


Monotonicity of entropy for random matrices

Based on joint works with Benjamin Dadoun
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Artstein-Ball-Barthe-Naor '04:

X r.v., $E X = 0$, $E X^2 = 1$.

X has density f ; $Ent(X) = -E \log f(X)$
 $= -\int f \log f$

$S_m = X_1 + \dots + X_m$; $(X_i)_i$ i.i.d. copies.

$$Ent\left(\frac{S_m}{\sqrt{m}}\right) \leq Ent\left(\frac{S_n}{\sqrt{n}}\right) \quad \forall n \leq m$$

Remark:

• Conservation law: system centered
unit variance.

• $Z \sim N(0,1) \rightsquigarrow$ maximizes Entropy among

continuous centered, unit variance
n.v.

Ent \Leftrightarrow rate fct^o in LDP.

Answers Q. of Shannon '40.

Stam '59, Lieb '78 : monotonicity for $n=2^k$.

Related quantities:

Score function : $f_X = \frac{f'_X}{f_X}$

Fisher information : $\phi_X = \mathbb{E} f_X^2$

De Bruijn identity:

$$\text{Ent}(X) = \frac{\log 2\pi e}{2} + \frac{1}{2} \int_0^\infty \left(\frac{1}{1+t} - \phi(X + \sqrt{t}Z) \right) dt$$

Covtade's proof of monotonicity of $\phi_{\frac{S_m}{\sqrt{m}}}$

Take $n \leq m$.

$$p_{S_m} = p_{\underbrace{S_n}_{\text{indep}} + \underbrace{S_m - S_n}_{\text{indep}}} = \mathbb{E}[p_{S_m} | S_m]$$

$$\phi_{S_m} = \mathbb{E} p_{S_m}^2 = \mathbb{E} \left[p_{S_m} \cdot \mathbb{E}[p_{S_m} | S_m] \right]$$

$$= \text{cov}(g(S_n), h(S_m))$$

Dembo-Kagan-Shepp '01:

$$X, Y \text{ r.v.}, \quad R(X, Y) = \sup_{g, h} \frac{\text{cov}(g(X), h(Y))}{\sigma_{g(X)} \sigma_{h(Y)}}$$

$$R(S_n, S_m) = \sqrt{\frac{n}{m}} \quad \forall n \leq m$$

$$\rightarrow \leq \sqrt{\frac{n}{m}} \sqrt{\phi_{S_n}} \sqrt{\phi_{S_m}}$$

$$\hookrightarrow m \phi_{S_m} \leq n \phi_{S_n} \Rightarrow \phi_{\frac{S_m}{\sqrt{m}}} \leq \phi_{\frac{S_n}{\sqrt{n}}} \quad \square$$

In Free Probability.

classical Prob

Free Probability

(Ω, \mathbb{P})

(A, τ)

distributions

$$\tau(a^k) = \int a^k d\mu_a$$

Indep

freeness

Gaussian

semi-circle s

Entropy

Free entropy

$$\chi^*(\mu) = \iint \log|x-y| d\mu(x) d\mu(y)$$

Score fct.

conjugate variable ξ

$$\xi(\xi) \in L^2(\mathfrak{A})$$

Fisher information

De Bruijn

$$\tau\left(\hat{f}(z) \cdot z^{\alpha}\right) = \sum_{k=0}^{\alpha-1} \tau\left(\frac{f^{(k)}}{z^k}\right) \tau\left(\frac{f^{(\alpha-k)}}{z^{\alpha-k}}\right)$$

Free Fisher

$$\Phi^{\alpha}(z) = \|\hat{f}(z)\|_2^2$$

replace Z by s .

Shlyakhtenko '07:

$(a_i)_i$ free NC α -v
identically distributed

$$S_n = a_1 + \dots + a_n$$

$$\chi^{\alpha}\left(\frac{S_n}{\sqrt{n}}\right) \leq \chi^{\alpha}\left(\frac{S_m}{\sqrt{m}}\right) \quad \forall n \leq m$$

(Shlyakhtenko-Tao '21: Fractional free convolution)

Dadoun-Y '21: Another proof.

