

Airy Line ensemble

Goal:- Motivate and define Airy line ensemble.

A bit of random matrices :-

Physics:- Wigner introduced random symmetric / Hermitian matrices to study energy spectrum of heavy nuclei. Interested in energy spectral gaps.

Statistics:- Wishart studied sample covariance of iid sample of a population.

Gaussian Unitary Ensemble (GUE) :-

$$M_n = (\varepsilon_{ij})_{i,j=1,2,\dots,n}$$

$\varepsilon_{ij} = \varepsilon_{ji}^*$ - complex conjugate.

$$\varepsilon_{ij} \sim N_C(0, I) \quad i \neq j$$

$$\varepsilon_{ji} \sim N_R(0, I).$$

and entries of M_n are independent
save for the condition $M_n^+ = M_n$.

Density of M_n w.r.t. Lebesgue :-

$$\mu_{M_n} = C_n \left(\prod_{1 \leq i < j \leq n} e^{-|\varepsilon_{ij}|^2} \right) \left(\prod_{i \leq n} e^{-|\varepsilon_{ii}|^2} \right) dM_n$$

$$= C_n e^{-\frac{\text{tr}(M_n^2)}{2}} dM_n$$

For a deterministic unitary U ,

$$\text{Tr}[(U M U^+)^2] = \text{Tr}(U M^2 U^+) = \text{Tr}(M^2)$$

Hence M_{M_n} is invariant under unitary transformations, hence the name GUE.

Ginibre formula (density of e.v.) :-

$$f_n(\lambda_1, \dots, \lambda_n) = C_n \cdot |\Delta_n(\lambda)|^2 e^{-\sum_{i=1}^n \lambda_i^2/2}$$

where $\lambda_1 > \lambda_2 > \dots > \lambda_n$ are eigenvalues of M_n (GUE) and

$$\Delta_n(\lambda) = \prod_{i < j} (\lambda_i - \lambda_j)$$

$\lambda = (\lambda_1, \lambda_2, \dots, \lambda_n)$ is Vandermonde determinant.

"Proof" :- Write $M_n = U D_n U^+$

This representation is unique upto

(i) permutation of diagonal entries

of D_n

$$(ii) U \mapsto \begin{pmatrix} e^{i\theta_1} & & \\ & \ddots & 0 \\ 0 & \ddots & e^{i\theta_n} \end{pmatrix} U$$

$$\text{So } (R^n/S_n) \times (U(n)/\Pi^n) \xrightarrow{\text{bijection, } n^2} R$$

Hermitian
matrices

Π -circle

S_n - permutation group

Let $\lambda_1, \lambda_2, \dots, \lambda_n, \omega_1, \dots, \omega_{n(n-1)}$

be the local coordinates of

$$(R^n/S_n) \times (U(n)/\Pi^n)$$

Want to compute Jacobian of

$$(\lambda_1, \lambda_2, \dots, \lambda_n, \omega_1, \dots, \omega_{n(n-1)}) \mapsto$$

$$(\epsilon_{11}, \epsilon_{22}, \dots, \operatorname{Re}(\epsilon_{12}), \dots)$$

we get that

$$\text{Jacobian} = |\Delta_n(\lambda)|^2 \times \text{fun}^c \text{ of } w_1, w_2, \dots$$

Hence, U and D_n are independent!

Since GUE has the measure

$$C_n e^{-\frac{1}{2} \text{Tr}(M_n^2)} dM_n$$

and noting $\text{Tr}(M_n^2) = \sum_{i=1}^n \lambda_i^2$, change of variables

we have the density for e.v.

$$p_n(\lambda) = C_n \cdot |\Delta_n(\lambda)|^2 \cdot e^{-\frac{1}{2} \sum \lambda_i^2}$$

Exercise:- Do the computation for

2×2 case i.e. M_2 .

Remarkably distribution of U (unitary)

in $M_n = U D_n U^+$ is uniform,

i.e. $U \stackrel{d}{=} \text{Haar}(U(n))$.
↳ group of unitary
matrices.

This is the magic of Gaussianity !!

For $Z \sim N_C(0, I)$, note that

$|Z|$ and $\arg(Z)$ are independent
and $\arg(Z)$ is uniform.

