lambda}_1^{(N)} \xrightarrow{\text{a.s.}} 2.$$

**Tracy-Widom Law** [Forrester '93] [Tracy-Widom '94 '96] etc.

$$\lambda_1^{(N)} \equiv N^{2/3}(\tilde{\lambda}_1^{(N)} - 2) \implies \text{TW}_\beta.$$

**Question**[Kalai '13] Any logarithmic type law?

## Paquette-Zeitouni's LFL for GUE

**Theorem**(Law of Fractional Logarithm) [Paquette-Zeitouni '17] For **GUE**

$$\limsup \frac{\lambda_1^{(N)}}{(\log N)^{2/3}} = \left(\frac{1}{4}\right)^{2/3}, \quad \text{a.s.}$$

$$-c_1 \leq \liminf \frac{\lambda_1^{(N)}}{(\log N)^{1/3}} \leq -c_2, \quad \text{a.s.}$$

**Conjecture 1:**  $c_1 = c_2 = 4^{1/3}$

**Conjecture 2:** Universality (other **symmetry** class and entry **distribution**)?

**Remark:** The proof in [Paquette-Zeitouni '17] heavily relies on the **determinantal structure** of GUE minor process [Johansson-Nordenstam '06], [Forrester-Nagao '08], etc.

## BBBK's lower limit for GUE

**Theorem**[Baslingker-Basu-Bhattacharjee-Krishnapur '24b] For **GUE**

$$\liminf \frac{\lambda_1^{(N)}}{(\log N)^{1/3}} = -4^{1/3}, \quad \text{a.s.}$$

**Remark** Two key inputs in [BBBK '24b]

- (i) A sharp **lower tail estimate** of  $\lambda_1^{(N)}$  for all  $\beta$ -ensembles from [Baslingker-Basu-Bhattacharjee-Krishnapur '24a]
- (ii) A **decorrelation estimate** via passage times in Brownian last passage percolation [Baryshnikov '01]

## LFL for Wigner matrices

**Theorem** [B.-Cipolloni-Erdős-Henheik-Kolupaiev '25a] For **Wigner** minor process in either symmetric class,  $\beta = 1$  for real symmetric and  $\beta = 2$  for complex Hermitian, we have almost surely

$$\limsup_{N \rightarrow \infty} \frac{\lambda_1^{(N)}}{(\log N)^{2/3}} = \left(\frac{1}{2\beta}\right)^{2/3}, \quad \liminf_{N \rightarrow \infty} \frac{\lambda_1^{(N)}}{(\log N)^{1/3}} = -\left(\frac{8}{\beta}\right)^{1/3}.$$

**Remark:** Our overall proof strategy is not via a comparison as there is no GOE result, although for the one point tail probability estimate we do rely on comparison with [BBBK '24a].

## All Possible Limit Points

**Corollary** [B.-Cipolloni-Erdős-Henheik-Kolupaiev '25a] For Wigner minor process in either symmetric class,  $\beta = 1$  for real symmetric and  $\beta = 2$  for complex Hermitian, we have almost surely

$$\bigcap_{m=1}^{\infty} \overline{\left\{ \lambda_1^{(N)} / (\log N)^{2/3} : N \geq m \right\}} = \left[ 0, \left( \frac{1}{2\beta} \right)^{2/3} \right]$$
$$\bigcap_{m=1}^{\infty} \overline{\left\{ \lambda_1^{(N)} / (\log N)^{1/3} : N \geq m \right\}} = \left[ - \left( \frac{8}{\beta} \right)^{1/3}, \infty \right]$$

where  $\overline{\mathcal{E}}$  is the closure of a set  $\mathcal{E} \subset \mathbb{R}$ .