Correlation coefficient:

$$\text{Cov}(a, b) = \mathcal{E} \left((a - \mathcal{E}(a)) \cdot (b - \mathcal{E}(b)) \right)$$

$$R(M_1, M_2) = \sup_{\substack{x \in M_1 \\ y \in M_2}} \frac{\text{Cov}(x, y)}{\sigma_x \cdot \sigma_y}$$

Down - Y'21: $(a_i)_i$ free identically dist

$$S_m = a_1 + \dots + a_m$$

$$R(S_n, S_m) = \sqrt{\frac{n}{m}} \quad \forall n \leq m$$

proof: assume n divides m .

ξ polynomial in S_m

ξ' in S_m

Denote

$$\text{Proj}_I = \mathcal{E} \left[(a_i)_{i \in I} \right]$$

$$\langle \xi, \xi' \rangle = \langle \xi, \text{Proj}_{\{1, \dots, n\}} \xi' \rangle \stackrel{\text{c.s.}}{\leq} \|\xi\|_2 \cdot \|\text{Proj}_{[n]} \xi'\|_2$$

$$\mathcal{E}(\xi) = \mathcal{E}(\xi') = 0$$

$$z' = \sum_{j=1}^m \text{Proj}_{\{1, \dots, j\}} z' - \text{Proj}_{\{1, \dots, j-1\}} z'$$

$$\|z'\|_2^2 = \sum_{j=1}^m \left\| \text{Proj}_{\{1, \dots, j\}} z' - \text{Proj}_{\{1, \dots, j-1\}} z' \right\|_2^2$$

$$\left\| \text{Proj}_{[n]} z' \right\|_2^2 = \sum_{l=1}^m \left\| \text{Proj}_{\{1, \dots, l\}} z' - \text{Proj}_{\{1, \dots, l-1\}} z' \right\|_2^2$$

$$= \frac{1}{\frac{m}{m}} \sum_{k=0}^{\frac{m}{m}-1} \sum_{l=1}^m \left\| \text{Proj}_{\{k m + 1, \dots, k m + l\}} z' - \text{Proj}_{\{k m + 1, \dots, k m + l-1\}} z' \right\|_2^2$$

Idempotence $\Rightarrow \text{Proj}_I \circ \text{Proj}_J = \text{Proj}_{I \cap J}$

$$= \frac{m}{m} \sum_{k=0}^{\frac{m}{m}-1} \sum_{l=1}^m \left\| \text{Proj}_{\{k m + 1, \dots, k m + l\}} \left(\text{Proj}_{\{1, \dots, k m + l\}} z' - \text{Proj}_{\{1, \dots, k m + l-1\}} z' \right) \right\|_2^2$$

$$\leq \frac{m}{m} \|z'\|_2^2$$

Entropy in RMT:

X_N $N \times N$ random matrix
i.i.d entries, centered, unit variance.

Wigner semi-circle theorem:

$$W_N = \frac{X_N + X_N^*}{\sqrt{2}}$$

$$\mathbb{E} \mu_{\frac{1}{\sqrt{N}} W_N} = \frac{1}{N} \sum_{i=1}^N \delta_{\lambda_i(\frac{1}{\sqrt{N}} W_N)} \xrightarrow[N \rightarrow \infty]{\text{a.s.}} \mu_{\text{s.c}}$$

Girko's circular law:

$$\mathbb{E} \mu_{\frac{1}{\sqrt{N}} X_N} \xrightarrow[N \rightarrow \infty]{\text{a.s.}} \mu_{\text{circ}}$$

Marchenko-Pastur (square case)

$$\mathbb{E} \mu_{\frac{1}{N} X_N X_N^a} \xrightarrow[N \rightarrow \infty]{\text{a.s.}} \mu_{\text{MP}}$$

What is entropy?

- ① rate fct. in LDP
- ② optimizer of which functional?