Physical intuition:- Think of e.v.
 $\lambda_1, \lambda_2, \dots, \lambda_n$ as charged particles
in a quadratic potential well.

i.e. $-\log p_n(\lambda) = \frac{1}{2} \|\lambda\|_2^2 +$

$$\sum_{i \neq j} \log \frac{1}{|\lambda_i - \lambda_j|} + C.$$

Determinantal point-processes:-

Fix a set S (say \mathbb{R}^2 or \mathbb{R}).

Let $K: S \times S \rightarrow \mathbb{R}$ be positive

semi-definite function, called a
kernel.

(Informal) A (simple) point process
on S is a determinantal point process
with kernel K if

$$dP(x_1, \dots, x_n) = C_n \cdot \det(K(x_i, x_j))_{i,j \leq n} \times \prod_{k=1}^n dx_k$$

Coming back to e.v. density of GUE,

Observe,

$$\Delta_n(\lambda) = \det \begin{pmatrix} 1 & \lambda_1 & \lambda_1^2 - \lambda_1^{n-1} \\ \vdots & \lambda_2 & \lambda_2^2 - \dots - \lambda_2^{n-1} \\ & & \ddots \end{pmatrix}$$

called Vandermonde identity.

By column operations,

$$\Delta_n(\lambda) = \det \begin{pmatrix} p_0(\lambda_1) & p_1(\lambda_1) & \dots & p_{n-1}(\lambda_1) \\ p_0(\lambda_2) & & & \\ \vdots & & & \\ p_0(\lambda_n) & & & p_{n-1}(\lambda_n) \end{pmatrix}$$

for any monic polynomial p_k of degree k .

Hence,

$$P_n(\lambda) = C_n \cdot \det \begin{pmatrix} p_0(\lambda_1) e^{-\lambda_1^2/4} & p_1(\lambda_1) e^{-\lambda_1^2/4} & \dots \\ p_0(\lambda_2) e^{-\lambda_2^2/4} & & \\ \vdots & \vdots & \vdots \\ p_0(\lambda_n) e^{-\lambda_n^2/4} & & \end{pmatrix}^2$$

$$= C_n'' \det(A^T A)$$

$$\text{where } A_{ij} = p_{j-1}(\lambda_i) e^{-\frac{1}{4}\lambda_i^2}$$

Take $\{p_k(x)\}_{k>0}$ to be a system
of orthogonal polynomials w.r.t.
measure $e^{-x^2/2} dx$. i.e.

$$\int_{\mathbb{R}} p_i(x) p_j(x) e^{-x^2/2} dx = \lambda_i^2 \delta_{ij}.$$

Example :- Hermite polynomials

$$p_k(x) = (-1)^k e^{x^2/2} \frac{d^k}{dx^k} e^{-x^2/2}$$

In this case, $\lambda_j^2 = j! \sqrt{2\pi}$.

Define $\phi_j(x) := \lambda_j^{-1} \cdot p_j(x) e^{-x^2/4}$
(Hermite func)

$\{\phi_j\}_{j>0}$ are orthonormal basis
of $L^2(\mathbb{R})$.

$$\text{Hence, } P_n(\lambda) = C_n \cdot \det(A^T A)$$

$$(A^T A)_{ij} = \sum_{k=1}^n A_{ki} A_{kj} = \sum_{k=0}^{n-1} \phi_k(\lambda_i) \phi_k(\lambda_j)$$

$$K_n(x, y) := \sum_{k=0}^{n-1} \phi_k(x) \phi_k(y) e^{-\frac{x^2+y^2}{4}}$$

$$= \frac{1}{\sqrt{2\pi}} \sum_{k=0}^{n-1} \frac{1}{k!} P_k(x) P_k(y)$$

Hence, $(\lambda_1, \lambda_2, \dots, \lambda_n)$ is a determinantal point process

with kernel $K_n =$

Wigner's semi-circle law :-

For M_n GUE matrix, spectral

radius = L^2 -matrix norm

Hence, $v \in \mathbb{R}^n, \|v\|_2 = 1,$

$$\text{Var}(\|M_n v\|_2^2) \sim n$$

To get meaningful scaling for the eigenvalues, need to scale by \sqrt{n} .

Theorem: Let $f: \mathbb{R} \rightarrow \mathbb{R}$ be a bounded

continuous fun^c (test fun^c). Then,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n f\left(\lambda_i/\sqrt{n}\right) = \int_{\mathbb{R}} f(x) d\sigma(x)$$

$$d\sigma(x) = \frac{1}{2\pi} \sqrt{4-x^2} dx \cdot 1[|x| \leq 2].$$

Convergence in expectation and
almost surely.

