

# Physical random matrices

## Heavy nuclei and quantum mechanics

- Introduced by Wigner in 1950's to explain statistics of highly excited nuclei resonances.
- Global time reversal constrains the elements to real.
- No preferential basis:  $P(X) = P(O^T X O)$ , e.g.  $P(X) \propto e^{-\text{Tr } X^2/2}$ . Orthogonal invariance.

In the case of no time reversal symmetry, elements will be complex.  
Unitary invariance.

In the case of a time reversal symmetry  $T^2 = -1$  (relevant to a finite dimensional system with an odd number of spin 1/2 particles),  
 $T = \mathbb{Z}_{2N} K$ , where  $K$  denotes complex conjugation, and

$$\mathbb{Z}_{2N} = \mathbb{I}_N \otimes \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}.$$

- Matrix commuting with  $T$  must have the additional property

$$X = \mathbb{Z}_{2N} \bar{X} \mathbb{Z}_{2N}^{-1}.$$

- Viewed as an  $N \times N$  matrix,  $X$  must have  $2 \times 2$  blocks

$$\begin{bmatrix} z & w \\ -\bar{w} & \bar{z} \end{bmatrix}.$$

Real quaternions.

- Invariance with respect to conjugation by  $N \times N$  unitary matrices with real quaternion elements.

## Dirac operators and QCD

Leads to random Hermitian matrices with a special block structure.

- Non-zero eigenvalues of the massless Dirac operator occur in pairs  $\pm\lambda$ .
- In the chiral basis, all eigenfunctions are also eigenfunctions of  $i\gamma_5$ , with eigenvalue  $+1$  or  $-1$ . Matrix elements between eigenfunctions with same eigenvalue must vanish. Implies block structure, with zero blocks in top left and bottom right.
- Application to QCD requires Dirac operator has given number  $\nu$  say of zero eigenvalues.

- Hence the structure

$$\begin{bmatrix} 0_{n \times n} & X_{n \times m} \\ X_{m \times n}^\dagger & 0_{m \times m} \end{bmatrix}$$

where  $n - m = \nu$ .

- The positive eigenvalues of this matrix are given by the positive square root of the eigenvalues of  $X^\dagger X$ .

**Question** What is the eigenvalue distribution of  $X^\dagger X$ ?

## Random scattering matrices

Scattering within an irregular shaped domain, connected to a wave guide.

- Wave guide permits  $N$  distinct plane wave states.
- (Complex) amplitudes denoted  $\vec{I}$  for incoming,  $\vec{O}$  for outgoing states.
- Scattering matrix  $S$ ,

$$S\vec{I} = \vec{O}.$$

$S$  must be unitary.

- Time reversal symmetry requires  $T^{-1}ST = S^\dagger$ .
- For  $T^2 = 1$ , implies  $S = S^T$ .
- For  $T^2 = -1$  implies  $S = \mathbb{Z}_{2N}S^T\mathbb{Z}_{2N}^{-1} =: S^D$ .

Seek a measure on  $\{S\}$  such that it is invariant under appropriate conjugations.

- For no time reversal symmetry  $S \in U(N)$  with Haar measure

$$(d_H S) = (S^\dagger dS).$$

- For  $S = S^T$ ,  $S = U_N U_N^T$ ,

$$(d_H S) = ((U_N^T)^\dagger dS U_N^\dagger).$$

- For  $S = S^D$ ,  $S = U_{2N} U_{2N}^D$ ,

$$(d_H S) = ((U_{2N}^D)^\dagger dS U_{2N}^\dagger).$$

## Quantum conductance problems

Scattering within a quasi one-dimensional conductor (lead).

- $n$  available channels at left edge,  $m$  at right edge. At each end a reservoir causes current to flow.
- $\vec{I}_n$  left incoming states,  $\vec{O}_n$  left outgoing states.
- $\vec{I}'_m$  right incoming states,  $\vec{O}'_m$  right outgoing states.
- Scattering matrix  $S$ ,

$$S \begin{bmatrix} \vec{I} \\ \vec{I}' \end{bmatrix} = \begin{bmatrix} \vec{O} \\ \vec{O}' \end{bmatrix}.$$

- Scattering matrix  $S$ ,

$$S = \begin{bmatrix} r_{n \times n} & t'_{n \times m} \\ t_{m \times n} & r'_{m \times m} \end{bmatrix}.$$

- Landauer formula

$$G/G_0 = \text{Tr}(t^\dagger t)$$

where  $G_0 = 2e^2/h$  is twice the fundamental quantum unit of conductance.

**Question** What is the eigenvalue distribution of  $t^\dagger t$ ?

## Calculation of eigenvalue PDFs

### Hermitian matrices

- $H = [x_{jk}]_{j,k=1,\dots,N}$  real symmetric matrix —  $N(N + 1)/2$  independent variables.
- Diagonalization  $H = OLO^T$ .
- Want Jacobian for change of variables from independent elements of  $X$  to eigenvalues  $\lambda_1, \dots, \lambda_N$  and  $N(N - 1)/2$  linearly independent variables formed out of  $O$ .

### Wedge products

Define

$$du_1 \wedge \cdots \wedge du_n := \det[du_i(\vec{r}_j)]_{i,j=1,\dots,n}.$$

Suppose we change variables  $\{u_1, \dots, u_N\}$  to  $\{v_1, \dots, v_N\}$ . Since

$$du_i = \sum_{l=1}^n \frac{\partial u_i}{\partial v_l} dv_l$$

and

$$\left[ \sum_{l=1}^n \frac{\partial u_i}{\partial v_l} dv_l(\vec{r}_j) \right]_{i,j=1,\dots,n} = \left[ \frac{\partial u_i}{\partial v_j} \right]_{i,j=1,\dots,n} [dv_i(\vec{r}_j)]_{i,j=1,\dots,n}$$

it follows

$$du_1 \wedge \cdots \wedge du_n = \det \left[ \frac{\partial u_i}{\partial v_j} \right]_{i,j=1,\dots,n} dv_1 \wedge \cdots \wedge dv_n$$

thus allowing the Jacobian to be read off.

