



Asymptotic expansion of the hard-to-soft edge transition

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Introduction

The Wishart-Laguerre ensemble

The Wishart-Laguerre ensemble \mathbf{X} is a complex random matrix defined by

$$\mathbf{X} = (X_1 + iX_2)(X_1 + iX_2)^*,$$

where X_1 and X_2 are two $n \times (n + \nu)$, $\nu \geq 0$, random matrices, whose element is chosen to be an independent normal random variable.

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Question: Where are the eigenvalues of \mathbf{X} ?

Marchenko-Pastur law

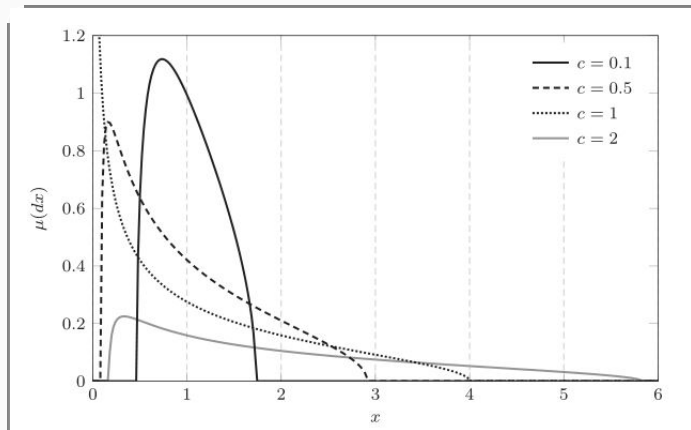
As $n \rightarrow \infty$, it is well known that the empirical spectral measure of \mathbf{X} converges weakly to the Marchenko-Pastur (MP) distribution μ given by

$$\mu(dx) = \frac{\sqrt{(x-x_-)(x_+ - x)}}{2\pi cx} dx, \quad x \in (x_-, x_+),$$

where $c = \lim n/(n + \nu) \in (0, 1]$ and $x_{\pm} = (1 \pm \sqrt{c})^2$.

[Marchenko-Pastur, '67]

Marchenko-Pastur law



Marchenko-Pastur distribution for different values of c . Source: [\[Couillet-Liao, '22\]](#)

Determinantal point process

Joint probability density function for the eigenvalues of \mathbf{X} :

$$\frac{1}{C_{n,\nu}} \prod_{l=1}^n x_l^\nu e^{-x_l} \prod_{1 \leq j < k \leq n} (x_j - x_k)^2 = \det [K_n^\nu(x_j, x_k)]_{j,k=1,\dots,n}$$

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This is a **determinantal point process** (DPP) with the correlation kernel $K_n^\nu(x, y)$ expressed in terms of classical Laguerre polynomials. All the information of DPP is contained in the correlation kernel.

Microscopic limits of the correlation kernel

Microscopic limits \Rightarrow local fluctuation

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The local limits and the associated DPPs are **universal** in RMT and beyond.

Ulam's problem

Let S_n be the set of all permutations on $\{1, 2, \dots, n\}$. Given $\sigma \in S_n$,

$L_n(\sigma) :=$ length of longest increasing subsequence in σ ,

i.e., the maximum of all k such that $1 \leq i_1 < i_2 < \dots < i_k \leq n$ with $\sigma(i_1) < \sigma(i_2) < \dots < \sigma(i_k)$.

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Example

Let $\sigma = 4635217$. Since both 457 and 357 are the longest increasing subsequences of length 3, it follows that

$$L_7(\sigma) = 3.$$

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- $\mathbb{E}(L_n) \sim c\sqrt{n}$

[Hammersley, '72]

- $\mathbb{E}(L_n) \sim 2\sqrt{n}$

[Vershik-Kerov, '77; Logan-Shepp, '77]

- $\lim_{n \rightarrow \infty} \mathbb{P} \left(\frac{L_n - 2\sqrt{n}}{n^{1/6}} \leq t \right) = F(t)$, where $F(t)$ is the Tracy-Widom distribution.

[Baik-Deift-Johansson, '99]

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The connections with other problems such as last passage percolation and RMT can be found in the book of Romik (2015).

