

7.4. Random matrix theory

conjecture: level statistics of classically chaotic systems

is well described by random matrix theory (RMT)

Casati, Vals-Gris, Guarneri 1980; Bohigas, Giannoni, Schmit 1983

remarks: . lots of numerical & experimental evidence

. semiclassical "proof" \rightarrow semiclassics

. intuition: no system specific information except symmetries

3 universality classes:

Systems with time reversal symmetry: $\beta = 1$

Systems without time reversal symmetry: $\beta = 2$

Spin systems (4×4 matrix) : $\beta = 4$

\Rightarrow 3 groups of transformations

$O(N)$ group of real orthogonal matrices

$U(N)$ " " unitary "

symplectic group

$$O^T O = \mathbb{1}$$

$$U^\dagger U = \mathbb{1}$$

7.4.1. Gaussian Ensembles

2x2 matrix, $\beta=1$: H real hermitean: $H = \begin{pmatrix} H_{11} & H_{12} \\ H_{12} & H_{22} \end{pmatrix}$

aim: find prob. dist. $\mathcal{P}(H) = \mathcal{P}(H_{11}, H_{22}, H_{12})$

under conditions:

1. normalization: $\int_{-\infty}^{\infty} dH_{11} \int_{-\infty}^{\infty} dH_{22} \int_{-\infty}^{\infty} dH_{12} \mathcal{P}(H) = 1$

2. invariance of $\mathcal{P}(H)$ under $O(2)$: Let $H' = O^T H O$ with $O^T O = \mathbb{1}$
 $\Rightarrow \mathcal{P}(H') = \mathcal{P}(H)$

3. independence: $\mathcal{P}(H) = \mathcal{P}_{11}(H_{11}) \cdot \mathcal{P}_{22}(H_{22}) \cdot \mathcal{P}_{12}(H_{12})$

sufficient: infinitesimal orthogonal transformation

$$O = \begin{pmatrix} 1 & \Theta \\ -\Theta & 1 \end{pmatrix} \quad \text{with } \Theta \ll 1 \quad \text{and neglect terms } O(\Theta^2)$$

$$H' = O^T H O = \begin{pmatrix} 1 & -\Theta \\ \Theta & 1 \end{pmatrix} \begin{pmatrix} H_{11} & H_{12} \\ H_{12} & H_{22} \end{pmatrix} \begin{pmatrix} 1 & \Theta \\ -\Theta & 1 \end{pmatrix}$$

$$= \dots \approx \begin{pmatrix} H_{11} + \Theta(-2H_{12}) & H_{12} + \Theta(H_{11} - H_{22}) \\ H_{12} + \Theta(H_{11} - H_{22}) & H_{22} + \Theta(2H_{12}) \end{pmatrix}$$

$$P_n(H'_n) = P_n(H_{11} + \Theta(-2H_{12})) = P_n(H_{11}) + \Theta(-2H_{12}) \frac{dP_n}{dH_{11}} + O(\Theta^2)$$

$$= P_n(H_{11}) \left(1 + \Theta(-2H_{12}) \cdot \frac{1}{P_n(H_{11})} \frac{dP_n}{dH_{11}} \right) \frac{d \ln P_n}{dH_{11}}$$

$$P_{22}(H'_{22}) = P_{22}(H_{22}) \left(1 + \Theta \cdot 2H_{12} \frac{d \ln P_{22}}{d H_{22}} \right)$$

$$P_{12}(H'_{12}) = P_{12}(H_{12}) \left(1 + \Theta \cdot (H_{11} - H_{22}) \frac{d \ln P_{12}}{d H_{12}} \right)$$

multiply:

$$P(H') \stackrel{3.}{=} P(H) \left(1 - \Theta \cdot \underbrace{\left[2H_{12} \left(\frac{d \ln P_{11}}{d H_{11}} - \frac{d \ln P_{22}}{d H_{22}} \right) - (H_{11} - H_{22}) \frac{d \ln P_{12}}{d H_{12}} \right]}_{\Rightarrow 0} \right)$$