Let  $H$  be real symmetric, and let  $dH$  denote the matrix of differentials. We have

$$dH = dO L O^T + O dL O^T + O L dO^T.$$

Noting from  $O^T O = \mathbb{I}_N$  that  $dO^T O = -O^T dO$  it follows from this that

$$O^T dH O = O^T dO L - L O^T dO + dL = \begin{bmatrix} d\lambda_1 & (\lambda_2 - \lambda_1)\vec{o}_1^T \cdot d\vec{o}_2 & \cdots & (\lambda_N - \lambda_1)\vec{o}_1^T \cdot d\vec{o}_N \\ (\lambda_2 - \lambda_1)\vec{o}_1^T \cdot d\vec{o}_2 & d\lambda_2 & \cdots & (\lambda_N - \lambda_2)\vec{o}_2^T \cdot d\vec{o}_N \\ \vdots & \vdots & & \vdots \\ (\lambda_N - \lambda_1)\vec{o}_1^T \cdot d\vec{o}_N & (\lambda_N - \lambda_2)\vec{o}_2^T \cdot d\vec{o}_N & \cdots & d\lambda_N \end{bmatrix}$$

Require the following result.

**Proposition 1.** *Let  $A$  and  $M$  be  $N \times N$  matrices, where  $A$  is non-singular. For  $A$  real ( $\beta = 1$ ), complex ( $\beta = 2$ ) or real quaternion ( $\beta = 4$ ), and  $M$  real symmetric ( $\beta = 1$ ), complex Hermitian ( $\beta = 2$ ) or quaternion real Hermitian ( $\beta = 4$ )*

$$(A^\dagger dM A) = \left( \det A^\dagger A \right)^{\beta(N-1)/2+1} (dM).$$

Applying the proposition with  $\beta = 1$  to the LHS, and taking the wedge product directly on the RHS gives

$$(dH) = \prod_{1 \leq j < k \leq N} (\lambda_k - \lambda_j) \bigwedge_{j=1}^N d\lambda_j (O^T dO).$$

## Scaling argument

- There are  $N(N + 1)/2$  independent differentials in  $(dX)$ , and so

$$(dcH) = c^{N(N+1)/2} (dH).$$

- Since  $cH = OcLO^T$ ,  $(dcH)$  is a homogeneous polynomial of degree  $N(N - 1)/2$  in  $\{\lambda_j\}$ .
- Because the probability of repeated eigenvalues occurs with zero probability,  $(dX)$  must vanish for  $\lambda_j = \lambda_k$ .
- According to the last two facts, the dependence on the eigenvalues is proportional to  $\prod_{1 \leq j < k \leq N} (\lambda_k - \lambda_j)$ .

For complex Hermitian matrices,

$$(dH) = \prod_{1 \leq j < k \leq N} (\lambda_k - \lambda_j)^2 \bigwedge_{j=1}^N d\lambda_j (U^\dagger dU).$$

For Hermitian matrices with real quaternion elements

$$(dH) = \prod_{1 \leq j < k \leq N} (\lambda_k - \lambda_j)^4 \bigwedge_{j=1}^N d\lambda_j (S^\dagger dS).$$

## Degeneracies

Consider a complex Hermitian matrix  $[x_{jk} + iy_{jk}]_{j,k=1,\dots,N}$ . It has the same eigenvalues as the  $2N \times 2N$  real symmetric matrix

$$\left[ \left[ \begin{array}{cc} x_{jk} & y_{jk} \\ -y_{jk} & x_{jk} \end{array} \right] \right]_{j,k=1,\dots,N}.$$

Scaling argument applied to a doubly degenerate real symmetric matrix gives that the dependence on the eigenvalues is proportional to

$$\prod_{1 \leq j < k \leq N} (\lambda_k - \lambda_j)^2.$$

Viewed as a  $2N \times 2N$  complex matrix, the  $N \times N$  real quaternion matrix  $[q_{jk}]_{j,k=1,\dots,N}$  is doubly degenerate. Now viewing this complex matrix as a  $4N \times 4N$  real symmetric matrix, we have a four fold degeneracy.

The scaling argument applied to a four fold degenerate real symmetric matrix gives that the dependence on the eigenvalues is proportional to

$$\prod_{1 \leq j < k \leq N} (\lambda_k - \lambda_j)^4.$$



## Wishart matrices

For  $X$  a random  $n \times m$  rectangular matrix ( $n \geq m$ ),  $A := X^\dagger X$  is referred to as a Wishart matrix.

Relevance to multivariate statistics comes from noting

$$\frac{1}{n} X^T X = \left[ \frac{1}{n} \sum_{j=1}^n x_{k_1}^{(j)} x_{k_2}^{(j)} \right]_{k_1, k_2=1, \dots, m} \approx [\langle x_{k_1} x_{k_2} \rangle]_{k_1, k_2=1, \dots, m}.$$

Require the Jacobian for changing variables from the elements of  $X$  to the elements of  $A$  (and other associated variables).

**Proposition 2.** *Let the  $n \times m$  matrix  $X$  have real ( $\beta = 1$ ), complex ( $\beta = 2$ ) or real quaternion ( $\beta = 4$ ) elements, and suppose it has a PDF of the form  $F(X^\dagger X)$ . The PDF of  $A := X^\dagger X$  is then proportional to*

$$F(A)(\det A)^{(\beta/2)(n-m+1-2/\beta)}.$$

For the proof we will use a scaling argument, due to Olkin. This in turn makes essential use of the result

$$(A^\dagger dM A) = \left( \det A^\dagger A \right)^{\beta(N-1)/2+1} (dM).$$

noted in Proposition 1 above. Further, for  $B$  a  $m \times m$  fixed matrix, and  $X = YB$ , we need the result that

$$(dX) = (\det B^\dagger B)^{\beta n/2} (dY).$$

This follows by noting that the Jacobian for  $\vec{x}^T = \vec{y}^T B$  is  $(\det B^\dagger B)^{\beta/2}$ .