The Tracy-Widom distribution

Definition:

$$F(t) = \det(I - \mathcal{K}_{\text{Ai}}),$$

where \mathcal{K}_{Ai} is the integral operator acting on $L^2(t, \infty)$ with the Airy kernel

$$K^{\text{Ai}}(x, y) = \frac{\text{Ai}(x)\text{Ai}'(y) - \text{Ai}'(x)\text{Ai}(y)}{x - y},$$

i.e.,

$$F(t) = 1 + \sum_{n=1}^{\infty} \frac{(-1)^n}{n!} \int_t^{\infty} \cdots \int_t^{\infty} \det(K^{\text{Ai}}(\xi_i, \xi_j))_{i,j=1}^n d\xi_1 \cdots d\xi_n.$$

It describes the limiting distributions of extreme eigenvalues for many random matrix ensembles.

Hammersley's Poissonization of L_n

A key role is played by the following Poissonization of L_n :

$$\mathbb{P}(L_{N_r} \leq l) := P(r, l) = e^{-r} \sum_{n=0}^{\infty} \mathbb{P}(L_n \leq l) \frac{r^n}{n!},$$

where $N_r \in \{0, 1, 2, \dots\}$ is a random variable with a Poisson distribution of intensity $r > 0$.

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Combinatorial interpretation: the cumulative distribution of the length of longest up/right path from $(0, 0)$ to $(1, 1)$ with nodes chosen randomly according to N_r .

Hammersley's Poissonization of L_n

To answer Ulam's question, one needs to show

$$\lim_{r \rightarrow \infty} \mathbb{P} \left(\frac{L_{N_r} - 2\sqrt{r}}{r^{1/6}} \leq t \right) = F(t).$$

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This, together with Johansson's de-Poissonization lemma, will lead to the distribution of L_n .

[Johansson, '98]

Hammersley's Poissonization of L_n

There are several representations of $P(r, l)$:

- a Toeplitz determinant of modified Bessel functions,

[Gessel, '90]

- Fredholm determinants of various (discrete) integral operators,

[Baik-Deift-Johansson, '00; Borodin-Okounkov-Olshanski, '00; Johansson, '01; ...]

- a unitary group integral.

[Rains, '98]

Asymptotics of $P(r, l)$

The fluctuation of $P(r, l)$ around $2\sqrt{r}$ can be proved by

- the Toeplitz determinant representation + Riemann-Hilbert approach,

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Asymptotics of $P(r, l)$

The fluctuation of $P(r, l)$ around $2\sqrt{r}$ can be proved by

- the Toeplitz determinant representation + Riemann-Hilbert approach, [Baik-Deift-Johansson, '99]
- the Fredholm determinat representation

$$P(r, l) = E_2^{\text{hard}}(4r, l),$$

where

$$E_2^{\text{hard}}(s; \nu) := \det(I - K_\nu^{\text{Bes}}) \Big|_{L^2(0, s)}$$

with

$$K_\nu^{\text{Bes}}(x, y) := \frac{J_\nu(\sqrt{x})\sqrt{y}J'_\nu(\sqrt{y}) - \sqrt{x}J'_\nu(\sqrt{x})J_\nu(\sqrt{y})}{2(x - y)}$$

being the Bessel kernel.

[Borodin-Forrester, '03; Forrester-Hughes, '94]

Asymptotics of $P(r, l)$

- The Fredholm determinat representation

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being the Bessel kernel.

- The distribution of $P(r, l)$ for large r is equivalent to the hard-to-soft edge transition

$$\lim_{\nu \rightarrow \infty} E_2^{\text{hard}} \left(\left(\nu - t(\nu/2)^{1/3} \right)^2 ; \nu \right) = F(t).$$

Today's topic

Can we improve the distribution formula of L_n by establishing the first few finite-size correction terms or the asymptotic expansion? This is also known as **edgeworth expansions** in the literature.