$$\stackrel{2.}{=} P(H)$$

$$\Rightarrow \frac{1}{H_{12}} \frac{d \ln P_{12}}{d H_{12}} = \frac{2}{H_{11} - H_{22}} \left(\frac{d \ln P_{11}}{d H_{11}} - \frac{d \ln P_{22}}{d H_{22}} \right) = a \quad \text{some constant}$$

$$\Rightarrow \frac{d \ln P_{11}}{d H_{11}} - \frac{a}{2} H_{11} = \frac{d \ln P_{22}}{d H_{22}} - \frac{a}{2} H_{22} = b \quad \text{some constant}$$

$$\frac{d \ln P_{12}}{d H_{12}} = a H_{12} \Rightarrow \ln P_{12} = \frac{a}{2} H_{12}^2 + c \Rightarrow P_{12}(H_{12}) = c_{12} e^{\frac{a}{2} H_{12}^2}$$

$$\frac{d \ln P_{11}}{d H_{11}} = \frac{a}{2} H_{11} + b \Rightarrow P_{11}(H_{11}) = c_{11} e^{\frac{a}{4} H_{11}^2 + b H_{11}}$$

$$\Rightarrow P_{22}(H_{22}) = c_{22} e^{\frac{a}{4} H_{22}^2 + b H_{22}}$$

$$\Rightarrow P(H) = C e^{\frac{a}{4} (H_{11}^2 + H_{22}^2 + 2H_{12}^2) + b(H_{11} + H_{22})}$$

normalization: $a < 0$

choose zero of energy: $0 \stackrel{!}{=} \langle E_1 + E_2 \rangle = \langle H_{11} + H_{22} \rangle = 2 \langle H_{11} \rangle \Rightarrow b = 0$

$$P(H) = C e^{-\frac{1}{\epsilon^2} T_V(H^2)}$$

↑
normalization

↑
 ϵ energy scale

Gaussian Orthogonal Ensemble (GOE)

remarks:

- same formula for $N \times N$ matrix
- " " " unitary transformation (GUE)
- " " " symplectic " (GSE)

• explicitly for GOE: $\text{Tr}(H^2) = \sum_{i=1}^N H_{ii}^2 + \sum_{i < j}^N 2H_{ij}^2$

compare with Gaussian: $P(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{x^2}{2\sigma^2}}$ with variance σ^2

$\Rightarrow H_{ii}$: Gaussian, mean zero, variance $\langle H_{ii}^2 \rangle = \frac{\epsilon^2}{2}$

$\Rightarrow H_{ij}$: Gaussian, mean zero, variance $\langle H_{ij}^2 \rangle = \frac{\epsilon^2}{4}$

numerics: K_{ij} with $\sigma^2 = \frac{\epsilon^2}{8}$

$H_{ij} := K_{ij} + K_{ji}$

7.4.2. Eigenvalue distribution

aim: find $P(E_1, E_2)$ for GOE: real symmetric 2×2 matrix

$$\begin{pmatrix} H_{11} & H_{12} \\ H_{12} & H_{22} \end{pmatrix} = O^T \begin{pmatrix} E_1 & 0 \\ 0 & E_2 \end{pmatrix} O \quad \text{with } O = \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}$$

$$H_{11}, H_{22}, H_{12} \xrightarrow{\quad} E_1, E_2, \theta$$

Jacobi determinant: $J = \det \frac{\partial (H_{11}, H_{22}, H_{12})}{\partial (E_1, E_2, \theta)} = \dots = E_1 - E_2$

$$\Rightarrow P(E_1, E_2) = C |E_1 - E_2| e^{-\frac{1}{2}(E_1^2 + E_2^2)}$$

general result for $N \times N$ matrix and all symmetry classes β ($= 1, 2, 4$)

$$P(E_1, \dots, E_N) = C \cdot \left(\prod_{\substack{\mu, \nu=1 \\ \mu < \nu}}^N |E_\mu - E_\nu|^\beta \right) e^{-\frac{1}{2} \sum_{\mu=1}^N E_\mu^2}$$