## Proof

- The PDF of  $A$  must equal  $F(A)h(A)$  for some  $h$ .
- Write  $A = B^\dagger V B$  where  $V$  is positive definite. According to Prop. 1, PDF of  $V$  equals

$$F(B^\dagger V B)h(B^\dagger V B) \det(B^\dagger B)^{(\beta/2)(m-1+2/\beta)}.$$

- Let  $X = YB$ , where  $Y$  is such that  $V = Y^\dagger Y$ . As already noted  $(dX) = (\det B^\dagger B)^{\beta n/2} (dY)$  and so the PDF of  $Y$  is

$$F(B^\dagger Y^\dagger Y B) (\det B^\dagger B)^{\beta n/2}.$$

- This is a function of  $Y^\dagger Y$ , so the PDF of  $V = Y^\dagger Y$  is

$$F(B^\dagger V B) (\det B^\dagger B)^{\beta n/2} h(V).$$

- Equating the two expressions for the PDF of  $V$  gives

$$h(B^\dagger V B) = h(V) (\det B^\dagger B)^{(\beta/2)(n-m+1-2/\beta)}.$$

- Set  $V = \mathbb{I}$  and note  $h(\mathbb{I}) = c$  to get the result.

## Eigenvalue PDF for Wishart matrices

- We have shown that the Jacobian for the change of variables  $A = X^\dagger X$  is proportional to

$$(\det A)^{(\beta/2)(n-m+1-2/\beta)}.$$

- We have shown that for Hermitian matrices, the Jacobian for the change of variables to its eigenvalues and eigenvectors is proportional to

$$\prod_{1 \leq j < k \leq N} |\lambda_k - \lambda_j|^\beta.$$

Hence, if  $X$  is distributed according to

$$\frac{1}{C} e^{-\text{Tr}(V(X^\dagger X))},$$

then the PDF of  $A = X^\dagger X$  is equal to

$$\frac{1}{C} e^{-\sum_{j=1}^m V(\lambda_j)} \prod_{j=1}^m \lambda_j^{(\beta/2)(n-m+1-2/\beta)} \prod_{1 \leq j < k \leq m} |\lambda_k - \lambda_j|^\beta,$$

where  $0 \leq \lambda_j < \infty$ .

Recall that the eigenvalues of

$$\begin{bmatrix} 0_{n \times n} & X_{n \times m} \\ X_{m \times n}^\dagger & 0_{m \times m} \end{bmatrix}$$

$\{x_j\}$  say, are related to the eigenvalues  $\{\lambda_j\}$  of  $X^\dagger X$  by  $x_j^2 = \lambda_j$ , and so  $\{x_j\}$  have PDF

$$\frac{1}{C} e^{-\sum_{j=1}^m V(x_j)} \prod_{j=1}^m |x_j|^{\beta(n-m+1-2/\beta)+1} \prod_{1 \leq j < k \leq m} |x_k^2 - x_j^2|^\beta.$$

## Unitary matrices

We seek the eigenvalue PDF corresponding to the Haar volume form  $(U^\dagger dU)$ .

Our strategy is to make use of the Cayley transform, by parametrizing  $U$  in terms of an Hermitian  $H$  so that

$$U = \frac{\mathbb{I}_N + iH}{\mathbb{I}_N - iH}.$$

Making use of the general operator identity

$$\frac{d}{da}(1 - K)^{-1} = (1 - K)^{-1} \frac{dK}{da} (1 - K)^{-1},$$

where  $K$  is assumed to be a smooth function of  $a$ , it follows

$$U^\dagger dU = 2i(\mathbb{I}_N + iH)^{-1} dH (\mathbb{I}_N - iH)^{-1}.$$

Holds with  $H$  real (real quaternion) for  $U$  symmetric (self dual quaternion).

Consequently, using Proposition 1,

$$(U^\dagger dU) = 2^{N(\beta(N-1)/2+1)} \det(\mathbb{I}_N + H^2)^{-\beta(N-1)/2-1} (dH).$$

Since  $H$  is complex Hermitian, the eigenvalue PDF in terms of  $\{\lambda_j\}$  is

$$\frac{1}{C} \prod_{l=1}^N \frac{1}{(1 + \lambda_l^2)^{\beta(N-1)/2+1}} \prod_{j < k} |\lambda_k - \lambda_j|^\beta.$$

But

$$\lambda_j = i \frac{1 - e^{i\theta}}{1 + e^{i\theta}}.$$

Thus the eigenvalue PDF in terms of  $\{\theta_j\}$  is

$$\frac{1}{C} \prod_{j < k} |e^{i\theta_k} - e^{i\theta_j}|^\beta.$$

## Blocks of unitary matrices

We seek the distribution of the non-zero eigenvalues of  $t^\dagger t$  in the decomposition

$$S = \begin{bmatrix} r_{n \times n} & t'_{n \times m} \\ t_{m \times n} & r'_{m \times m} \end{bmatrix}.$$

- Make use of singular value decompositions of individual blocks, for example

$$t = U_t \Lambda_t V_t^\dagger.$$

- $\Lambda_t$  is a rectangular diagonal matrix, diagonal entries consisting of the positive square roots of the non-zero eigenvalues of  $t^\dagger t$ .
- $U_t$  and  $V_t$  are  $m \times m$  and  $n \times n$  unitary matrices.