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Relevant works:

- finite size corrections relating to distributions of the length of longest increasing subsequences,
[Forrester-Mays, '23; Bornemann, '23]
- finite size corrections relating to the Laguerre ensemble,
[Perret-Schehr, '16; Forrester-Trinh, '19]
- finite size corrections relating to the elliptic Ginibre matrices,
[Byun-Lee, '23]
- ...

Main results

An expansion from the Bessel kernel to the Airy kernel

Theorem [Yao-Z., arXiv:2309.06733]

Let

$$\hat{K}_\nu^{\text{Bes}}(x, y) := \sqrt{\phi'_\nu(x)\phi'_\nu(y)} K_\nu^{\text{Bes}}(\phi_\nu(x), \phi_\nu(y))$$

with $\phi_\nu(t) = \nu^2(1 - h_\nu t)^2$ and $h_\nu = 2^{-\frac{1}{3}}\nu^{-\frac{2}{3}}$ be the symmetrically transformed Bessel kernel. For any $m \in \mathbb{N}$, we have, as $h_\nu \rightarrow 0^+$,

$$\hat{K}_\nu^{\text{Bes}}(x, y) = K^{\text{Ai}}(x, y) + \sum_{j=1}^m K_j(x, y) h_\nu^j + h_\nu^{m+1} \cdot \mathcal{O}\left(e^{-(x+y)}\right),$$

uniformly valid for $t_0 \leq x, y < h_\nu^{-1}$ with t_0 being any fixed real number. Preserving uniformity, the expansion can be repeatedly differentiated w.r.t. the variable x and y .

An expansion from the Bessel kernel to the Airy kernel

Theorem [Yao-Z., arXiv:2309.06733]

Here, K^{Ai} is the Airy kernel and

$$K_j(x, y) = \sum_{\kappa, \lambda \in \{0, 1\}} p_{j, \kappa \lambda}(x, y) \text{Ai}^{(\kappa)}(x) \text{Ai}^{(\lambda)}(y)$$

with $p_{j, \kappa \lambda}(x, y)$ being polynomials in x and y . Moreover, we have

$$K_1(x, y) = \frac{1}{10} (-3(x^2 + xy + y^2) \text{Ai}(x) \text{Ai}(y) + 2(\text{Ai}(x) \text{Ai}'(y) + \text{Ai}'(x) \text{Ai}(y)) \\ + 3(x + y) \text{Ai}'(x) \text{Ai}'(y)).$$

The above theorem was stated by Bornemann (arXiv:2301.02022) under the condition that $0 \leq m \leq m_* = 100$, and it was conjectured therein to be true without such a restriction.

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One can lift the expansion of the kernel to the associated Fredholm determinants, as rigorously shown by Bornemann.

Asymptotic expansion of the hard-to-soft edge transition

Corollary [Yao-Z., arXiv:2309.06733]

With E_2^{hard} defined as before, we have, for any $m \in \mathbb{N}$, as $h_\nu \rightarrow 0^+$,

$$E_2^{\text{hard}}(\phi_\nu(t); \nu) = F(t) + \sum_{j=1}^m F_j(t) h_\nu^j + h_\nu^{m+1} \cdot \mathcal{O}\left(e^{-3t/2}\right),$$

uniformly valid for $t_0 \leq t < h_\nu^{-1}$ with t_0 being any fixed real number. Preserving uniformity, the expansion can be repeatedly differentiated w.r.t. the variable t . Here, F denotes the Tracy-Widom distribution and F_j are certain smooth functions.

It can be shown that

$$F_1(t) = \frac{3t^2}{10} F'(t) - \frac{1}{5} F''(t),$$

$$F_2(t) = \left(\frac{2}{175} + \frac{32t^2}{175} \right) F'(t) + \left(-\frac{16t}{175} + \frac{9t^4}{200} \right) F''(t) - \frac{3t^2}{50} F'''(t) + \frac{1}{50} F^{(4)}(t).$$

[Bornemann, '23]

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[Bornemann, '23]

The above expansion serves as a preparatory step in establishing the expansion of L_n .

