

Exercises: ① Show the formula for the joint distributions of eigenvalues in the $\beta=2$ and $\beta=4$ cases.

② Check the operator identity $CUC = \bar{U}$ where C is the complex conjugation.

③ Consider the "entropy" functional

$$S(P) = - \int p(H) \ln(p(H)) dH$$

Show that under the condition $\mathbb{E}\left(\frac{\text{Tr}(H^2)}{2N}\right) = \frac{n}{2}$,

where $n = N + \frac{\beta N(N-1)}{2}$ (degrees of freedom),

the measure with

$$p(H) = \frac{1}{Z_N} \exp\left(-\frac{\text{Tr}(H^2)}{2N}\right)$$

maximize $S(P)$.

Solutions of the exercises:

① For $\beta=2$, the strategy is the same as for $\beta=1$.

We have also in this case $\delta H = g \cdot \delta \tilde{H} \cdot g^{-1}$ with

$$\delta \tilde{H} = \delta A + [\tilde{g}^{-1} \delta g, \Lambda].$$

In components: $\delta \tilde{H}_{i,i} = \delta \lambda_i$ and

for $j > i$ $\delta(\text{Re } \tilde{H}_{ij}) = (\lambda_j - \lambda_i) \cdot \delta(\text{Re } R_{ij})$

and $\delta(\text{Im } \tilde{H}_{ij}) = (\lambda_j - \lambda_i) \cdot \delta(\text{Im } R_{ij})$

Thus, the Jacobian between \tilde{H} and (A, R) is

$$J = \det \begin{pmatrix} \begin{array}{c|c|c} 1 & & 0 \\ \hline & \ddots & \\ \hline & & 0 \end{array} & 0 & c \\ \hline & \lambda_1 - \lambda_2 & 0 \\ \hline 0 & & \lambda_{N-1} - \lambda_N \\ \hline 0 & c & \begin{array}{c} \lambda_1 - \lambda_2 \\ \vdots \\ \lambda_{N-1} - \lambda_N \end{array} \end{pmatrix}$$

$$= \left[\prod_{1 \leq i < j \leq N} (\lambda_i - \lambda_j) \right]^2.$$

For $\beta=4$, it is similar. Now, we have instead of only the real and imaginary parts, we have the 4 components of the basis of the quaternionic numbers.

② Let us see how they apply to functions: Ψ :

$$\begin{aligned} (CUC\Psi)(x) &= C \cdot \int u(x,y) \overline{\Psi(y)} dy = \int \bar{u}(x,y) \Psi(y) dy \\ &= (\bar{U}\Psi)(x). \end{aligned}$$

③ Consider the constraint

$$C = \int p(H) \text{Tr}(H^2) dH.$$

Let $B = -(1 + \ln(A))$ and λ be the Lagrange multiplier, i.e.,

$$S(p) = - \int p(H) \ln p(H) dH - \lambda \cdot \left(\int p(H) \text{Tr}(H^2) dH - C \right) + (\ln A + 1) \left(\int p(H) dH - 1 \right)$$

• let p_0 be the distribution which maximizes $S(p)$, then at first order in δp ,

$$\delta S(p_0) = S(p_0 + \delta p) - S(p_0) = 0, \text{ i.e.,}$$

$$-1 - \ln(p_0) - \lambda \cdot \text{Tr}(H^2) + 1 + \ln(A) = 0,$$

which implies $p_0(H) = A \cdot \exp(-\lambda \cdot \text{Tr}(H^2))$.

• The normalization constant fixes the value of A ,

$$A^{-1} = \int e^{-\lambda \text{Tr}(H^2)} dH \equiv q(\lambda).$$

• Thus, the second constraint writes

$$C = \frac{1}{q(\lambda)} \int \text{Tr}(H^2) \cdot e^{-\lambda \text{Tr}(H^2)} dH = - \frac{1}{q(\lambda)} \frac{dq(\lambda)}{d\lambda}$$

• let us get $q(\lambda)$ now. By the change of variable

$X = \sqrt{\lambda} H$, we obtain

$$q(\lambda) = \frac{1}{\lambda^{n/2}} \int e^{-\text{Tr}(X^2)} dX, \text{ where } n \text{ is the}$$

dimension of the space where the integral is made.

• The integral over dX is independent of λ , thus

$$\frac{dq(\lambda)}{d\lambda} = -\frac{n}{2\lambda} \cdot q(\lambda), \text{ and finally } \lambda = \frac{n}{2C}.$$

Since we want $\lambda = \frac{1}{2N}$, $C = N \cdot n$. This finishes the proof.