This leads to the parameterization

$$S = \begin{bmatrix} U_r & 0 \\ 0 & U_{r'} \end{bmatrix} L \begin{bmatrix} V_r^\dagger & 0 \\ 0 & V_{r'}^\dagger \end{bmatrix}$$

where

$$L = \begin{bmatrix} \sqrt{1 - \Lambda_t \Lambda_t^T} & i\Lambda_t \\ i\Lambda_t^T & \sqrt{1 - \Lambda_t^T \Lambda_t} \end{bmatrix}.$$

For  $S$  symmetric

$$V_r^\dagger = U_r^T, \quad V_{r'}^\dagger = U_{r'}^T,$$

while for  $S$  self dual quaternion

$$V_r^\dagger = U_r^D, \quad V_{r'}^\dagger = U_{r'}^D.$$

Using the method of wedge products, can show that the non-zero elements of  $\Lambda_t$  have the distribution

$$\prod_{j=1}^m \lambda_j^{\beta\alpha} \prod_{1 \leq j < k \leq m} |\lambda_k^2 - \lambda_j^2|^\beta, \quad \alpha = n - m + 1 - 2/\beta$$

where  $0 < \lambda_j < 1$ .

For  $\beta = 2$ , different approaches also work, and further a more general result holds.

**Proposition 3.** *Let  $U$  be an  $N \times N$  random unitary matrix chosen with Haar measure. Decompose  $U$  into blocks*

$$U = \begin{bmatrix} A_{n_1 \times n_2} & C_{n_1 \times (N-n_2)} \\ B_{(N-n_1) \times n_2} & D_{(N-n_1) \times (N-n_2)} \end{bmatrix}$$

where  $n_1 \geq n_2$ . The eigenvalue PDF of  $Y := A^\dagger A$  is proportional to

$$\prod_{j=1}^{n_2} y_j^{(n_1-n_2)} (1-y_j)^{(N-n_1-n_2)} \prod_{j < k}^{n_2} (y_k - y_j)^2.$$

Possible strategies:

- Wedge products.
- Matrix integrals.
- Orthogonal projections relating to the matrix structure  $A(A+B)^{-1}$ .

## Proof using matrix integrals

- Use the Ingham-Seigel type integral

$$\int e^{(i/2)\text{Tr}(HQ)} \left( \det(H - \mu \mathbb{I}_m) \right)^{-n} (dH) \propto (\det Q)^{(n-m)} e^{(i/2)\mu \text{Tr}Q},$$

valid for  $Q$  Hermitian, and  $\text{Re}(\mu) > 0$ , and the integration is over the space of  $m \times m$  Hermitian matrices.

- Regard  $A$  and  $C$  as general  $n_1 \times n_2$  and  $n_1 \times (N - n_2)$  complex rectangular matrices. Then the distribution of  $A$  is

$$\int \delta(AA^\dagger + CC^\dagger - \mathbb{I}_{n_2})(dC).$$

- The delta function can itself be written as a matrix integral

$$\int e^{-i\text{Tr}(H(AA^\dagger + CC^\dagger - \mathbb{I}_{n_2}))}(dH),$$

where the integration is over the space of  $n_2 \times n_2$  Hermitian matrices.

- Interchange order of integration, doing integration over  $C$  first. For this must regularize  $H \mapsto H - i\mu \mathbb{I}_n$ ,  $\mu > 0$ . Gives

$$\lim_{\mu \rightarrow 0^+} \int (\det(H - i\mu \mathbb{I}_{n_1}))^{-(N-n_2)} e^{i\text{Tr}(H(\mathbb{I}_{n_1} - AA^\dagger))}$$

- Evaluate using Ingham-Seigel type integral to get

$$(\det(\mathbb{I}_{n_1} - AA^\dagger))^{(N-n_1-n_2)}.$$

- Use Propostion 2 above to conclude the distribution of  $Y := A^\dagger A$  is proportional to

$$(\det Y)^{(n_1-n_2)} (\det(\mathbb{I}_{n_2} - Y))^{(N-n_1-n_2)}.$$

## Classical random matrix ensembles

Let  $\beta = 1, 2$  or  $4$  according to the elements being real, complex or quaternion real respectively.

In the case of Hermitian matrices, the eigenvalue PDFs derived above all have the general form

$$\frac{1}{C} \prod_{l=1}^N g(x_l) \prod_{1 \leq j < k \leq N} |x_k - x_j|^\beta.$$

Choosing the entries to be matrices to be independent Gaussians, when there is a choice, the form of  $g(x)$  is, up to scaling  $x_l \mapsto cx_l$ ,

$$g(x) = \begin{cases} e^{-x^2}, & \text{Gaussian} \\ x^a e^{-x} (x > 0) & \text{Laguerre} \\ x^a (1-x)^b (0 < x < 1) & \text{Jacobi} \\ (1+x^2)^{-\alpha} & \text{Cauchy} \end{cases}$$

These are the four classical weight functions from orthogonal polynomial theory, which can be characterized by the property that

$$\frac{d}{dx} \log g(x) = \frac{a(x)}{b(x)}$$

where

$$\text{degree } a(x) \leq 1, \quad \text{degree } b(x) \leq 2.$$

The corresponding eigenvalue PDF is said to define a classical random matrix ensemble.



## Gaussian $\beta$ ensemble

So far we've seen that the eigenvalue PDF

$$\frac{1}{C} \prod_{l=1}^N e^{-x_l^2} \prod_{1 \leq j < k \leq N} |x_k - x_j|^\beta$$

corresponds to random Hermitian matrices with probability measure

$$\frac{1}{C} e^{-\text{Tr } X^2},$$

where the elements are real ( $\beta = 1$ ), complex ( $\beta = 2$ ) or quaternion real ( $\beta = 4$ ). Said to define the GOE (Gaussian orthogonal ensemble), GUE (Gaussian unitary ensemble) and GSE (Gaussian symplectic ensemble).

We would like to give a meaning in the context of random matrix theory for general  $\beta > 0$ .

Inductively define a sequence of random matrices  $\{M_j\}_{j=1,2,\dots}$  by  $M_1 \sim a$  and

$$M_{N+1} = \begin{bmatrix} D_N & \vec{w} \\ \vec{w}^T & a \end{bmatrix}$$

where  $D_N = \text{diag}(a_1, \dots, a_N)$  where  $\{a_j\}$  denotes the eigenvalues of  $M_N$ .