## Key steps for classical LIL

For some sequence  $a_n \rightarrow \infty$  but  $a_n/\sqrt{n} \rightarrow 0$ , we set the event

$$E_n := \left\{ S_n \geq a_n \sqrt{n} \right\}$$

1: One point small deviation

$$\mathbb{P}(E_n) = \exp(-a_n^2(1 + o(1))/2)$$

2: Decorrelation

$$\mathbb{P}(E_n \cap E_{n+m}) = \mathbb{P}(E_n)\mathbb{P}(E_{n+m})(1 + o(1)), \quad \text{if } m \gg n$$

3: Correlation

$$E_n \approx E_{n+m}, \quad \text{if } m \ll n$$



## Small deviation of extreme eigenvalue

**Proposition** Consider general Wigner matrices. For any (big)  $K > 0$  and (small)  $\varepsilon > 0$ , there exist  $C_1 \equiv C_1(K, \varepsilon)$  and  $C_2 \equiv C_2(K, \varepsilon)$ , s.t.

(i) [Right tail] For any  $1 \leq x \leq K(\log N)^{2/3}$ ,

$$C_1^{-1} \exp\left(-\frac{2\beta}{3}(1+\varepsilon)x^{3/2}\right) \leq \mathbb{P}(\lambda_1^{(N)} \geq x) \leq C_1 \exp\left(-\frac{2\beta}{3}(1-\varepsilon)x^{3/2}\right)$$

(ii) [Left tail] For any  $1 \leq x \leq K(\log N)^{1/3}$ ,

$$C_2^{-1} \exp\left(-\frac{\beta}{24}(1+\varepsilon)x^3\right) \leq \mathbb{P}(\lambda_1^{(N)} \leq -x) \leq C_2 \exp\left(-\frac{\beta}{24}(1-\varepsilon)x^3\right)$$

**Remark** Part of the results are from [Erdős-Xu '23]. The rest is obtained via a comparison with [BBBK '24a], using the Green function comparison approach in [Erdős-Xu '23] for small deviation regime. Also see [Aubrun '05], [Ledoux-Rider '10], [Paquette-Zeitouni '14] etc. for earlier results.

## Correlation-Decorrelation transition

**Theorem**[B.-Cipolloni-Erdős-Henheik-Kolupaiev '25b] Consider general real ( $\beta = 1$ ) or complex ( $\beta = 2$ ) Wigner minor process.

(i) For  $1 \ll k \leq N^{2/3-\varepsilon}$

$$\frac{N^{1/3}(\lambda_1^{(N)} - \lambda_1^{(N-k)})}{\sqrt{k}} \sim \mathcal{N}(0, 2/\beta),$$

(ii) For  $N^{2/3+\varepsilon} \leq k \leq N^{1-\varepsilon}$ , and any smooth compactly supported test-functions  $F, G$

$$\mathbb{E}[F(\lambda_1^{(N)})G(\lambda_1^{(N-k)})] - \mathbb{E}[F(\lambda_1^{(N)})]\mathbb{E}[G(\lambda_1^{(N-k)})] \rightarrow 0.$$

**Remark 1:** The threshold  $N^{2/3}$  is indicated by [Forrestor-Nagao '11].

**Remark 2:** For LFL, we need above transition for small/moderate deviation regime.

## Upper bound for lim sup: Subsequence

**Aim** Prove

$$\limsup_{N \rightarrow \infty} \frac{\lambda_1^{(N)}}{(\log N)^{2/3}} \leq c \quad a.s., \quad \forall c > \left(\frac{1}{2\beta}\right)^{2/3}$$
$$\iff \mathbb{P}\left(\frac{\lambda_1^{(N)}}{(\log N)^{2/3}} > c \quad i.o.\right) = 0$$

**Subsequence:** Let  $\alpha < 3$  and  $N_k = \lfloor k^\alpha \rfloor$  s.t.

$$N_{k+1} - N_k \ll N_k^{2/3}.$$

Since

$$\sum_k \mathbb{P}\left[\lambda_1^{(N_k)} \geq c(\log N_k)^{2/3}\right] \lesssim \sum_k k^{-\alpha \frac{2\beta}{3}(1-\varepsilon)c^{3/2}} < \infty$$

By BC I

$$\limsup_{k \rightarrow \infty} \frac{\lambda_1^{(N_k)}}{(\log N_k)^{2/3}} \leq c.$$

