

Eigenvalue density correlation functions

We define the n -point eigenvalue correlation function as

$$R_{N, n}^{(\beta)}(x_1, \dots, x_n) = \frac{N!}{(N-n)!} \int_D dx_1 \dots dx_n P_N^{(\beta)}(x_1, \dots, x_n)$$

where the jpdf is given by

$$P_N^{(\beta)}(x_1, \dots, x_n) = \frac{C_N^{(\beta)}}{Z_N^{(\beta)}} \prod_{j=1}^n \frac{1}{x_j} e^{-V(x_j)} \left| \Delta_N(x) \right|^\beta$$

$\beta = 1, 2, 4$ for $(OE/UE/SE)$

we dropped the $C_N^{(\beta)}$'s

with $D = \mathbb{R}$

$$P_N^{(\beta)}(H) = e^{-\text{Tr} V(H)}$$

$$\frac{C_N^{(\beta)}}{Z_N^{(\beta)}} \prod_{j=1}^n \frac{1}{x_j} \underbrace{x_j^{\frac{\beta}{2}(\nu+1) - 1}}_{w(x_j)} e^{-V(x_j)} \left| \Delta_N(x) \right|^\beta$$

with $D = \mathbb{R}_+$

$$P_N^{(\beta)}(H) = e^{-\text{Tr} V(H+H^\dagger)}$$

Wishart-Laguerre-divided $O/U/SE$

$H \ N \times (N+\nu)$ $\left\{ \begin{array}{l} \text{real asym.} \\ \text{complex non-Her} \\ \text{quats. real non-self dual} \end{array} \right.$
 $\nu \in \mathbb{N}$

6 ensembles of RMT with real eigenvalues

• in the second 3 ensembles we compute either

singular values $H = U \Lambda V$ $\lambda_i^2 = x_i$ for $\lambda_i > 0$
 or eigenvalues of $HH^\dagger = U \Lambda U^\dagger$ $x_i \geq 0$

- the Jacobian depends on N, ν for rectangular matrices

- characteristic equations in the 2 cases

$$\det \left(\frac{1}{N} HH^\dagger \right) = \prod_{j=1}^N (\lambda - x_j)$$

vs $\det \left(\gamma \frac{1}{2N+\nu} - \begin{pmatrix} 0_N & H \\ H^\dagger & 0_{N+\nu} \end{pmatrix} \right) = \gamma^\nu \prod_{j=1}^N (\gamma^2 - \lambda_j^2)$
 Dirac matrix

which can be seen from

(18)

$$\det \begin{pmatrix} A & B \\ C & D \end{pmatrix} \stackrel{A^{-1} \text{ exists}}{=} \det \begin{pmatrix} A & 0 \\ C & D - CA^{-1}B \end{pmatrix} = \det A \det (D - CA^{-1}B)$$

$$\stackrel{D^{-1} \text{ exists}}{=} \det A \det (A - B D^{-1} C)$$

Computation of all n-point function for $\beta = 2$:

using $\Delta_N(\{x\}) = \det_{i,j} [P_{i-j}(x_i)]$

and $\det A = \det A^T$, $\det A \cdot \det B = \det A \cdot B$

$$\Delta_N(\{x\})^2 = \det_{i,j} [P_{i-j}(x_i)]^2 = \frac{N-1}{N} h_j \det_{i,j} [P_{i-j}(x_i)]^2$$

$$= \frac{N-1}{N} h_j \det_{i,k} \left[\sum_{j=1}^N A_{ij} A_{kj} \right]$$

$$= \frac{N-1}{N} h_j \det_{i,k} \left[K_N(x_i, x_k) = \sum_{e=0}^{N-1} P_e(x_i) P_e(x_k) \right]$$

- by multiplying the square root of the weights into rows and columns we may write the entire jpdf as a det

$$P_N^{(\beta=2)} = \frac{N-1}{N} h_j \det_{i,j,k \in N} \left[\omega(x_i)^{\frac{1}{2}} \omega(x_k)^{\frac{1}{2}} K_N(x_i, x_k) \right]$$

This is called determinantal point process.