## Relationship to GOE

Let  $a \in \mathbb{N}[0, 1]$ , and let  $w_j \in \mathbb{N}[0, 1/\sqrt{2}]$ . Analogous to above, define

$$M_{N+1} = \begin{bmatrix} M_N & \vec{w} \\ \vec{w}^T & a \end{bmatrix}$$

Let  $O_N$  be the real orthogonal matrix which diagonalizes  $M_N$ , so that  $M_N = O_N D_N O_N^T$ , and observe

$$\begin{bmatrix} O_N & \vec{0} \\ \vec{0}^T & 1 \end{bmatrix} \begin{bmatrix} M_N & \vec{w} \\ \vec{w}^T & a \end{bmatrix} \begin{bmatrix} O_N & \vec{0} \\ \vec{0}^T & 1 \end{bmatrix}^T \sim \begin{bmatrix} D_N & \vec{w} \\ \vec{w}^T & a \end{bmatrix}$$

**Question:** Can we understand this result using different reasoning?

## A random rational function

Recall

$$M_{N+1} = \begin{bmatrix} D_N & \vec{w} \\ \vec{w}^T & a \end{bmatrix}$$

where

$$D_N = \text{diag}(a_1, \dots, a_N).$$

For given  $\{a_j\}$ , we would like to compute the eigenvalue distribution of  $M_{N+1}$ . We have

$$\begin{aligned} \det(\lambda \mathbb{I}_{N+1} - M_{N+1}) &= \det \begin{bmatrix} \lambda \mathbb{I}_N - D_N & \vec{w} \\ \vec{w}^T & \lambda - a \end{bmatrix} \\ &= \det \begin{bmatrix} \lambda \mathbb{I}_N - D_N & -\vec{w} \\ \vec{0}^T & \lambda - a - \vec{w}^T (\lambda \mathbb{I}_N - D_N)^{-1} \vec{w} \end{bmatrix} \\ &= p_N(\lambda) (\lambda - a - \vec{w}^T (\lambda \mathbb{I}_N - D_N)^{-1} \vec{w}) \end{aligned}$$

But  $\lambda \mathbb{I}_N - D_N$  is diagonal, so its inverse is also diagonal, allowing us to conclude

$$\frac{p_{N+1}(\lambda)}{p_N(\lambda)} = \lambda - a - \sum_{i=1}^N \frac{q_i}{\lambda - a_i}, \quad q_i := w_i^2.$$

**Conclusion** The eigenvalues of  $M_{N+1}$  are given by the zeros of the above rational function.

The corresponding PDF can be computed for a certain choice of the distribution of the  $q_i$ , generalizing  $q_i \sim (\text{N}[0, 1/\sqrt{2}])^2 \sim \Gamma[1/2, 1]$  which corresponds to the GOE.

**Proposition 4.** Let  $w_i^2 \sim \Gamma[\beta/2, 1]$  where  $\Gamma[s, \sigma]$  refers to the gamma distribution, specified by the PDF  $\sigma^{-s} x^{s-1} e^{-x/\sigma} / \Gamma(s)$  ( $x > 0$ ). Given

$$a_1 > a_2 > \cdots > a_N$$

the PDF for the zeros of the random rational function

$$\lambda - a - \sum_{i=1}^N \frac{q_i}{\lambda - a_i}$$

is equal to

$$\begin{aligned} & \frac{e^{a^2/2}}{(\Gamma(\beta/2))^N} \frac{\prod_{1 \leq j < k \leq N+1} (\lambda_j - \lambda_k)}{\prod_{1 \leq j < k \leq N} (a_j - a_k)^{\beta-1}} \prod_{j=1}^{N+1} \prod_{p=1}^N |\lambda_j - a_p|^{\beta/2-1} \\ & \times \exp\left(-\frac{1}{2} \left( \sum_{j=1}^{N+1} \lambda_j^2 - \sum_{j=1}^N a_j^2 \right)\right) \end{aligned}$$

where

$$\infty > \lambda_1 > a_1 > \lambda_2 > \cdots > a_N > \lambda_{N+1} > -\infty$$

and

$$\sum_{j=1}^{N+1} \lambda_j = \sum_{j=1}^N a_j + a.$$

## Proof

- Because the  $q_i$  are positive, graphical considerations imply the interlacing condition.
- The summation constraint is equivalent to the statement that  $\text{Tr } M_{N+1} = \text{Tr } D_N + a$ .
- To compute the PDF we change variables from  $\{q_i\}_{i=1,\dots,N}$  to  $\{\lambda_j\}_{j=1,\dots,N}$ .
- The translations  $\lambda_j \mapsto \lambda_j - a$ ,  $a_j \mapsto a_j - a$  shows it suffices to consider the case  $a = 0$

With  $a = 0$  we have

$$\lambda - \sum_{i=1}^N \frac{q_i}{\lambda - a_i} = \frac{\prod_{j=1}^{N+1} (\lambda - \lambda_j)}{\prod_{l=1}^N (\lambda - a_l)}$$

From the residue at  $\lambda = a_i$  it follows

$$\frac{\prod_{j=1}^{N+1} (a_i - \lambda_j)}{\prod_{l=1, l \neq i}^N (a_i - a_l)} = -q_i.$$

We have just shown that

$$\frac{\prod_{j=1}^{N+1} (a_i - \lambda_j)}{\prod_{l=1, l \neq i} (a_i - a_l)} = -q_i.$$

Now use the method of wedge products to compute the Jacobian:

$$\bigwedge_{j=1}^N dq_j = \prod_{j=1}^N q_j \det \left[ \frac{1}{a_i - \lambda_j} \right] \bigwedge_{j=1}^N d\lambda_j.$$

Making use of the [Cauchy double alternant](#) identity, we read off that the Jacobian is equal to

$$\prod_{j=1}^N q_j \left| \frac{\prod_{1 \leq i < j \leq N} (a_i - a_j)(\lambda_i - \lambda_j)}{\prod_{i,j=1}^N (a_i - \lambda_j)} \right|.$$