## Upper bound for lim sup: Full sequence

**Lemma**[Correlation estimate] For any  $\delta > 0$

$$\sum_k \mathbb{P}(\mathcal{E}_k(\delta)) < \infty$$

where

$$\mathcal{E}_k(\delta) = \left\{ \exists n \in [N_{k-1} + 1, N_k] : \lambda_1^{(n)} \geq (c + \delta)(\log n)^{2/3} \right\} \\ \cap \left\{ \lambda_1^{(N_k)} \leq c(\log N_k)^{2/3} \right\}$$

Consequently

$$\limsup_{N \rightarrow \infty} \frac{\lambda_1^{(N)}}{(\log N)^{2/3}} \leq c + \delta.$$

**Remark:** Essentially, we need bound  $\lambda_1^{(N_k)} - \lambda_1^{(n)}$  from below, uniformly in  $n$ .

## Correlation: Key equation

Write

$$H^{(n)} = \begin{bmatrix} \sqrt{\frac{n-1}{n}} H^{(n-1)} & \mathbf{a}^{(n)} \\ (\mathbf{a}^{(n)})^* & h_{nn}^{(n)} \end{bmatrix}$$

### Key Equation

$$\tilde{\lambda}_1^{(n)} = h_{nn}^{(n)} + \frac{1}{n} \sum_{\alpha=1}^{n-1} \frac{|\xi_\alpha^{(n)}|^2}{\tilde{\lambda}_1^{(n)} - \sqrt{\frac{n-1}{n}} \tilde{\lambda}_\alpha^{(n-1)}}, \quad \xi_\alpha^{(n)} = \sqrt{n} (\mathbf{w}_\alpha^{(n-1)})^* \mathbf{a}^{(n)}$$

We rewrite it as

$$\underbrace{\tilde{\lambda}_1^{(n)}}_{\approx 2} = \underbrace{h_{nn}^{(n)}}_{\approx 0} + \frac{1}{n} \frac{|\xi_1^{(n)}|^2}{\tilde{\lambda}_1^{(n)} - \sqrt{\frac{n-1}{n}} \tilde{\lambda}_1^{(n-1)}} + \underbrace{\frac{1}{n} \sum_{\alpha=2}^{n^\varepsilon} \frac{|\xi_\alpha^{(n)}|^2}{\tilde{\lambda}_1^{(n)} - \sqrt{\frac{n-1}{n}} \tilde{\lambda}_\alpha^{(n-1)}}}_{>0} + \underbrace{\frac{1}{n} \sum_{\alpha > n^\varepsilon} \frac{|\xi_\alpha^{(n)}|^2}{\tilde{\lambda}_1^{(n)} - \sqrt{\frac{n-1}{n}} \tilde{\lambda}_\alpha^{(n-1)}}}_{\approx 1}$$

## Correlation: Martingale Approach

By local law and rigidity [Erdős-Yau-Yin '12]

$$\tilde{\lambda}_1^{(n)} - \tilde{\lambda}_1^{(n-1)} \geq \frac{1}{n} \left( |\xi_1^{(n)}|^2 - 1 \right) + O_{\prec}(n^{-4/3})$$

Sum up for any  $n \in [N_{k-1} + 1, N_k]$ ,

$$\begin{aligned} \tilde{\lambda}_1^{(N_k)} - \tilde{\lambda}_1^{(n)} &\geq \sum_{\ell=n+1}^{N_k} \frac{1}{\ell} (|\xi_1^{(\ell)}|^2 - 1) + O_{\prec}(N_k^{-2/3}) \\ &= \sum_{\ell=N_{k-1}+1}^{N_k} \frac{1}{\ell} (|\xi_1^{(\ell)}|^2 - 1) - \sum_{\ell=N_{k-1}+1}^n \frac{1}{\ell} (|\xi_1^{(\ell)}|^2 - 1) + O_{\prec}(N_k^{-2/3}) \end{aligned}$$

Conclude the proof by **maximal inequality of martingale**.

