

• In any change of variables the number of variables before and after

should agree: $H = H^T$ has N diag + $\frac{N(N-1)}{2}$ (strictly) upper triag. degrees of freedom (dof)
 $= \frac{N(N+1)}{2}$

• X has N dof, O has N^2 dof. minus the number of constraints from $OO^T = \mathbb{1}_N = O^T O$: N diag and $\frac{N(N-1)}{2}$ off diag
 $\Rightarrow N^2 - N - \frac{N(N-1)}{2} = \frac{N^2 - N}{2}$ ok.

- note that O contains the orthonormalised eigenvectors of H

• dO is the Haar measure on group of orthogonal matrices, it is invariant under orthogonal transformations $O' = O_1 O O_1^T$ with $O_1 O_1^T = \mathbb{1}_N$ so $dO = dO'$

\Rightarrow we need to compute the integral $\int dO$ and the Jacobian

$$\left[J(\underline{x}, O) \approx \prod_{j>k}^N (x_j - x_k) \right] \leftarrow \text{to derive } g(x_1, \dots, x_N) \text{ incl. normalisation, this is independent of } O!$$

\Rightarrow we only need $\text{Vol}(dO) = \int dO = \frac{2^N}{\Gamma_N(\frac{N}{2})} \frac{N^{\frac{N}{2}}}{\pi^{\frac{N}{2}}}$ Exercise for $N=2$ and general N

with $\Gamma_m(a) = \frac{1}{\pi^{\frac{m(m-1)}{4}}} \prod_{i=1}^m \Gamma(a - \frac{i-1}{2})$

Theorem [Edelman] Let $H = H^T$ be a real sym. matrix with jpdf of matrix elements $P(H) = \phi(\text{Tr} H, \text{Tr} H^2, \dots, \text{Tr} H^m)$ invariant under conjugation by an orthogonal matrix. Then the jpdf of the ordered eigenvalues $x_1 \geq x_2 \geq \dots \geq x_N$ of H reads

$$S_{\text{ord}}(x_1, \dots, x_N) = \frac{1}{\Gamma_N(\frac{N}{2})} \frac{1}{2^{N/2} \pi^{N/4}} \underbrace{\phi\left(\sum_i x_i, \sum_i x_i^2, \dots, \sum_i x_i^m\right)}_{\text{invariant}} \prod_{j>k}^N (x_j - x_k)$$

\rightarrow in our example $P(H) = e^{-\text{Tr} H^2} \Rightarrow \phi\left[-\sum_{i=1}^N \lambda(x_i)\right]$

* How are jpdf of ordered and unordered eigenvalues related

Example $N=2$:

unordered

$$\int_{\mathbb{R}^2} dx_1 dx_2 e^{-V(x_1) - V(x_2)} |x_1 - x_2|$$

\mathbb{R}^2 \mathbb{R}^2 invariant under permutations $x_1 \leftrightarrow x_2$

$$= \int_{x_1 > x_2} dx_1 dx_2 \mathcal{S}_2(x_1, x_2) + \int_{x_1 < x_2} dx_1 dx_2 \mathcal{S}_2(x_1, x_2) = 2 \int_{-\infty}^{\infty} dx_1 \int_{-\infty}^{x_1} dx_2 \mathcal{S}_2(x_1, x_2)$$

Similarly one can show for general N that for $\mathcal{S}_N(x_1, \dots, x_N)$ invariant under any permutations of indices

$$\int_{\mathbb{R}^N} dx_1 \dots dx_N \mathcal{S}_N(x_1, \dots, x_N) = N! \int_{x_1 > x_2 > \dots > x_N} dx_1 \dots dx_N \mathcal{S}_N(x_1, \dots, x_N)$$

unordered ordered, here $\prod_{i < j}^N (x_i - x_j) = \prod_{i < j}^N (x_j - x_i)$

\Rightarrow taking all together for the GOE we have

$$P(H) = \prod_{i=1}^N \frac{N}{i} \frac{e^{-H_i^2/2}}{\sqrt{2\pi}} \prod_{i < j}^N \frac{e^{-H_i^2}}{\sqrt{2\pi}} = \frac{1}{(2\pi)^{\frac{N}{2}} \frac{N!}{2^{\frac{N(N-1)}{2}}}} e^{-\frac{1}{2} \sum_{i=1}^N x_i^2} = C_N e^{-\frac{1}{2} \text{tr} H^2}$$

and $\mathcal{S}(x_1, \dots, x_N) = \int dO |J(X, O)| C_N e^{-\frac{1}{2} \sum_{i=1}^N x_i^2}$