Take GUE

↳ distributed on H_N according

$$e^{-\frac{N}{2} \text{Tr}(A^2)} dA$$

↳ joint law of eig

$$C_N e^{-\frac{N}{2} \sum_{i=1}^N \lambda_i^2} \prod_{i < j} |\lambda_i - \lambda_j|^2$$

$$\mathbb{P}(\mu_N \approx \nu) = \int_{d\mu_N \approx \nu} e^{-\frac{N}{2} \sum_{i=1}^N \lambda_i^2} \prod_{i < j} |\lambda_i - \lambda_j|^2$$

$$= c_N \int_{\{\mu_N \approx \nu\}} e^{-N^2 \left(\frac{1}{2N} \sum_{i=1}^N \lambda_i^2 - \frac{1}{N^2} \sum_{i,j} \log |\lambda_i - \lambda_j| \right)}$$

$$\approx c_N e^{-N^2 \left(\underbrace{\frac{1}{2} m_2(\nu)}_{\text{second moment}} - \iint \log |x-y| d\nu(x) d\nu(y) \right)}$$

Ben Arous-Guionnet '97: made this rigorous.

$$\mathcal{E}(\mu) = -\chi^2(\mu) = -\iint \log |x-y| d\mu(x) d\mu(y)$$

$$\mu_{sc} = \operatorname{argmin}_{m_2(\mu)=1} \mathcal{E}(\mu)$$

$$= \operatorname{argmin} \mathcal{E} \left(\frac{1}{\sqrt{m_2(\mu)}} \mu \right)$$

$$= \operatorname{argmin} \left(\mathcal{E}(\mu) + \frac{1}{2} m_2(\mu) \right)$$

$$\downarrow \\ \epsilon_1^{(2)}(\mu)$$

Question 1: Is $\epsilon_1^{(2)}$ monotone (in terms of matrix dimension) along Wigner's th?

Question 2: Is $\epsilon_{\frac{1}{2}}^{(2)}$ monotone in Girkko's th?

Q 3: Is $\epsilon^{(1)}$ monotone in MP th?

Results:

Let X_N^{Gin} $N \times N$ (complex) Gaussian matrix properly normalized

$$X_N^{\text{GUE}} = \frac{X_N^{\text{Gin}} + (X_N^{\text{Gin}})^a}{2}$$

$$X_N^{\text{LUE}} = X_N^{\text{Gin}} X_N^{\text{Gin}^a}$$

Chafri-Dadoun - Y'22:

$$\sum_1^{(2)} \left(\mathbb{E} \mu_{X_N^{\text{GUE}}} \right) = \frac{3}{4} + \frac{1}{2} \left(\log N + \gamma + \frac{1}{2N} - H_N \right)$$

$\gamma =$ Euler constant

$$H_N = \sum_{i=1}^N \frac{1}{i} \rightarrow N\text{-th harmonic nb.}$$

→ decreasing in N .

Chafri-Dadoun - Y'22:

$$\sum_{\frac{1}{2}}^{(2)} \left(\mathbb{E} \mu_{X_N^{\text{Gin}}} \right) = \frac{3}{4} + \frac{1}{2} \left(\log N + \gamma + \frac{1}{2N} - H_N \right) + \frac{1}{2} \sum_{k=N+1}^{\infty} \frac{4^{-k} \binom{2k}{k}}{k(k-1)}$$

$$\sum^{(1)} \left(\mathbb{E} \mu_{X_N^{\text{LUE}}} \right) = \frac{3}{2} + \left(\log N + \gamma + \frac{1}{2N} - H_N \right).$$

Open problems:

• Prove/disprove Q 1, 2, 3.

• Prove it for $N=2^k$.

Wigner

$$\left(\begin{array}{c|c} \text{Wigner} & \\ \hline & \text{Wigner} \end{array} \right)$$

• Explain $\sum_1^{(2)}(\text{GUE}) = \sum_{\frac{1}{2}}^{(2)}(\text{Gin})$

$$\left(\begin{array}{c|c} 0 & \text{Gin} \\ \hline \text{Gin} & 0 \end{array} \right)$$

↑

• $2 \sum_1^{(2)}(\text{GUE}) = \sum^{(1)}(\text{LUE})$.

$$+ \frac{1}{2} \sum_{k=N+1}^{\infty} \frac{4^{-k} \binom{2k}{k}}{k(k-1)}$$