## Lower bound for lim sup: Decorrelation

**Aim** Prove

$$\limsup_{N \rightarrow \infty} \frac{\lambda_1^{(N)}}{(\log N)^{2/3}} \geq c \quad a.s., \quad \forall c < \left(\frac{1}{2\beta}\right)^{2/3}$$
$$\iff \mathbb{P}\left(\frac{\lambda_1^{(N)}}{(\log N)^{2/3}} \geq c \quad i.o.\right) = 1$$

**Proposition**[Decorrelation estimate] For any  $N_1, N_2 \in \mathbb{N}$  and any  $x_1, x_2 \in [K^{-1}, K]$  with any large constant  $K > 0$ , we set the tail events

$$\mathcal{F}_\ell(x_\ell) := \left\{ \lambda_1^{(N_\ell)} \geq x_\ell (\log N_\ell)^{2/3} \right\}, \quad \ell = 1, 2.$$

If  $N_2^{\frac{2}{3}+\varepsilon} \leq N_2 - N_1 \leq N_2^{1-\varepsilon}$  we have

$$\mathbb{P}(\mathcal{F}_1(x_1) \cap \mathcal{F}_2(x_2)) = \mathbb{P}(\mathcal{F}_1(x_1))\mathbb{P}(\mathcal{F}_2(x_2))(1 + O(N_2^{-\delta}))$$

**Remark** In the spirit of BC II, decorrelation allows one to prove the lower bound for a sufficiently independent subsequence.

## Decorrelation: a dynamic approach

We rely on the dynamic approach [Cipolloni-Erdős-Schröder '23].

### Dynamic of matrices

$$dH_t^{(n)} = \frac{dB_t^{(n)}}{\sqrt{n}}, \quad H_0^{(n)} = H^{(n)}, \quad n = N, N - k, \quad k \gg N^{2/3}$$

### Dynamic of minor process

$$d\tilde{\lambda}_i^{(n)} = \frac{db_i^{(n)}(t)}{\sqrt{\beta n}} + \frac{1}{N} \sum_{j \neq i} \frac{1}{\tilde{\lambda}_j^{(n)}(t) - \tilde{\lambda}_i^{(n)}(t)} dt, \quad n = N, N - k$$

Here  $\{b_i^{(N)}(t)\}$  and  $\{b_i^{(N-k)}(t)\}$  are two collections of independent standard BM, but these two collections are dependent

$$d[b_i^{(N)}(t), b_j^{(N-k)}(t)] = \left| \langle \mathbf{w}_i(H^{(N)}(t)), \mathbf{w}_j(H^{(N-k)}(t)) \rangle \right|^2 dt =: \Theta_{ij}^{(N, N-k)}(t) dt.$$

**Heuristic:**  $\tilde{\lambda}_1^{(N)}(t)$  and  $\tilde{\lambda}_1^{(N-k)}(t)$  will be indep. after sufficiently long time ( $t \gtrsim N^{-1/3+\varepsilon}$ ) if  $\Theta_{ij}^{(N, N-k)}(t)$  is small. For the indep. of the  $\tilde{\lambda}_1^{(N)}(0)$  and  $\tilde{\lambda}_1^{(N-k)}(0)$ , we do a Green function comparison with  $\tilde{\lambda}_1^{(N)}(t)$  and  $\tilde{\lambda}_1^{(N-k)}(t)$ .



# Eigenvector overlap

## Green function

$$G^{(n)}(z) = (H^{(n)} - z)^{-1}.$$

**Overlap bound** For  $z_a = E_a + i\eta_a$ , with  $\eta_a \sim N^{-2/3+\varepsilon}$

$$\begin{aligned} \frac{1}{N} \text{Tr} \Im G^{(N)}(z_1) \Im G^{(N-k)}(z_2) \\ \geq \frac{1}{N} \frac{\eta_1}{(\tilde{\lambda}_i^{(N)} - E_1)^2 + \eta_1^2} \frac{\eta_2}{(\tilde{\lambda}_j^{(N-k)} - E_2)^2 + \eta_2^2} \Theta_{ij}^{(N, N-k)} \end{aligned}$$

The estimate of the LHS is done by two-G local law [Cipolloni-Erdős-Schröder '22].

**THANK YOU!**

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**HAPPY BIRTHDAY, PETER!**