Edelman \Rightarrow

$$\int dO \frac{N!}{\prod_{i=1}^N (N-i)!} \prod_{i < j}^N (x_i - x_j) e^{-\frac{1}{2} \sum_{i=1}^N x_i^2}$$

ordered $x_1 > x_2 > \dots > x_N$

$$= N! \int dO \frac{N!}{\prod_{i=1}^N (N-i)!} \prod_{i < j}^N |x_i - x_j| e^{-\frac{1}{2} \sum_{i=1}^N x_i^2}$$

unordered

* note that when diagonalising $H = O \Sigma O^T$ the choice of eigenvectors is not unique (with $H \vec{v}_i = x_i \vec{v}_i$ also $-\vec{v}_i$ is an eigenvector)

\Rightarrow in $\int dO$ we integrate over all eigenvector modulo this change

\Rightarrow we have to divide by 2^N : $\frac{\int dO}{2^N} = \frac{2^{\frac{N(N-1)}{2}}}{\prod_{i=1}^N (N-i)!} \cdot 2^N$

\Rightarrow the desired normalisation $Z_{N, \beta=1} = \frac{1}{(2\pi)^{\frac{N}{2}}} \prod_{i=1}^N \frac{N}{i} \frac{\Gamma(N-i+1/2)}{\Gamma(1/2)}$ from p 3 \checkmark

7. The Vandermonde Determinant and Derivation of the JPF Part Co

• the Jacobian $J(\bar{x}, 0) \equiv \Delta_N(\{x\}) = \prod_{i>j}^N (x_i - x_j)$

Can be written as a determinant $= \det [x_j^{i-1}]_{i,j=1}^N = \begin{vmatrix} 1 & 1 & \dots & 1 \\ x_1 & x_2 & \dots & x_N \\ \vdots & \vdots & \dots & \vdots \\ x_1^{N-1} & x_2^{N-1} & \dots & x_N^{N-1} \end{vmatrix}$

Proof: Exercise (N=2 trivial)

7.1. Properties of the Vandermonde - determinant:

1) $\Delta_N(\{x\})$ is a totally antisymmetric polynomial in all x_1, \dots, x_N as any exchange of 2 columns in the det gives a minus

2) $\Delta_N(\{x\})$ contains $\frac{N(N-1)}{2}$ factors (which is easy to see by induction)

$\Delta_N(\{x\}) = \prod_{i=1}^{N-1} (x_N - x_i) \Delta_{N-1}(\{x\})$ only contains x_1, \dots, x_{N-1}

\Rightarrow rescaling all x_i by the same constant $x_i \rightarrow ax_i$ (e.g. $a = \sqrt{N}$) yields $\Delta_N(\{ax\}) = a^{\frac{N(N-1)}{2}} \Delta_N(\{x\})$

3) Linear Algebra properties of Δ_N :

multiples of rows can be added to each other without changing the value of the det:

$\Rightarrow \Delta_N(\{x\}) = \det \begin{vmatrix} 1 & 1 & \dots & 1 \\ x_1 & x_2 & \dots & x_N \\ \vdots & \vdots & \dots & \vdots \\ \frac{1}{N-1} (x_1) & \frac{1}{N-1} (x_2) & \dots & \frac{1}{N-1} (x_N) \end{vmatrix} = \dots = \det \begin{vmatrix} \frac{1}{N-1} (x_1) & \frac{1}{N-1} (x_2) & \dots & \frac{1}{N-1} (x_N) \\ \vdots & \vdots & \dots & \vdots \\ \frac{1}{N-1} (x_1) & \frac{1}{N-1} (x_2) & \dots & \frac{1}{N-1} (x_N) \end{vmatrix} = \frac{1}{a_0 a_1 \dots a_{N-1}} \det \begin{vmatrix} \frac{1}{N-1} (x_1) & \dots & \frac{1}{N-1} (x_N) \\ \vdots & \dots & \vdots \\ \frac{1}{N-1} (x_1) & \dots & \frac{1}{N-1} (x_N) \end{vmatrix}$

$\frac{1}{N-1} (x) = x^{N-1} + a_{N-2} x^{N-2} + \dots + a_1 x + a_0$ can obtain arbitrary poly $\frac{1}{N-1} (x) = a_N x^{N-1} + \dots$ by mult in and out, provided all leading coeff $\neq 0$

