

Free Probability

L1

A, B are $N \times N$ Hermitian matrices.

We are interested in the eigenvalues of the sum $A+B$.

Not sufficient to know eig. of A and B separately, one needs also the relative position of the eigenvectors of A and

B : for example the decomposition of B -eigenvectors in the eigenbasis of A .

Define a non-standard addition of matrices that depends only on the eigenvalues of individual matrices in the sum and not on their eigenvectors

$$A \boxplus B = U^T A U + V^T B V$$

where U, V are independent unitary matrices, distributed according to the uniform measure on the unitary group.



$A \boxplus B$ is a random matrix
(even if A and B are deterministic/fixed)

The transformation $A \rightarrow U^T A U$
 $B \rightarrow V^T B V$

preserve the eigenvalues of the individual terms, but uniformly randomize their eigenvectors.

Observation

$$U^T A U + V^T B V = U^T (A + W^T B W) U$$

$W = VU^{-1}$ also Haar unitary

It is sufficient to randomize the eigenvectors of one matrix against the other.

More general setting

A, B are independent random matrices themselves.

Can one determine the eigenvalue density $\rho_{A \oplus B}(\lambda)$ from the eigenvalue densities of A and B separately ($\rho_A(\lambda)$ and $\rho_B(\lambda)$)?

The answer is affirmative for $N \rightarrow \infty$.

FREENESS (~~FREE ADDITION~~)
 (generalization of statistical independence to non-commutative objects)

$N \rightarrow \infty$
 A and B statistically independent
 Eigenvectors are in generic position.

Theorem (Sufficient condition for freeness)

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Let $N \times N$ random matrices A_N and B_N such that:

- A_N, B_N have an asymptotic eigenvalue density for $N \rightarrow \infty$.
- A_N, B_N are independent
- B_N is a unitarily invariant ensemble

Then A_N and B_N are asymptotically free.

$$A + B \iff A \boxplus B$$

if B is rotationally invariant.

Classical (commutative) probability

X_1, \dots, X_L independent r.v. drawn from $p_{X_i}(x)$.

$$S = X_1 + X_2 + \dots + X_L$$

$$\varphi_X(t) = \langle e^{itx} \rangle = \int_{-\infty}^{+\infty} dx p_X(x) e^{itx}$$

$$\boxed{\varphi_S(t) = \varphi_{X_1}(t) \varphi_{X_2}(t) \dots \varphi_{X_L}(t)}$$

$\varphi_X(t)$ is the generating function of moments

$$m_X^{(n)} = \int_{-\infty}^{+\infty} dx p_X(x) x^n$$

via

$$\varphi_X(t) = \int_{-\infty}^{+\infty} dx p_X(x) e^{itx} = \int_{-\infty}^{+\infty} dx p_X(x) \sum_{n=0}^{\infty} \frac{(it)^n}{n!} x^n$$

$$= \sum_{n=0}^{\infty} \frac{(it)^n}{n!} m_X^{(n)} \Rightarrow m_X^{(n)} = \frac{1}{i^n} \frac{d^n}{dt^n} \varphi_X(t) \Big|_{t=0}$$

If we take the log of $\varphi_X(t)$

$$g_{X_i}(t) \equiv \log \varphi_{X_i}(t)$$

$$g_S(t) = g_{X_1}(t) + g_{X_2}(t) + \dots + g_{X_L}(t)$$

$g_X(t)$ is the generating function of cumulants.

$$K_S^{(n)} = K_{X_1}^{(n)} + K_{X_2}^{(n)} + \dots + K_{X_L}^{(n)}$$

What is the analogue ~~concept~~ for matrices?

$$G_A(z) = \left\langle \frac{1}{N} \text{Tr} \frac{1}{z - A} \right\rangle = \int \frac{p_A(\lambda) d\lambda}{z - \lambda}$$

Resolvent or Green's function

A. Zeil, "Law of addition in random matrix theory"

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Nucl. Phys. B 476, 726 (1996)

[36-37]
Handout

"For the sake of convenience, we may, with due respect to Green, somewhat fancifully define a "Blue's function" ...

$$G_A^{-1}(B_A(z)) = B_A(G_A(z)) = z$$

Functional inverse of the Green's function.

Remarkably, the Blue's function satisfies the following addition rule

$$B_{H_1 + \dots + H_L}(z) = B_{H_1}(z) + B_{H_2}(z) + \dots + B_{H_L}(z) + \frac{1-L}{z}$$

for free Hermitian matrices $\{H_1, \dots, H_L\}$.