Think of point process as random measure

$$\frac{1}{n} \sum_{i=1}^n \delta(\lambda_i/\sqrt{n}) \rightarrow \frac{1}{2\pi} \int_{-\infty}^{\infty} \sqrt{4-x^2} dx$$

Spacing of GUE eigenvalues :-

Microscopic repulsion :-

Exercise :- $n=2$, λ_1, λ_2 e.v. of M_2 ,

then $|\lambda_1 - \lambda_2|$ has density $s^2 e^{-s^2/8}$,

\Rightarrow density vanishes at $s=0$ (repulsion)

Similar density for $|\lambda_j - \lambda_{j+1}|$ for

$n > 2$. In fact, "any" Hermitian random matrix has quadratic repulsion

Since $d\sigma$ has uniformly positive density inside $[-2, 2]$, spacing between eigenvalues is roughly

$$\frac{1}{\sqrt{n}}.$$

Scaling at the edge of the spectrum!

By semi-circle law,

$$\#\{\lambda_i : \lambda_i/\sqrt{n} \geq 2 - \varepsilon\} \approx \frac{n}{2\pi} \int_{2-\varepsilon}^2 \sqrt{4-x^2} dx$$
$$\approx \frac{2}{3\pi} n \varepsilon^{3/2}$$

To see finite number of e.v.,
correct scale is $\varepsilon \approx n^{-2/3}$.

$$\text{i.e. } \frac{\lambda_i}{\sqrt{n}} = 2 + d_i n^{-2/3}$$

$$\Rightarrow \lambda_i = 2\sqrt{n} + d_i n^{-1/6}.$$

Hence, the correct scaling to get
a non-trivial limit is,

$$\tilde{K}_n(x, y) = n^{-1/6} K_n(2\sqrt{n} + x n^{-1/6}, 2\sqrt{n} + y n^{-1/6})$$

Lemma: - $\forall x \in \mathbb{R}$,

$$\lim_{n \rightarrow \infty} |Ai(x) - n^{1/2} \phi_n(2\sqrt{n} + x n^{-1/6})| = 0$$

$$Ai(x) = \frac{1}{\pi} \int_0^\infty \cos(t^3/3 + xt) dt$$

Airy function.

ϕ_n - nth Hermite fun^c.

Thm: - $\lim_{n \rightarrow \infty} \tilde{K}_n(x, y) =$

$$\frac{Ai(x) Ai(y) - Ai(x) Ai(y)}{(x-y)}$$

Summary:- At the edge of the spectrum,

$$\lambda_i = 2\sqrt{n} + d_i n^{-1/6} \text{ (location)}$$

Spacing is roughly $n^{-1/6}$

(Compared to $n^{-1/2}$ in the bulk).

To get constant mean spacing,

consider $\{n^{1/6} \lambda_i\}_{i=1,-n}$, so

location of n^{th} particle is of

order $n^{2/3}$.

We define stationary Airy line ensemble.

Defⁿ: Extended Airy Kernel

$\kappa: \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ is given by,

$$\kappa(s, x; t, y) = \int_0^\infty e^{u(t-s)} A_i(x+u) A_i(y+u) du$$

if $s > t$.

$$\kappa(s, x; t, y) = - \int_{-\infty}^0 e^{u(t-s)} A_i(x+u) A_i(y+u) du.$$

if $s < t$.

The stationary Airy line ensemble

$$A = (A_1, A_2, \dots) \in \mathbb{Z}_{\geq 1} \times C(\mathbb{R})$$

is a collection of random continuous

curves $A_i: \mathbb{R} \rightarrow \mathbb{R}$, $A_1(t) > A_2(t) > \dots$

$t \in \mathbb{R}$ such that the density

$(t_1, y_1), (t_2, y_2), \dots, (t_m, y_m) \in A$ is

$$\det \left[K(t_i, y_i; t_j, y_j) \right]_{i,j \leq n} \prod_{j=1}^m dy_j$$

Remarks:- ① Put $t = s$ in the kernel
to get Airy point process.

② Since $Ai(x) \approx \frac{1}{x^{1/4}} \exp(-\frac{2}{3} x^{3/2})$
the density of points decay exponentially, hence a.s there is a "top" point / curve.

③ Existence of such process:-
see Corwin-Hammond [2014]

Scale and location of Airy
point process.