Generalisations to det p.p. on \mathbb{C} exist, as well as to

Pfaffian point processes (later)

Theorem [Ryerson 1970, see Merka 5.1.4] here $D=2$ only

Let $k(x,y)$ be a complex valued function, with $k(x,y)^* = k(y,x)$

that is k hermitian in both x and y , s.t.k.

$$\int k(x,y) k(y,z) dy = k(x,z) \text{ holds}$$

Then
$$\int \det_{1 \leq i, j \leq N} [k(x_i, x_j)] dx_N = (C - N+1) \det_{1 \leq i, j \leq N-1} k(x_i, x_j)$$

with $C = \int k(x,x) dx$

Remark: in choosing $k(x,y) = \omega(x)\omega(y)^* \sum_{j=0}^{k-1} p_j(x) p_j(y)^*$

this reduces the $(N-k)$ -fold integral in $R_N^{(k)}(x_1, \dots, x_k)$ to a determinant of size $k \times k$.

Proof:

$$\det_{1..N} [k(x_i, x_j)] = \sum_{\sigma} (-1)^\sigma k(x_1, x_{\sigma(1)}) \dots k(x_N, x_{\sigma(N)})$$

- All $N!$ permutations can be reached starting from the identity $(-1)^{id} = +1$ and exchanging pairs of second arguments of k 's
- Start with Id and do all $(N-1)!$ permutations keeping the last arg. fixed $k(x_N, x_N)$
- $\Rightarrow \int dx_N k(x_N, x_N) = C$, which multiplies by definition

$$\det_{1..N-1} [k(x_i, x_j)]$$

- to do all remaining $(N-1) \cdot (N-1)!$ perm start again with Id (which we have counted already) and do 1 pair exchange with the second argument of the last kernel

$$k(x_1, x_2) \dots k(x_j, x_{j+1}) \dots k(x_N, x_N)$$

$\underbrace{\hspace{10em}}_{\uparrow}$

- there are $N-1$ possibilities (positions) to do so, we get a "-" from the 1 pair exchange

$$\int dx_N k(x_j, x_N) k(x_N, x_j) = k(x_j, x_j)$$

giving - Id of $N-1$ variables. Now we do the remaining $(N-1)!$ permutations of the $(N-1)$ arguments yielding

$$\det_{1..N-1} k$$

$$\Rightarrow \int dx_N \det_{1..N} [k(x_i, x_j)] = (c - (N-1)) \det_{1..N-1} [k(x_i, x_j)]$$

[alternatively see Mehta, Th 5.1.4 p 75, 76]

\Rightarrow compute all $\beta=2$ correlation functions

$$\begin{aligned}
 K_{N,k}^{(\beta=2)}(x_1, \dots, x_k) &= \frac{N!}{(N-k)!} \int dx_{k+1} \dots dx_N \prod_{j=1}^N \omega(x_j) \Delta_N(\{x\})^2 \cdot \frac{1}{N! \prod_{j=0}^{k-1} h_j} \\
 &= \frac{N!}{(N-k)!} \int dx_{k+1} \dots dx_N \prod_{j=0}^{k-1} h_j \det [\omega(x_i)^{\frac{1}{2}} \omega(x_j)^{\frac{1}{2}} \sum_{c=0}^{N-k} \rho(x_i) \rho(x_j)] \\
 &\qquad \qquad \qquad \frac{N! \prod_{j=0}^{k-1} h_j}{k(x_i, x_j), \text{ with } c=N}
 \end{aligned}$$