We are given that the distribution of  $\{q_i\}$  is equal to

$$\frac{1}{(\Gamma(\beta/2))^N} \prod_{j=1}^N q_j^{\beta/2-1} e^{-\sum_{j=1}^N q_j}.$$

Must multiply these last two equations together, and substitute for  $q_i$ , noting

$$\sum_{j=1}^N q_j = \frac{1}{2} \left( \sum_{j=1}^{N+1} \lambda_j^2 - \sum_{j=1}^N \mu_j^2 \right).$$

Suppose now that  $a \sim N[0, 1]$ . Thus we must multiply the PDF of Proposition 4 by  $\frac{1}{\sqrt{2\pi}}e^{-a^2/2}$  and integrate over  $a$ . Doing this gives the PDF

$$\frac{1}{\sqrt{2\pi}(\Gamma(\beta/2))^N} \frac{\prod_{1 \leq j < k \leq N+1} (\lambda_j - \lambda_k)}{\prod_{1 \leq j < k \leq N} (a_j - a_k)^{\beta-1}} \prod_{j=1}^{N+1} \prod_{p=1}^N |\lambda_j - a_p|^{\beta/2-1} \\ \times \exp\left(-\frac{1}{2}\left(\sum_{j=1}^{N+1} \lambda_j^2 - \sum_{j=1}^N a_j^2\right)\right)$$

where

$$\infty > \lambda_1 > a_1 > \lambda_2 > \dots > a_N > \lambda_{N+1} > -\infty$$

- Denote the above conditional PDF  $G_N(\{\lambda_j\}, \{a_k\})$ .
- Denote the domain of support by  $R_N$ .
- Let  $\{a_j\}$  have PDF  $p_N(a_1, \dots, a_N)$ .
- Let  $\{\lambda_j\}$  have PDF  $p_{N+1}(\lambda_1, \dots, \lambda_{N+1})$ .

The PDFs  $\{p_n\}$  must satisfy the recurrence

$$p_{N+1}(\lambda_1, \dots, \lambda_{N+1}) \\ = \int_R da_1 \cdots da_N G_N(\{\lambda_j\}, \{a_k\}) p_N(a_1, \dots, a_N)$$

We seek the solution with  $p_0 := 1$ .

**Question** How can we solve this recurrence.

For  $\beta = 1$  we know that the solution of the recurrence is

$$\frac{1}{C} \prod_{j=1}^N e^{-\lambda_j^2/2} \prod_{1 \leq j < k \leq N} |\lambda_k - \lambda_j|.$$

It must then be that there is an integration formula implying the recurrence.

**New question** How can we derive such integration formulas?

We again take inspiration from the case of the GOE, applied to a general formula:

- Let  $M_N$  be a general  $N \times N$  real symmetric matrix with eigenvalues  $\{\lambda_j\}$ .
- Let  $\mu_i$  denote the top entry of the normalized eigenvector corresponding to  $\lambda_i$ .
- We have

$$\sum_{j=1}^N \frac{\mu_j^2}{\lambda - \lambda_j} = \left( (\lambda \mathbb{I}_N - M_N)^{-1} \right)_{11} = \frac{p_{N-1}(\lambda)}{p_N(\lambda)}.$$

Here the first equality follows from a spectral decomposition, while the second follows from Cramer's rule. Note that we must have

$$\sum_{j=1}^N \mu_j^2 = 1.$$



In the case  $\beta = 1$ ,  $M_N \in \text{GOE}$ , and is thus orthogonally invariant.

Consequently

$$\mu_i^2 \sim \frac{w_i^2}{w_1^2 + \dots + w_N^2} =: \rho_i$$

where each  $w_i^2 \sim \Gamma[1/2, 1]$ .

Generally, if  $w_i^2 \sim \Gamma[\beta/2, 1]$  then the PDF of  $\rho_1, \dots, \rho_N$  is equal to the Dirichlet distribution

$$\frac{\Gamma(N\beta/2)}{(\Gamma(\beta/2))^N} \prod_{j=1}^N \rho_j^{\beta/2-1}$$

where each  $\rho_j$  is positive and  $\sum_{j=1}^N \rho_j = 1$ .

**Question** What is the distribution of the roots of the random rational function

$$\sum_{j=1}^N \frac{\mu_j^2}{\lambda - \lambda_j}$$

when the  $\{\mu_j^2\}$  have a Dirichlet distribution?

**Proposition 5.** *Let  $\{\rho_i\}$  have the Dirichlet distribution*

$$\frac{\Gamma(N\beta/2)}{(\Gamma(\beta/2))^N} \prod_{j=1}^N \rho_j^{\beta/2-1}$$

*and let  $\{b_j\}$  be given. The roots of the random rational function*

$$\sum_{j=1}^N \frac{\rho_j}{x - b_j},$$

*denoted  $\{x_1, \dots, x_{N-1}\}$  say, have the PDF*

$$\frac{\Gamma(N\beta/2)}{(\Gamma(\beta/2))^N} \frac{\prod_{1 \leq j < k \leq N-1} (x_j - x_k)}{\prod_{1 \leq j < k \leq N} (b_j - b_k)^{\beta-1}} \prod_{j=1}^{N-1} \prod_{p=1}^N |x_j - b_p|^{\beta/2-1}$$

*where*

$$x_1 > b_1 > x_2 > b_2 > \dots > x_{N-1} > b_N.$$

The proof is very similar to that of Proposition 4. We seek to change variables from  $\{\rho_j\}$  to  $\{x_j\}$ .