Example: for the (probabilist's) Hermite polynomials $H_n(x)$

orthogonal w.r.t $e^{-x^2/2}$: $\int_{-\infty}^{\infty} dx e^{-x^2/2} H_n(x) H_m(x) \sim \delta_{nm}$

we have $\Delta_N(x) = \begin{vmatrix} H_0(x_1) & \dots & H_0(x_N) \\ \vdots & & \vdots \\ H_{N-1}(x_1) & \dots & H_{N-1}(x_N) \end{vmatrix}$ as $H_n(x) = x^n + O(x^{n-1})$ are monic

remark: the Hermite polynomials $H_n(x)$ w.r.t e^{-x^2} , $\int_{-\infty}^{\infty} dx e^{-x^2} H_n(x) H_m(x) \sim \delta_{nm}$

yield prefactors as $H_n(x) = 2^{n/2} x^n + O(x^{n-1})$ are not monic, $a_n^{(n)} = 2^{n/2}$

$$\Delta_N(x) = \frac{1}{1 \cdot 2 \cdot \dots \cdot 2^{N-1}} \begin{vmatrix} H_0(x_1) & \dots & H_0(x_N) \\ \vdots & & \vdots \\ H_{N-1}(x_1) & \dots & H_{N-1}(x_N) \end{vmatrix}$$

7.2 Derivation of the jpdf Part II

Here we do not rely on the theorem by Edelman and thus on the invariance of $P(H)$ under rotations $H \rightarrow OHO^T$.

Consider an infinitesimal transformation of $H = O \Sigma O^T$ or differentiations

$$\Rightarrow dH = dO \Sigma O^T + O d\Sigma O^T + O \Sigma dO^T \quad \text{with } dO, d\Sigma, dO^T \text{ matrices}$$

using $\boxed{OO^T = \mathbb{I}_N = O^T O}$ we have $d(OO^T) = dO O^T + O dO^T = 0$

$$\text{i.e. } \underline{(dO O^T)^T = O dO^T = -dO O^T} = d(O^T O) = dO^T O + O^T dO$$

is antisymmetric, as is $\boxed{d\Sigma \equiv O^T dO}$

$$\Rightarrow dH = \underbrace{O O^T}_{d\Sigma} dO \Sigma O^T + O d\Sigma O^T + O \Sigma \underbrace{dO^T}_{d\Sigma^T = -d\Sigma} O^T$$

$$\underline{= O (d\Sigma \Sigma + d\Sigma - \Sigma d\Sigma) O^T = O (d\Sigma + [\Sigma, d\Sigma]) O^T} \quad d\hat{H}$$

because dH and $d\hat{H}$ are related by an

orthogonal trafo, it is sufficient to find the Jacobian for $d\hat{H}$: $\frac{\partial \hat{H}_i}{\partial (x_j, \Omega_{ij})}$

using that $\underline{X} = \text{diag}(x_1, \dots, x_N)$ and (since $d\underline{X}$ are diagonal) we have

$$d\hat{t}_{ij} = d\underline{X}_{ij} + (d\underline{\Omega}\underline{X} - \underline{X}d\underline{\Omega})_{ij} = dx_j \delta_{ij} + d\underline{\Omega}_{ij} x_j - x_i d\underline{\Omega}_{ij}$$

$$\Rightarrow \left[\frac{d\hat{t}_{ij}}{dx_k} = \delta_{ik} \delta_{ij}, \quad \frac{d\hat{t}_{ij}}{d\underline{\Omega}_{kl}} = \delta_{ik} \delta_{jl} (x_j - x_i) \right]$$

Example $N=3$: $\exists \underline{N(N-1)} = 6$ indep. dof.

$$d\underline{\Omega} = -d\underline{\Omega}^T \Rightarrow \text{parametrise } d\underline{\Omega} = \begin{pmatrix} 0 & d\underline{\Omega}_{21} & d\underline{\Omega}_{23} \\ -d\underline{\Omega}_{21} & 0 & d\underline{\Omega}_{33} \\ -d\underline{\Omega}_{23} & -d\underline{\Omega}_{33} & 0 \end{pmatrix} \text{ and } dx_1, dx_2, dx_3$$

