

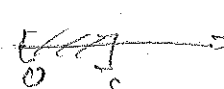
# Lecture 5

## Summary

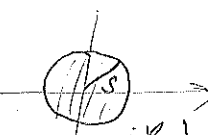
generating functional = Fredholm determinant for  $k$ -th gap on  $\mathbb{R}$

$$E(S; \beta) = 1 + \sum_{e=1}^N \frac{(-\beta)^e}{e!} \int_0^S dx_1 \dots \int_0^S dx_e \text{Re}(A_{x_1, \dots, x_e})$$

with  $E_k(S) = (-1)^k \frac{\partial^k}{\partial \beta^k} E(S; \beta) \Big|_{\beta=1}$



analogous quantity on  $\mathbb{C}$  - gap w/ radial domain

$$E(S; \beta) = 1 + \sum_{e=1}^N \frac{(-\beta)^e}{e!} \int_{\mathbb{D}} dx_1 \dots \int_{\mathbb{D}} dx_e \frac{e^{2i\theta}}{r} \text{Re}(A_{r_1 e^{i\theta_1}, \dots, r_e e^{i\theta_e}})$$


with the same relation to  $E_k(S)$  with gap on  $\mathbb{C}$

$\Rightarrow$  individual eigenvalue distributions w/ radial ordering

\* repeat: Andrić def integral identity

Andreas - integration formula (19th century) Exercise 14: Proof

$$\int d\mu(x_1) \dots \int d\mu(x_N) \det[\phi_i(x_j)]_{i,j=1}^N \det[\psi_i(x_j)]_{i,j=1}^N = N! \det \left[ \int d\mu(x) \phi_i(x) \psi_j(x) \right]_{i,j=1}^N$$

for any measurable functions  $\phi_i, \psi_j$  on any domain  $\subseteq \mathbb{R}$ , a.s. the int. exist.

The proof merely uses Laplace expansion

$$\sum_{\sigma \in S_N} (-1)^{\sigma} \prod_{e=1}^N \int d\mu(x_{\sigma(e)}) \phi(x_e) \psi_e(x_{\sigma(e)}) = N! \sum_{\sigma \in S_N} (-1)^{\sigma} \prod_{e=1}^N \int dx \phi_{\sigma(e)}(x) \psi_e(x)$$

$\downarrow$   
 $\phi_{\sigma(\sigma(e))}(x_{\sigma(e)})$

$$\Rightarrow E_0(v) = \det \left[ \int_{D_v} dz \psi_{j-1}(z) \overline{\psi_{k-1}(z)} \right]_{j,k=1}^N = \det \left[ \delta_{jk} - \int_{D_v} dz \frac{\psi_{j-1}(z) \overline{\psi_{k-1}(z)}}{j-1} \right]_{j,k=1}^N$$

these are orthonormal on  $\mathbb{C}$

which is the Fredholm determinant for the integration domain  $D_v$

\* We want to diagonalize the <sup>matrix inside the</sup> determinant and find its eigenvalues

$$E_0(v) = \prod_{j=1}^N (1 - f_j) \quad \text{where the } f_j \text{ are the ev of the}$$

integral eq

$$\lambda \varphi(u) = \int_{D_v} dz w(z) \sum_{j=0}^{N-1} \frac{P_j(u) \overline{P_j(z)}}{h_j} \varphi(z)$$

$= \sum_{j=0}^{N-1} \frac{h_j}{h_j} \varphi_j(u) \varphi_j(z)$

the same steps can be repeated for the generating function

$$E(v; z) = \det \left[ \delta_{jk} - \int_{D_v} dz w(z) \frac{P_{j-1}(z) \overline{P_{k-1}(z)}}{h_{j-1} h_{k-1}} \right] = \prod_{j=1}^N (1 - z f_j)$$

$\Rightarrow$  all h-h gaps, e.g.

$$k=1: \left( E_1(z) = -\partial_z E_0(v; z) \Big|_{z=0} = \prod_{j=1}^N (1 - f_j) \sum_{k=1}^N \frac{f_k}{1 - f_k} \right) \text{ etc.}$$

• Example Airy:  $w(z) = e^{-t^2}$ ,  $h_k = \sqrt{k!}$ ,  $P_e(z) = z^e$

$$\Rightarrow \int_{D_r} dz \psi_{j-1}(z) \psi_{k-1}(z) = \int_0^r dt e^{-t^2} \frac{(te^{it})^{j-1} (te^{-it})^{k-1}}{\sqrt{(j-1)!(k-1)!}}$$

$$z = te^{i\varphi}$$

$$= \delta_{jk} \frac{2\pi \int_0^r dt t^{2j-1} e^{-t^2}}{(j-1)!} = \delta_{jk} f_j$$

is already diagonal! true for weights  $w = w(z)$

$$\Rightarrow f_j = \frac{\int_0^\infty ds s^{j-1} e^{-s}}{(j-1)!} = \frac{1}{\Gamma(j)} \int_0^\infty ds s^{j-1} e^{-s} = \frac{\Gamma(j, \infty)}{\Gamma(j)}$$

for  $j=1, \dots, N$

$$= e^{-r^2} \sum_{l=0}^{j-1} \frac{r^{2l}}{l!}$$

$$\Rightarrow E_0(r) = \frac{N}{r} \prod_{j=1}^N \frac{\Gamma(j, r^2)}{\Gamma(j)}$$

is normalised as  $\frac{\Gamma(j, \infty)}{\Gamma(j)} = 1$

$$\Rightarrow P_d(r) = -r \partial_r E_0(r) = \sum_{k=1}^N \frac{2r^{2j} e^{-r^2}}{\Gamma(k)} \frac{N}{r} \prod_{j=1, j \neq k}^N \frac{\Gamma(j, r^2)}{\Gamma(j)}$$

the distribution of the smallest eigenvalue in radius

\* Other correlation functions - the spacing distribution:

outline: for the 3 classical Wigner-Dyson ensembles the spacing distribution

the probability that two consecutive ev have distance  $s$ ,

can be computed analytically for  $N=2$ : for  $V = \frac{1}{2}x^2, 4$  (GOE/GSE)