- the integrations yield

$$\begin{aligned}
 \int dx_N (N - (N-1)) \det_N &= \det_{N-1} \\
 \int dx_{N-1} \det_{N-1} &= (N - (N-1-1)) \det_{N-2} \\
 \int dx_{k+1} \det_{k+1} &= (N - (k+1-1)) \det_k
 \end{aligned}$$

$$\Rightarrow R_{N,k}^{(\beta=2)}(x_1, \dots, x_k) = \frac{(N-k)!}{(N-k)!} \det_{1 \dots k} \left[\omega(x_i)^{\frac{1}{2}} \omega(x_j)^{\frac{1}{2}} \prod_{l=0}^{k-1} P_l(x_i) P_l(x_j) \right] \quad (21)$$

$$= \prod_{e=1}^k \omega(x_e) \det_{1 \dots k} \left[\prod_{n=0}^{k-1} P_n(x_i x_j) \right]$$

Some authors define the kernel with or without the weights

You will also find the notation of a

$$\text{wave-function } \Psi_k(x) = \omega(x)^{\frac{1}{2}} P_k(x)$$

as then the Ψ are orthonormal functions

$$\boxed{\int dx \Psi_k(x) \Psi_l(x) = \delta_{kl}}$$

$$\Rightarrow R_{N,k}^{(\beta=2)}(x_1, \dots, x_k) = \det_{1 \leq i, j \leq k} \left[\prod_{e=0}^{k-1} \Psi_e(x_i) \Psi_e(x_j) \right]$$

• all correlation functions for $\beta=1, 4$ can be expressed as a Pfaffian ($\text{Pf} A = \det A^{\frac{1}{2}}$) or Qdet of a quaternion-valued (2×2 matrix valued) kernel of skew-orthogonal polynomials

$$\text{scalar products: } \langle f, g \rangle_2 = \int f(x) g(x) \omega(x) dx \quad \beta=2 \text{ ordinary}$$

$$\text{skew } \sim \sim: \langle f, g \rangle_4 = \int [f(x) g'(x) - f'(x) g(x)] \omega(x) dx = - \langle g, f \rangle_4 \quad \beta=4$$

$$\langle f, g \rangle_1 = \iint f(x) g(y) \varepsilon(x-y) \omega(x) \omega(y) dx dy = - \langle g, f \rangle_1 \quad \beta=1$$

$$\text{where } \varepsilon(x) = \begin{cases} +\frac{1}{2} & x > 0 \\ -\frac{1}{2} & x < 0 \end{cases}$$

Appearance of the scalar product:

$\beta=4$: replace $\Delta_N(\{x\})^4$ in jpdf.

$$\Delta_N^4 = \prod_{j>i} (x_j - x_i)^4 = \det \left[x_i^{0} \quad j x_i^{j-1} \right]_{\substack{1 \leq i \leq N \\ 0 \leq j \leq 2N-1}} = \det \left[\phi_j(x_i) \quad \phi_j'(x_i) \right]_{\substack{1 \leq i \leq N \\ 0 \leq j \leq 2N-1}}$$

use 2. de Bruijn formula 1955

$$\int dx_1 \dots \int dx_N \det \left[\phi_j(x_i) \quad \phi_j'(x_i) \right]_{\substack{1 \leq i \leq N \\ 0 \leq j \leq 2N-1}} = N! \text{Pf} \left[(\phi_j(x) \phi_k'(x) - \phi_k(x) \phi_j'(x)) dx \right]$$

$$\Rightarrow Z_N^{(\beta=4)} = \binom{2N}{N} \int dx_1 \dots \int dx_N \prod_{i=1}^N w(x_i) \det \left[\phi_j(x_i) \quad \phi_j'(x_i) \right]$$

$$= \binom{2N}{N} \prod_{j=0}^{2N-1} h_j^{(\beta=4)} N! \text{Pf} \left[\int \frac{\phi_j(x) \phi_k'(x)}{h_j^{(\beta=4)}} - \frac{\phi_k'(x) \phi_j(x)}{h_k^{(\beta=4)}} dx \right]$$

where ($N=2s$ even), A $2N \times 2N$, $A = -A^T$

$$\text{Pf} [A] = \frac{1}{(2^s s!)} \sum_{\substack{P \text{ all} \\ N! \text{ perms} \\ \text{of } N \text{ indices}}} (-1)^P A_{i_1 i_2} \dots A_{i_{2s-1} i_{2s}}$$