7.4.3. Nearest neighbor level spacing distribution $P(S)$

$$\begin{aligned}
 N=2: P(S) &= C \int_{-\infty}^{\infty} dE_1 \int_{-\infty}^{\infty} dE_2 \delta(S - |E_1 - E_2|) |E_1 - E_2|^\beta e^{-\frac{1}{\varepsilon^2} (E_1^2 + E_2^2)} \\
 &\quad \downarrow \\
 &\quad \Delta = E_1 - E_2 \\
 &= C \int_{-\infty}^{\infty} d\Delta \int_{-\infty}^{\infty} dE_2 \delta(S - |\Delta|) |\Delta|^\beta e^{-\frac{1}{\varepsilon^2} (\Delta + E_2)^2 - E_2^2} \\
 &\quad \underbrace{\int_{-\infty}^{\infty} dE_2 \delta(S - |\Delta|) |\Delta|^\beta e^{-\frac{1}{\varepsilon^2} (\Delta + E_2)^2 - E_2^2}}_{\sqrt{\frac{\pi}{2}} \varepsilon e^{-\frac{\Delta^2}{2\varepsilon^2}}} \\
 &= 2C \varepsilon \sqrt{\frac{\pi}{2}} S^\beta e^{-\frac{S^2}{2\varepsilon^2}}
 \end{aligned}$$

normalization: $\int_0^\infty dS P(S) = 1 \Rightarrow C(\beta)$

average spacing: $\int_0^\infty dS S P(S) = 1 \Rightarrow \varepsilon(\beta)$

remarks: • agrees with $P(S) \sim S^\beta$ for $S \rightarrow 0$

• $N \times N$ matrix with $N \rightarrow \infty$ gives $\approx 1\%$ correction

Wigner surmise (2x2):

GOE $\beta=1$:

$$P(S) = \frac{1}{2} S e^{-S/2} S^2$$

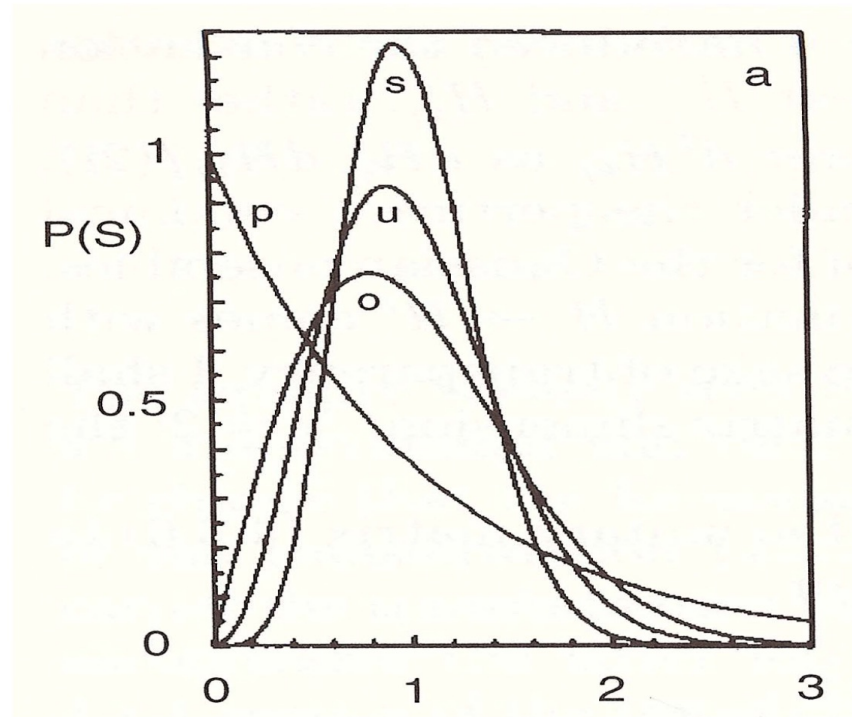
GUE $\beta=2$:

$$P(S) = \frac{32}{\pi^2} S^2 e^{-S/4} S^2$$

GSE $\beta=4$:

$$P(S) = \frac{2^{18}}{3^6 \pi^3} S^4 e^{-S/6} S^2$$

P(S) for 2x2 matrix: GOE, GUE, GSE



Haake, Fig 4.1

Difference in $P(S)$ for $N \times N$ matrix with $N \rightarrow \infty$ and $N=2$