Exercise: compute  $\rho^B(s)$  for  $N=2$  using Wigner surmise

$$\rho^B(s) = a_\beta s^\beta e^{-b_\beta s^2}$$

where  $a_\beta, b_\beta$  are normalisation constants that follow from

$$1 = \int_0^\infty ds \rho^B(s) = \int_0^\infty ds s \rho^B(s)$$

\* it turns out that this is an excellent approximation to  $\rho^B(s)$  in the limit  $N \rightarrow \infty$  on the bulk of the spectrum.

The exact result is a Fredholm det of the sine-kernel (or  $U$  of spherical func, see Mehta's book)

18

and the spacing distribution is maybe not as unambiguously defined. We will choose the following def:

psj: Spacing distributions: given that there is an ev at  $z_0 \in \mathbb{C}$ , what is the probability that in vacuum distance  $S = |z - z_0|$  the next ev is found at distance  $S$ ? Its derivative gives:  $P(S)$

\* For the Ginibre ensemble (and in particular others) we can easily compute a closed form expression for any  $N$  of the following: spacing at the origin: given an ev at  $z_0 = 0$ , what is the probab to find the next ev at vacuum  $S$ ?

→ because for Ginibre (and others) there is translation invariance in the bulk at large  $N$ , this is expected to be true at  $z_0 \neq 0$  too

$$P(z_1=0, |z_2|, \dots, |z_N| \geq S) = \int_S^\infty dz_2 v_2 \dots \int_S^\infty dz_N v_N \prod_{l=1}^N \int_0^\infty d\varphi_l \frac{1}{C_N} P(z_1=0, z_2, \dots, z_N)$$

$$= \frac{1}{N! \prod_{j=0}^{N-1} h_j} \int_S^\infty dz_2 v_2 \dots \int_S^\infty dz_N v_N \int_0^\infty d\varphi_1 w(z_1=0) \prod_{l=2}^N \int_0^\infty d\varphi_l w(z_l) \prod_{j>i>2} |z_j - z_i| \prod_{k=2}^N |z_k - 0|^2$$

⇒ this is proportional to the gap probability  $E_0(S)$  for  $N-1$  ev with modified weight  $w(z) \rightarrow |z|^2 w(z) \stackrel{V=1}{\approx} w(z)$

\* example Ginibre:  $e^{-|z|^2} \rightarrow |z|^2 e^{-|z|^2} \stackrel{V=1}{\approx}$  rectangular Ginibre with 1 zero mode ✓

$$\Rightarrow \int_0^{\infty} dt t^{j-1} e^{-t^2} = \int_0^{\infty} dt t^{j+1} e^{-t^2}$$

$$\Rightarrow P(z_1=0, |z_2|, \dots, |z_N| \geq r) = \prod_{j=1}^{N-1} \frac{\Gamma(j+1, r^2)}{\Gamma(j+1)} = E_0(r) \Big|_{N-1}^{(V=1)}$$

$\Rightarrow$  spacing distrib. at the origin

$$P(\psi) = -\partial_r E_0(\psi) = \sum_{k=1}^{N-1} \frac{1}{\Gamma(k+1)} \left( 2r r^{2k} e^{-r^2} \right) \frac{r^{N-1}}{4} \frac{\Gamma(N+1, r^2)}{\Gamma(N-k)}$$

$$\underset{r \rightarrow 0}{\sim} r^3 + \mathcal{O}(r^5)$$

Exercise: show this

\* the repulsion at small distance  $r \rightarrow 0$  is cubic compared to quadratic for the GUE and thus weaker. This cubic behaviour seems to be universal for other distributions, cf. cond. mat / 9610073.

\* Note that  $N=2$  is not a good approximation for the large- $N$  result (cf. 0907.4195)

# diag 2 products of Gaussian random matrices

consider  $J_i \in \mathbb{C}^{N \times N}$  with  $i=1, 2, \dots, m$ , with

distribution  $P(J_i) = c \exp[-\text{Tr} J_i J_i^\dagger]$

Q: What is the distribution of complex eigenvalues  $z_1, \dots, z_N$  of the

product matrix  $Y_m = J_1 J_2 \dots J_m \neq Y_m^\dagger$

- the same Q is much more difficult when multiplying  $GUE$  matrices!

- we could also make every factor rectangular

in analogy we define the partition function  $C_N^{(m)}$

$C_N^{(m)} \sim \frac{1}{N} \int \prod_{i=1}^m [dJ_i] e^{-\text{Tr}(J_i J_i^\dagger)}$

A: the resulting joint density of  $z_1, \dots, z_N$  is a OPP:

$P_N^{(m)}(z_1, \dots, z_N) = \frac{1}{N} \prod_{i=1}^m w_m(z_i) |\Delta_N(z)|^2$

with weight  $w_m(z) = \frac{1}{\pi^{m-1}} G_{0m}^{m0} \left( \frac{\cdot}{0 \dots 0} \mid |z|^2 \right)$

given by a Meijer G-function  $G_{pq}^{mn} \left( \begin{matrix} a_1 \dots a_n \\ b_1 \dots b_p \end{matrix} \mid x \right)$

→ in order to understand this result we need to understand

2 on gradients:

\* a generalized Schur decomposition ⇒ jacobian

\* how does the Meijer G arise?

⇒ in the next step we need to compute the norms  $h_j^{(m)} \sim (j!)^m \Rightarrow$  kernel