* Forget about eigenvalues scaling.
Want to study $(A_1(0), A_2(0), \dots)$

One-point correlation / density

of points of determinantal
point process with kernel K is

$$\mathbb{E}[\# \text{ pts. in } I] = \int_I K(x, x) dx$$

$I \subseteq \mathbb{R}$ (say).

For Airy point process,

$$K(x, x) = -x (A_i(x))^2 + (A_i'(x))^2$$

Using the asymptotics of Airy function, we have

$$\mathbb{E} [\# \text{pts in } (-T, \infty)] = \frac{2}{3\pi} T^{3/2} + O(1)$$

as $T \rightarrow \infty$.

and,

$$\mathbb{E} [\alpha_j(0)] = - \left(\frac{3\pi}{2} \cdot j \right)^{2/3} + O(1)$$

Fine location,

$$P(|\alpha_j(0) + \left(\frac{3\pi}{2} j \right)^{2/3}| > m j^{-1/3}) \leq e^{-\frac{m}{5}}$$

for $m > c_1 \log j$

Proof is out of scope for us.

(Determinantal str. of Airy pt. process.)

Now think of each line A_j as stationary perturbation of $(\frac{3\pi}{2} j)^{2/3}$.

Definition: Parabolic Airy line ensemble $R = (A_1(t) - t^2, A_2(t) - t^2, \dots)$

[Scaled version] $S = 2^{-1/2} R$.

Hence we have, with high probability

on paths.

$$S_j(t) + 2^{-1/2} t^2 = -2^{-7/6} (\frac{3\pi}{2})^{2/3} j^{2/3} + O(j^{0(1)-1/3})$$

Eq (1.2) in Aggarwal - Huang

If $j \asymp n$ and $t \asymp n^{1/3}$ then

S is of order $n^{2/3}$ in the box

$$[-n^{1/3}, n^{1/3}] \times [-n^{2/3}, 0] \quad (\text{t, x})\text{-space}$$

Plan:- Compare the Brownian Gibbsian line ensemble \mathcal{L} to S in the above box scale.

[Brownian motion characterisation of Airy line ensemble].

- Airy point process \hookrightarrow scaling of e.v. of GUE at the edge.

- Airy lines \leftrightarrow stationary perturbation of Airy points.

Dyson Brownian motion, $N \in \mathbb{N}$.

Let $B_{ij}(t)$ be independent complex

Brownian motion, $i < j$.

For $B_{ii}(t)$ be standard real BM.

All of them start at zero.

$$A(t) = \begin{pmatrix} B_{11}(t) & B_{12}(t) & \cdots \\ \overline{B_{12}(t)} & B_{22}(t) & \cdots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

Then the eigenvalues of $A(t)$ evolve according to the sol' of

$$\boxed{dx_i = dB_i + \sum_{j \neq i} \frac{dt}{\lambda_i - \lambda_j}} \quad (\#)$$

where B_i 's are iid real BM.

Fact:- As $N \rightarrow \infty$, the top k curves $\lambda_1(t) > \lambda_2(t) > \dots > \lambda_k(t)$

(after rescaling) converge to

Airy line ensemble.

$\boxed{\text{Ornstein-Uhlenbeck entries} \Rightarrow \text{Stationary Airy line ensemble. pictures.}}$

Non-intersecting Brownian motions!

Take N independent real BM

x_1, x_2, \dots, x_n and condition

it to never collide, say

$$x_1(t) > x_2(t) - \dots \rightarrow x_n(t).$$

By Doob's h -transform,

the new generator of the process

$$I^h(f) = \frac{1}{h} L(hf)$$

$$\text{where } h(x) = \Delta(x) = \prod_{i < j} (x_i - x_j)$$

$$\text{and } L(f) = \frac{1}{2} \sum_{i=1}^N \partial_{x_i}^2 f.$$

The induced drift is

$$\nabla \log h(x) = \left(\sum_{j \neq i} \frac{1}{x_i - x_j} \right)_{i=1}^n$$

$$\Rightarrow dx_i(t) = dB_i(t) + \sum_{j \neq i} \frac{dt}{x_i(t) - x_j(t)}$$

By uniqueness of strong solⁿ of

SDE, x_i 's are Dyson BM.

In conclusion, Airy line ensemble
is the ensemble of top curves of
non-intersecting Brownian motions(!)