- Begin by noting

$$\sum_{j=1}^N \frac{\rho_j}{x - b_j} = \frac{\prod_{l=1}^{N-1} (x - x_l)}{\prod_{l=1}^N (x - b_l)}.$$

- Equating the residue at  $x = b_j$  gives

$$\rho_j = \frac{\prod_{l=1}^{N-1} (b_j - x_l)}{\prod_{l=1, l \neq j}^N (b_j - b_l)}.$$

- Taking wedge products gives

$$\bigwedge_{j=1}^{N-1} d\rho_j = \prod_{j=1}^{N-1} \rho_j \det \left[ \frac{1}{b_j - x_k} \right]_{j,k=1, \dots, N-1} \bigwedge_{j=1}^{N-1} dx_j$$

- Now use the Cauchy double alternant identity to conclude the Jacobian for the change of variables is equal to

$$\prod_{j=1}^{N-1} \rho_j \left| \frac{\prod_{1 \leq j < k \leq N-1} (b_k - b_j)(x_k - x_j)}{\prod_{j,k=1}^{N-1} (b_j - x_k)} \right|.$$

- Multiply this Jacobian and the PDF for the distribution of the  $\{\rho_j\}$ , then substitute for the  $\rho_j$  to get the result.

## Consequence

Because the expression for the distribution of the roots of the rational function is a PDF for  $\{x_j\}$ , integrating over the region

$$x_1 > b_1 > x_2 > b_2 > \cdots > x_{N-1} > b_N$$

denoted  $R'_{N-1}$  say must give unity. Hence we have the integration formula

$$\begin{aligned} \int_{R'_{N-1}} dx_1 \cdots dx_{N-1} \prod_{1 \leq j < k \leq N-1} (x_j - x_k) \prod_{j=1}^{N-1} \prod_{p=1}^N |x_j - b_p|^{\beta/2-1} \\ = \frac{(\Gamma(\beta/2))^N}{\Gamma(N\beta/2)} \prod_{1 \leq j < k \leq N} (b_j - b_k)^{\beta-1}. \end{aligned}$$

This integration formula allows the solution of the recurrence to be verified.

**Proposition 6.** *The solution of the recurrence*

$$\begin{aligned} p_{N+1}(\lambda_1, \dots, \lambda_{N+1}) \\ = \int_R da_1 \cdots da_N G_N(\{\lambda_j\}, \{a_k\}) p_N(a_1, \dots, a_N) \end{aligned}$$

is given by

$$p_N(x_1, \dots, x_N) = \frac{1}{m_N(\beta)} \prod_{j=1}^N e^{-x_j^2/2} \prod_{1 \leq j < k \leq N} |x_k - x_j|^\beta$$

where

$$N! m_N(\beta) = (2\pi)^{N/2} \prod_{j=0}^{N-1} \frac{\Gamma(1 + (j+1)\beta/2)}{\Gamma(1 + \beta/2)}.$$

This PDF is said to define the Gaussian  $\beta$  ensemble.

## Proof

Substituting the stated form for  $p_N$  in the recurrence gives

$$\begin{aligned} \frac{1}{\sqrt{2\pi}} \frac{1}{(\Gamma(\beta/2))^N} \frac{1}{m_N(\beta)} e^{-\frac{1}{2} \sum_{j=1}^{N+1} \lambda_j^2} \prod_{1 \leq j < k \leq N+1} (\lambda_j - \lambda_k) \\ \times \int_{R_N} da_1 \cdots da_N \prod_{1 \leq j < k \leq N} (a_j - a_k) \prod_{j=1}^{N+1} \prod_{p=1}^N |\lambda_j - a_p|^{\beta/2-1}. \end{aligned}$$

Evaluating the integral according to the integration formula of the previous page verifies the recurrence.

## Three term recurrence

The characteristic polynomial  $p_N(x) := \prod_{l=1}^N (x - x_l)$ , where  $\{x_j\}$  is distributed according to the Gaussian  $\beta$ -ensemble, satisfies

$$\frac{p_{N-1}(x)}{p_N(x)} = \sum_{j=1}^N \frac{\rho_j}{x - x_j}$$

where

$$\rho_j \sim w_j^2 / (w_1^2 + \cdots + w_N^2), \quad w_j^2 \sim \Gamma[\beta/2, 1]$$

and

$$\frac{p_{N+1}(x)}{p_N(x)} = x - a - \sum_{j=1}^N \frac{w_j^2}{x - x_j}, \quad a \in \mathbb{N}[0, 1].$$

Hence

$$\frac{p_{N+1}(x)}{p_N(x)} = (x - a) - b_N^2 \frac{p_{N-1}(x)}{p_N(x)}$$

where

$$b_N^2 \sim (w_1^2 + \cdots + w_N^2) \sim \Gamma[N\beta/2, 1].$$

Rearranging gives the random three term recurrence

$$p_{N+1}(x) = (x - a)p_N(x) - b_N^2 p_{N-1}(x).$$

## Relationship to tridiagonal matrices

Consider a general real symmetric tridiagonal matrix

$$T_n = \begin{bmatrix} a_1 & b_1 & & & \\ b_1 & a_2 & b_2 & & \\ & b_2 & a_3 & & \\ & & & \ddots & b_{n-1} \\ & & & b_{n-1} & a_n \end{bmatrix}$$

By forming  $\lambda \mathbb{I}_n - T_n$  and expanding the determinant along the bottom row one sees

$$\det(\lambda \mathbb{I}_n - T_n) = (\lambda - a_n) \det(\lambda \mathbb{I}_{n-1} - T_{n-1}) - b_{n-1}^2 \det(\lambda \mathbb{I}_{n-2} - T_{n-2}).$$

## Conclusion

The Gaussian  $\beta$ -ensemble is realized by random tridiagonal matrices with

$$a_j \sim \text{N}[0, 1] \quad b_j^2 \sim \Gamma[j\beta/2, 1].$$

## Laguerre $\beta$ ensemble

Consider the LOE.