↓

It becomes additive if we define the R-transform

$$R_A(z) = B_A(z) - \frac{1}{z}$$

↓

$$R_{H_1 + \dots + H_L}(z) = R_{H_1}(z) + \dots + R_{H_L}(z)$$

Free Prob.
R-transform

Class. Prob.

L6

g (cumulant
generating
function).

Remarks

①

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$$B_A(G_A(z)) = z \implies R_A(G_A(z)) + \frac{1}{G_A(z)} = z$$

$$G_A(z) = \frac{1}{z - R_A(G_A(z))}$$

↓
Dyson-Schwinger eq.

$R_A(G_A(z))$ is the self-energy
 $\Sigma(z)$ of physicists.....

② R-transform is the generating function of free cumulants

$$R(z) = \sum_{n=1}^{\infty} \kappa_n z^{n-1}$$

(Trace) Moments : $m_n(A) = \left\langle \frac{1}{N} \text{Tr} A^n \right\rangle$

free cum. moments

$$K_1 = m_1(A)$$

$$K_2 = m_2(A) - [m_1(A)]^2$$

$$K_3 = m_3(A) - 3m_2(A)m_1(A) + 2[m_1(A)]^3$$

$$K_4 = \dots$$

different.

as
in
classical
probability

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Example :

In classical probability

$$p(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

has all cumulants = 0 except $K_2 = 1$.

What is the corresponding RM eigenvalue density that has all free cumulants = 0 except $K_2 = 1$?

$$R(z) = \sum_{n=1}^{\infty} K_n z^{n-1} = z.$$

So from $G_A(z) = \frac{1}{z - R_A(G_A(z))} = \frac{1}{z - G_A(z)}$

↓

$$G(z) = \frac{1}{2}(z - \sqrt{z^2 - 4}) \Rightarrow \left\{ p(\lambda) = \frac{1}{2\pi} \sqrt{4 - \lambda^2} \right\}$$

semicircle

Semicircle law is stable with respect to free addition in the same way as the Gaussian law is stable with respect to addition in classical probability. L8

Examples.

① Addition of a deterministic matrix and a random matrix

$$H = H_d + M$$

$$P(M) = \frac{1}{Z} e^{-N \text{Tr} \underbrace{V[M]}_{\frac{1}{2} M^2}}$$

$$G(z) = \frac{\downarrow 1}{z - G(z)}$$

$$z = \frac{\downarrow}{B(z) - z}$$

$$B(z) - z = \frac{1}{z}$$

$$\boxed{B(z) = z + \frac{1}{z}}$$

The Blue's function has a simpler form than $G(z)$ itself.

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$$H_d \begin{cases} c \mathbb{1} \rightarrow G_{H_d}(z) = \frac{1}{N} \text{Tr} \frac{1}{(z-c)\mathbb{1}} = \frac{1}{z-c} \Rightarrow z = \frac{1}{B(z)-c} \\ \text{diag}(\epsilon_1, \dots, \epsilon_N) \rightarrow G_{H_d}(z) = \frac{1}{N} \text{Tr} \frac{1}{z\mathbb{1} - \text{diag}(\epsilon_1, \dots, \epsilon_N)} \end{cases}$$

\Downarrow
 $B(z) = c + \frac{1}{z}$

$$= \frac{1}{N} \sum_{i=1}^N \frac{1}{z - \epsilon_i}$$

$$\Downarrow$$

$$z = \frac{1}{N} \sum_{i=1}^N \frac{1}{B(z) - \epsilon_i} \quad \text{implicitly.}$$

$$B_H(z) = B_{H_d}(z) + B_M(z) - \frac{1}{z} \quad \text{law of addition.}$$

$$= B_{H_d}(z) + z + \frac{1}{z} - \frac{1}{z} \quad (*)$$

To find $G_H(z)$, we use $B_H(G_H(z)) = z$, or using (*)

$$B_H(g_H(z)) = B_{H_d}(g_H(z)) + g_H(z)$$

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$$B_{H_d}(g_H(z)) = z - G_H(z)$$

Applying $G_{H_d}(\cdot)$ to both sides :

$$\begin{aligned} G_H(z) &= G_{H_d}(z - G_H(z)) \\ &= \frac{1}{N} \sum_{i=1}^N \frac{1}{z - \epsilon_i - G_H(z)} \end{aligned}$$