$$= \sum_{\substack{P \text{ all perms} \\ \text{s.t. } i_1 < i_2, i_3 < i_4, \dots, i_{2s-1} < i_{2s}}} (-1)^P A_{i_1 i_2} \dots A_{i_{2s-1} i_{2s}}$$

example 2×2 : $\text{Pf} A = A_{12}$

4×4 : $\text{Pf} A = A_{12} A_{34} - A_{13} A_{24} + A_{14} A_{23}$

note in $\det A = \sum_{\sigma} (-1)^\sigma A_{1\sigma(1)} \dots A_{N\sigma(N)}$ every index appears twice, here only once

If we choose the $Q_j(x)$ s.t.

$$\langle Q_{2j}, Q_{2k+1} \rangle_{\mathcal{H}} = -\langle Q_{2k+1}, Q_{2j} \rangle_{\mathcal{H}} = h_j^{(a, \alpha)} S_{kj}$$

$$\langle Q_{2j}, Q_{2k} \rangle_{\mathcal{H}} = \langle Q_{2j+1}, Q_{2k+1} \rangle_{\mathcal{H}} = 0$$

We have that $\text{pf}[\langle Q_i, Q_j \rangle] = 1$ and we can construct a kernel from them

Example:
[Laguerre 1835]

Laguerre weight $w(x) = x^a e^{-x}$ on \mathbb{R}_+

$$Q_{2j+1}(x) = -L_{2j+1}^{(a)}(x) + L_{2j}^{(a)}(x) \Rightarrow Q_{2j+1}^{(1)}(x) = L_{2j}^{(a)}(x)$$

$$Q_{2j}^{(1)}(x) = L_{2j}^{(a)}(x) - L_{2j-1}^{(a)}(x) - \frac{(2j+a-1)}{(2j-1)} Q_{2j-2}^{(1)}(x)$$

$$\Rightarrow Q_{2j}^{(1)}(x) = -L_{2j-1}^{(a)}(x) - \frac{(2j+a-1)}{(2j-1)} Q_{2j-2}^{(1)}(x)$$

$$h_j^{(1)} = \frac{\Gamma(2j+a+1)}{(2j)!}$$

• the skew orthogonal polynomials enjoy a determinantal rep:

$$Q_{2j}(x) = \langle \det(x-H) \rangle$$

$$Q_{2j+1}(x) = \langle \det(x-H) (\text{Tr}(H) + x + \text{const.}) \rangle$$

as the odd polynomials are only defined up to $\text{const.} \cdot Q_{2j}(x)$
(skew-product!)

$\beta=1$ is analogous, except we need another integration formula.

replace $\det x_i^{j-1} = \det P_j(x_i)$, remove absolute value of $|\Delta_N|^{-1}$ by ordering eigenvalues

$$\begin{aligned} Z_N^{(\beta=1)} &= \tilde{C}_N^{(\beta=1)} \int dx_1 \dots dx_N w(x_1) \dots w(x_N) |\Delta_N(\{x\})|^{-1} \\ &= N! C_N^{(\beta=1)} \int_{x_1 \leq x_2 \leq \dots \leq x_N} dx_1 \dots \left(dx_N \frac{N}{N} w(x_j) \right) \det [P_j(x_i)] \end{aligned}$$

1. de Bruijn formula:

$$\int_{x_1 \leq \dots \leq x_N} dx_1 \dots dx_N \det_{1 \leq i, j \leq N} [q_i(x_j)] = Pf \left[\iint 2\varepsilon(x-y) q_i(x) q_j(y) dx dy \right]$$

\Rightarrow skew-product for $\beta=1$ $\langle d_i, g \rangle_1$, same construction of SOP $\langle P_{i1}, P_{i2}, \dots \rangle_1 = \frac{1}{g} \sum_{j=1}^{(\beta-1)} S_{ij}$ other's 0, same rep of P's terms of $\langle \det(\tau_n) \rangle$