$$\Rightarrow \begin{pmatrix} \frac{d\hat{t}_{21}}{d\underline{\Omega}_{21}} & \frac{d\hat{t}_{21}}{d\underline{\Omega}_{23}} & \frac{d\hat{t}_{21}}{d\underline{\Omega}_{33}} & \frac{d\hat{t}_{21}}{dx_1} & \frac{d\hat{t}_{21}}{dx_2} & \frac{d\hat{t}_{21}}{dx_3} \\ \frac{d\hat{t}_{12}}{d\underline{\Omega}_{21}} & \frac{d\hat{t}_{12}}{d\underline{\Omega}_{23}} & \frac{d\hat{t}_{12}}{d\underline{\Omega}_{33}} & \frac{d\hat{t}_{12}}{dx_1} & \frac{d\hat{t}_{12}}{dx_2} & \frac{d\hat{t}_{12}}{dx_3} \\ \frac{d\hat{t}_{32}}{d\underline{\Omega}_{21}} & \frac{d\hat{t}_{32}}{d\underline{\Omega}_{23}} & \frac{d\hat{t}_{32}}{d\underline{\Omega}_{33}} & \frac{d\hat{t}_{32}}{dx_1} & \frac{d\hat{t}_{32}}{dx_2} & \frac{d\hat{t}_{32}}{dx_3} \\ \frac{d\hat{t}_{13}}{d\underline{\Omega}_{21}} & \frac{d\hat{t}_{13}}{d\underline{\Omega}_{23}} & \frac{d\hat{t}_{13}}{d\underline{\Omega}_{33}} & \frac{d\hat{t}_{13}}{dx_1} & \frac{d\hat{t}_{13}}{dx_2} & \frac{d\hat{t}_{13}}{dx_3} \\ \frac{d\hat{t}_{23}}{d\underline{\Omega}_{21}} & \frac{d\hat{t}_{23}}{d\underline{\Omega}_{23}} & \frac{d\hat{t}_{23}}{d\underline{\Omega}_{33}} & \frac{d\hat{t}_{23}}{dx_1} & \frac{d\hat{t}_{23}}{dx_2} & \frac{d\hat{t}_{23}}{dx_3} \\ \frac{d\hat{t}_{31}}{d\underline{\Omega}_{21}} & \frac{d\hat{t}_{31}}{d\underline{\Omega}_{23}} & \frac{d\hat{t}_{31}}{d\underline{\Omega}_{33}} & \frac{d\hat{t}_{31}}{dx_1} & \frac{d\hat{t}_{31}}{dx_2} & \frac{d\hat{t}_{31}}{dx_3} \end{pmatrix}$$

$$= \begin{pmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ x_2 - x_1 & 0 & 0 & 0 & 0 & 0 \\ 0 & x_3 - x_1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & x_3 - x_2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

reorder columns to bring to identity form

$$- \text{we have to compute } \left| \det \left[\frac{\partial \hat{t}_{ij}}{\partial x_k \partial \underline{\Omega}_{lm}} \right] \right| = \begin{vmatrix} 1 & & & & & \\ & x_2 - x_1 & & & & \\ & & x_3 - x_1 & & & \\ & & & 1 & & \\ & & & & x_3 - x_2 & \\ & & & & & 1 \end{vmatrix}$$

$$= |(x_2 - x_1)(x_3 - x_1)(x_3 - x_2)|$$

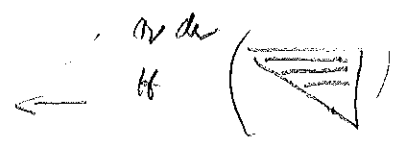
$$= |\Delta_3(x)|.$$

Also for general N such an ordering with diagonal Jacobian-determinant can be found

$$\left[\frac{dh_{11}}{dx_1} \right] \quad \frac{dh_{12}}{dx_1} \quad \frac{dh_{13}}{dx_1} \quad \frac{dh_{21}}{dx_1} \quad \frac{dh_{22}}{dx_1} \quad \frac{dh_{23}}{dx_1}$$

$$\frac{dh_{11}}{dx_{12}} \quad \left[\frac{dh_{12}}{dx_{12}} \right]$$

$$\frac{dh_{11}}{dx_{13}} \quad \frac{dh_{12}}{dx_{13}} \quad \left[\frac{dh_{13}}{dx_{13}} \right]$$



order

$$\begin{pmatrix} 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \end{pmatrix}$$

$$\frac{dh_{11}}{dx_{11}}$$

$$\left[\frac{dh_{11}}{dx_{11}} \right]$$

$$\frac{dh_{11}}{dx_2}$$

$$\left[\frac{dh_{22}}{dx_2} \right] \quad \frac{dh_{23}}{dx_2}$$

$$\frac{dh_{11}}{dx_{23}}$$

$$\left[\frac{dh_{23}}{dx_{23}} \right]$$

$$\frac{dh_{11}}{dx_{21}}$$

$$\frac{dh_{11}}{dx_3}$$

$$\frac{dh_{11}}{dx_{31}}$$

$$\left[\frac{dh_{33}}{dx_3} \right]$$

with the result that $|\det(\frac{\partial h}{\partial x_i})| = |\Delta_n(\{x_i\})|$