$N \rightarrow \infty$

$$G_H(z) = \int d\varepsilon \frac{\rho(\varepsilon)}{z - \varepsilon - G_H(z)}$$

"Pastur equation" (derived by other means in

L. Pastur, Theor. Math. Phys.

10, 67 (1972) see HO 38-39]

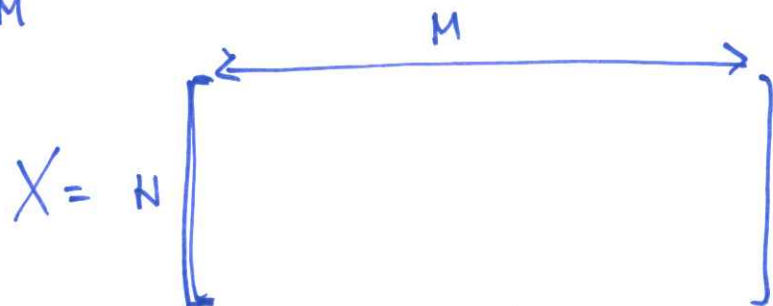
Check : if $\rho(\varepsilon) = \delta(0) \Leftarrow H_d$ is not there

$$G_H(z) = \frac{1}{z - G_H(z)} \quad \text{standard Gaussian equation}$$

Sum of GOE and Wishart random matrices.

LM

$$W = \frac{1}{M} X \cdot X^T$$



$$X_{ij} \sim \mathcal{N}(0, 1)$$

$r = \frac{N}{M}$ fixed as $N, M \rightarrow \infty$.

$$P(\lambda_1, \dots, \lambda_N) \propto e^{-\frac{M}{2} \sum_{i=1}^N \lambda_i} \prod_{i=1}^N \lambda_i^{\frac{1}{2}(M-N-1)} \prod_{j < k} |\lambda_j - \lambda_k|.$$

$\{\lambda_i\}$ are all positive.

Using the resolvent method (careful!)

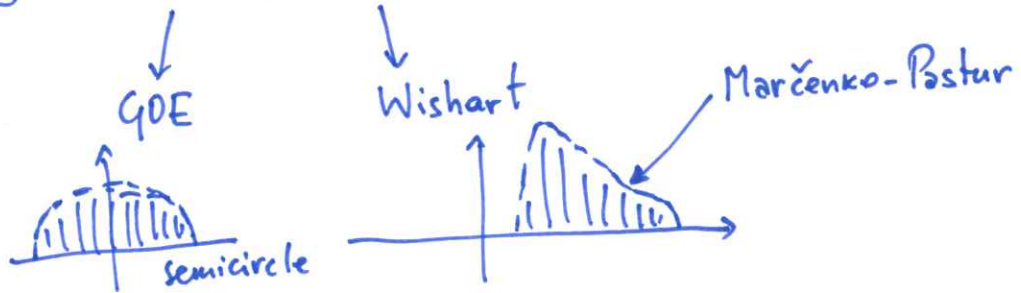
one obtains the following equation for $G_W(z)$

$$\frac{1}{2r} G_W(z) + \left(\frac{1}{2} - \frac{1}{2r}\right) \frac{1}{z} \left[\frac{1}{1-r} + G_W(z)\right] = \frac{1}{2} G_W^2(z),$$

from which the Blue function

$$B_W(z) = \frac{1}{z} + \frac{1}{1-rz}.$$

$$S = a G + b W$$



$$B_G(z) = z + \frac{1}{z}$$

$$B_W(z) = \frac{1}{z} + \frac{1}{1-rz}$$



$$R_G(z) = z$$

$$R_W(z) = \frac{1}{1-rz}$$

Using the rescaling property of the R-transform

$$R_{cH}(z) = c R_H(cz)$$

$$R_S(z) = a R_G(az) + b R_W(bz)$$

$$= a^2 z + \frac{b}{1-rbz}$$

$$B_S(z) = R_S(z) + \frac{1}{z}$$

$$= \frac{1}{z} + a^2 z + \frac{b}{1-rbz}$$



The Green's function for the sum satisfies:

$$z = \frac{1}{G_S(z)} + a^2 G_S(z) + \frac{b}{1-rb G_S(z)}$$

p. 26-27-28 Codes.

[Application to random supergravity 45-46-47].