About the proofs

The proof of Bornemann

Recall that

$$K_\nu^{\text{Bes}}(x, y) = \frac{J_\nu(\sqrt{x})\sqrt{y}J'_\nu(\sqrt{y}) - \sqrt{x}J'_\nu(\sqrt{x})J_\nu(\sqrt{y})}{2(x - y)}.$$

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It is natural to use Olver's expansion of Bessel functions of large order in the transition region:

$$J_\nu(\nu + \tau\nu^{1/3}) \sim \frac{2^{1/3}}{\nu^{1/3}} \text{Ai}(-2^{1/3}\tau) \sum_{k=0}^{\infty} \frac{A_k(\tau)}{\nu^{2k/3}} + \frac{2^{1/3}}{\nu^{1/3}} \text{Ai}'(-2^{1/3}\tau) \sum_{k=0}^{\infty} \frac{B_k(\tau)}{\nu^{2k/3}},$$

for $|\arg \nu| \leq \pi/2 - \delta$ and fixed τ . Here, $A_k(\tau)$ and $B_k(\tau)$ are certain rational polynomials of increasing degree.

[Olver, '52]

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[Olver, '52]

It is **difficult** to check the divisibility by $x - y$ for all the coefficients.

A Riemann-Hilbert characterization of the Bessel kernel

The Bessel kernel admits the following representation:

$$K_\nu^{\text{Bes}}(x, y) = \frac{1}{2\pi i(x-y)} \begin{pmatrix} -e^{-\frac{\pi i\nu}{2}} & e^{\frac{\pi i\nu}{2}} \end{pmatrix} \Psi_+(y)^{-1} \Psi_+(x) \begin{pmatrix} e^{\frac{\pi i\nu}{2}} \\ e^{-\frac{\pi i\nu}{2}} \end{pmatrix}.$$

where

$$\Psi(z) = \sqrt{\pi} e^{-\frac{\pi i}{4}} \begin{pmatrix} I_\nu \left((-z)^{\frac{1}{2}} \right) & -\frac{i}{\pi} K_\nu \left((-z)^{\frac{1}{2}} \right) \\ (-z)^{\frac{1}{2}} I'_\nu \left((-z)^{\frac{1}{2}} \right) & -\frac{i}{\pi} (-z)^{\frac{1}{2}} K'_\nu \left((-z)^{\frac{1}{2}} \right) \end{pmatrix} \begin{cases} \begin{pmatrix} 1 & 0 \\ -e^{-\pi i\nu} & 1 \end{pmatrix}, & \arg z \in (0, \frac{\pi}{3}), \\ I, & \arg z \in (\frac{\pi}{3}, \frac{5\pi}{3}), \\ \begin{pmatrix} 1 & 0 \\ e^{\pi i\nu} & 1 \end{pmatrix}, & \arg z \in (\frac{5\pi}{3}, 2\pi), \end{cases}$$

with I_ν and K_ν being the modified Bessel functions of order ν .

The matrix-valued function Ψ is solves a Riemann-Hilbert problem uniquely.

- Asymptotic analysis of Ψ for large $\nu \longrightarrow$ kernel expansion.
- This direct approach provides a better understanding of the coefficients in the expansion.

Ideas of the proof

Let $\text{Ai}(x)$ be the classical Airy function. It is readily seen from the differential equation

$$\frac{d^2 \text{Ai}(z)}{dz^2} = z \text{Ai}(z)$$

that

$$\text{Ai}^{(m)}(x) = P_m(x) \text{Ai}(x) + Q_m(x) \text{Ai}'(x),$$

where $P_m(x)$ and $Q_m(x)$ are polynomials with $P_0(x) = 1$ and $Q_0(x) = 0$. They satisfy the recurrence relations

$$P_{n+1}(x) = P_n'(x) + xQ_n(x), \quad Q_{n+1}(x) = Q_n'(x) + P_n(x).$$

Lemma

With polynomials P_m and Q_m defined through the m -th derivative of $\text{Ai}(x)$, we have

$$\sum_{j=0}^N \frac{1}{j!(N-j)!} (P_j(x)Q_{N+1-j}(x) - Q_j(x)P_{N+1-j}(x)) = 0, \quad N \geq 1.$$

Thanks for your attention!

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Questions?