- This is realized by matrices  $X_{(n)}^T X_{(n)}$  where  $X_{(n)}$  is an  $n \times N$  rectangular matrix with Gaussian entries  $N[0, 1]$ .
- Have the recurrence

$$X_{(n+1)}^T X_{(n+1)} = X_{(n)}^T X_{(n)} + \vec{x}_{(1)} \vec{x}_{(1)}^T.$$

- For  $n < N$ ,  $X_{(n)}^T X_{(n)}$  has  $n$  non-zero eigenvalues, and  $N - n$  eigenvalues.
- Suggests an inductive construction of  $N \times N$  positive definite matrices  $\{A^{(n)}\}_{n=1, \dots, N}$ ,

$$A_{(n+1)} = \text{diag } A_{(n)} + \vec{x}_{1 \times m} \vec{x}_{1 \times m}^T$$

where  $A_{(0)} = [0]_{m \times m}$  and

$$x_j^2 \sim \Gamma[\beta/2, 1] \quad (j = 1, \dots, n), \quad x_j^2 \sim \Gamma[a\beta/2, 1] \quad (j = n + 1, \dots, N)$$

For this must study the eigenvalues of the  $N \times N$  matrix

$$Y := \text{diag}(a_1, \dots, a_n, \underbrace{a_{n+1}, \dots, a_{n+1}}_{N-n}) + \vec{x} \vec{x}^T$$



Since

$$\det(\lambda \mathbb{I}_N - Y) = \det(\lambda \mathbb{I}_N - A) \det(\mathbb{I}_N - (\lambda \mathbb{I}_N - A)^{-1} \vec{x} \vec{x}^T)$$

it follows

$$\frac{\det(\lambda \mathbb{I}_N - Y)}{\det(\lambda \mathbb{I}_N - A)} = 1 - \sum_{j=1}^n \frac{x_j^2}{\lambda - a_j} - \frac{\sum_{j=n+1}^N x_j^2}{\lambda - a_{n+1}}$$

## Question

What is the density of zeros of the random rational function

$$1 - \sum_{j=1}^{n+1} \frac{w_j}{\lambda - a_j}$$

for  $w_j \in \Gamma[s_j, 1]$ .

**Proposition 7.** *The zeros of the above rational function have PDF*

$$\begin{aligned} & \frac{1}{\Gamma(s_1) \cdots \Gamma(s_{n+1})} e^{-\sum_{j=1}^{n+1} (\lambda_j - a_j)} \prod_{1 \leq i < j \leq n+1} \frac{(\lambda_i - \lambda_j)}{(a_i - a_j)^{s_i + s_j - 1}} \\ & \times \prod_{i,j=1}^{n+1} |\lambda_i - a_j|^{s_j - 1} \end{aligned}$$

where

$$\lambda_1 > a_1 > \lambda_2 > \cdots > a_{n+1}.$$

The case of interest is

$$s_1 = \cdots = s_n = \beta/2, \quad s_{n+1} = (N - n)\alpha\beta/2, \quad a_{n+1} = 0.$$

## A recurrence relation

- Recall the case of interest is

$$s_1 = \cdots = s_n = \beta/2, \quad s_{n+1} = (N - n)a\beta/2, \quad a_{n+1} = 0.$$

- Denote the conditional PDF for  $\{\lambda_j\}$  in the case of interest by  $G(\{\lambda_j\}_{j=1,\dots,n+1}; \{a_j\}_{j=1,\dots,n})$ .
- Let the PDF of  $\{\lambda_j\}_{j=1,\dots,n+1}$  be denoted  $p_{n+1}(\{\lambda_j\})$ .
- Let the PDF of  $\{a_j\}_{j=1,\dots,n+1}$  be denoted  $p_n(\{a_j\})$ .

For  $n < N$  the recursive construction of  $\{A^{(n)}\}$  gives that

$$p_{n+1}(\{\lambda_j\}) = \int_{\lambda_1 > a_1 > \cdots > \lambda_{n+1} > 0} da_1 \cdots da_n \\ \times G(\{\lambda_j\}_{j=1,\dots,n+1}; \{a_j\}_{j=1,\dots,n}) p_n(\{a_j\})$$

subject to the initial condition  $p_0 = 1$ .

## A special solution

For  $\beta = 1$ , the recurrence has solution

$$p_n(\{\lambda_j\}) = \frac{1}{C_n} \prod_{l=1}^n \lambda_l^{(N-n-1)/2} e^{-\lambda_l} \prod_{1 \leq j < k \leq n} |\lambda_k - \lambda_j|$$

since this corresponds to the LOE.

For general  $\beta > 0$ , want to check that the recurrence has solution

$$p_n(\{\lambda_j\}) = \frac{1}{C_{n,\beta}} \prod_{l=1}^n \lambda_l^{(N-n+1)\beta/2-1} e^{-\lambda_l} \prod_{1 \leq j < k \leq n} |\lambda_k - \lambda_j|^\beta.$$

Since

$$\begin{aligned} & G(\{\lambda_j\}_{j=1,\dots,n+1}; \{a_j\}_{j=1,\dots,n}) \\ &= \frac{1}{(\Gamma(\beta/2))^n \Gamma((N-n)\beta/2)} e^{-\sum_{j=1}^n (\lambda_j - a_j) - \lambda_{n+1}} \frac{\prod_{i < j}^{n+1} (\lambda_i - \lambda_j)}{\prod_{i < j}^n (a_i - a_j)^{\beta-1}} \\ & \times \frac{\prod_{i=1}^{n+1} \lambda_i^{(N-n)\beta/2+1}}{a_i^{(N-n+1)\beta/2-1}} \prod_{i,j=1}^n |\lambda_i - a_j|^{\beta/2-1} \end{aligned}$$

we see we need to evaluate

$$\int_{\lambda_1 > a_1 > \dots > \lambda_{n+1} > 0} da_1 \cdots da_n \prod_{i < j}^n (a_i - a_j) \prod_{i,j=1}^n |\lambda_i - a_j|^{\beta/2-1}$$

According to the integration formula which follows from Proposition 5 this equals

$$\frac{(\Gamma(\beta/2))^{n+1}}{\Gamma((n+1)\beta/2)} \prod_{1 \leq j < k \leq n+1} (\lambda_j - \lambda_k)^{\beta-1}.$$

leaving us with

$$\frac{C_{n+1,\beta}}{C_{n,\beta}} \frac{\Gamma(\beta/2)}{\Gamma((n+1)\beta/2) \Gamma((N-n)\beta/2)} p_{n+1}(\{\lambda_j\}).$$