

# Hadamard product of random matrices and their limiting spectral distributions

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# A standard result of GUE

First, we review some basics of random matrix theory.

- $\{X_{i,j} : 1 \leq i \leq j\}$  is a family of i.i.d. standard normal random variables, i.e.  $X_{i,j} \sim \mathcal{N}(0, 1)$ .
- $W_N$  is called a **Gaussian Unitary Ensemble** (or shortly, **GUE**) if it is a random matrix with the form

$$W_N(i, j) := N^{-\frac{1}{2}} X_{i \wedge j, i \vee j} \text{ for all } 1 \leq i, j \leq N.$$

# A standard result of GUE

- $\text{AED}[W_N] \in \mathbf{P}(\mathbb{R})$  is the **average eigenvalue distribution** of  $W_N$ , which is defined by

$$\text{AED}[W_N] := \frac{1}{N} \mathbb{E} [\delta_{\lambda_1(\omega)} + \cdots + \delta_{\lambda_N(\omega)}],$$

where  $\lambda_i$  are the eigenvalues of  $W_N$ .

- **Theorem** [Wigner, 1955]

$$\text{AED}[W_N] \rightharpoonup w \quad \text{as } N \rightarrow \infty,$$

where  $w$  with density  $\frac{1}{2\pi} \sqrt{4 - x^2}$  is called the **standard semi-circular distribution**.

# Hadamard product and GUE

**Question:** How to generalize this result?

Chakrabarty gives an idea as follows:

Let  $f : [0, 1]^2 \rightarrow \mathbb{R}$  be a symmetric Riemann integrable function (or shortly, **SRIF**), i.e.  $f(x, y) = f(y, x)$ . Then define a random matrix

$$Z_{f,N}(i, j) := f\left(\frac{i}{N+1}, \frac{j}{N+1}\right) W_N(i, j),$$

for  $1 \leq i, j \leq N$ . In other words,  $Z_{f,N} = A_{f,N} \circ W_N$  is a Hadamard product of matrices  $A_{f,N}(i, j) = f(i/(N+1), j/(N+1))$  and  $W_N$ .

# Hadamard product and GUE

- **Theorem** [A. Chakrabarty, 2017] There exists a symmetric probability measure  $\mu_f \in \mathbf{P}_{\text{sym}}(\mathbb{R})$  such that

$$\text{AED}[Z_{f,N}] \rightharpoonup \mu_f \quad \text{as } N \rightarrow \infty.$$

- For  $f(x, y) = r(x)r(y)$ , Chakrabarty shows that  $\mu_f^2 = \mu^{\boxtimes 2} \boxtimes \sigma$ , where  $\mu$  is the law of  $r^2(U)$ ,  $U$  is the standard uniform random variable,  $\boxtimes$  is the **free multiplicative convolution**, and  $\sigma$  is the **Free Poisson distribution** with rate 1.

**Remark:** The density of  $\sigma$  is  $\frac{\sqrt{4x-x^2}}{2\pi x} \mathbb{1}_{(0,4]}(x)$ .

Chakrabarty A. *Stat Probab Lett* 2017; **127** : 150-157.

# Introduction to Free Probability

- $(\mathcal{A}, \phi)$ : a  $*$ -algebra  $\mathcal{A}$  over  $\mathbb{C}$  and a tracial linear functional  $\phi$ . We call such a pair a **Non-Commutative Probability Space** (or shortly, NCPS). An element  $a \in \mathcal{A}$  is called a **random variable**.
- For a random variable  $a \in \mathcal{A}$ , we called  $\mu \in \mathbf{P}(\mathbb{R})$  is **the distribution of  $a$**  if

$$\phi(a^n) = \int_{\mathbb{R}} t^n d\mu(t) \quad \forall n \geq 0.$$

- $\mathcal{B}$  and  $\mathcal{C}$  are two subalgebras of  $\mathcal{A}$ . We say they are **freely independent** if for any  $n \geq 1$ ,  $b_i \in \mathcal{B}$ , and  $c_i \in \mathcal{C}$  satisfy  $\phi(b_i) = \phi(c_i) = 0$  for all  $1 \leq i \leq n$ , we have

$$\phi(b_1 c_1 b_2 c_2 \cdots b_n c_n) = 0.$$

# Introduction to Free Probability

- For  $a, b \in \mathcal{A}$ , we say  $a$  and  $b$  are **free** if  $\text{span}\{1_{\mathcal{A}}, a, a^*\}$  and  $\text{span}\{1_{\mathcal{A}}, b, b^*\}$  are free.
- For  $a, b \in \mathcal{A}$  are free and  $a, b$  have the distributions  $\mu, \nu$ , respectively.
  - If the distribution of  $a + b$  exists, then we denote it by  $\mu \boxplus \nu$ , the **free additive convolution of  $a$  and  $b$** .
  - If the distribution of  $ab$  exists, then we denote it by  $\mu \boxtimes \nu$ , the **free multiplicative convolution of  $a$  and  $b$** .
- If  $\mathcal{A}$  is a  $C^*$ -algebra and  $\phi$  is **positive state**, then for any **self-adjoint element**  $a$ , by **Gelfand transform**, the distribution exists. Moreover, if  $a$  is positive, then the distribution is defined on  $\mathbb{R}^+$ .



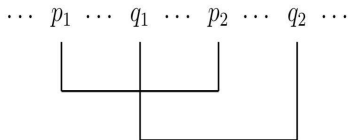
# Introduction to Free Probability

## Theorem 2.1 (Existence of Free Product)

Given  $\{(\mathcal{A}_i, \phi_i)\}_{i \in I}$  a family of NCPS. Then there exists a NCPS  $(\mathcal{A}, \phi)$  and monomorphisms  $\psi_i : \mathcal{A}_i \rightarrow \mathcal{A}$  such that  $\phi_i = \phi \circ \psi_i$  for all  $i \in I$ , which satisfy  $\{\psi_i(\mathcal{A}_i)\}_{i \in I}$  are free in  $(\mathcal{A}, \phi)$ . We denote  $(\mathcal{A}, \phi)$  by  $(\ast_{i \in I} \mathcal{A}_i, \phi)$  and call it the **free product** of  $\{(\mathcal{A}_i, \phi_i)\}_{i \in I}$ .

# Introduction to Free Probability

- A partition  $\pi$  of the set  $[n]$  is called **crossing** if there exist  $p_1 < q_1 < p_2 < q_2$  in  $S$  such that  $p_1 \sim_\pi p_2 \not\sim_\pi q_2 \sim_\pi q_1$ :



If  $\pi$  is not crossing, then it is called **non-crossing**.

- The set of all non-crossing partitions of  $[n]$  is denoted by  $NC(n)$ .

# Main Theorem I

Now, we introduce an approximation lemma.

**Lemma 1 (M. Hsiao, H-W. Huang, 2023)**

*Consider a class*

$$\mathcal{F} := \left\{ \sum_{k=1}^N a_k(x) b_k(y) : \begin{array}{l} \text{for all } N \geq 1 \text{ and } a_k, b_k : [0, 1] \rightarrow \mathbb{R} \\ \text{is continuous } \forall k = 1, \dots, N \end{array} \right\}.$$

*For any SRIF  $f : [0, 1]^2 \rightarrow \mathbb{R}_{\geq 0}$ , there exists a sequence of SRIF's  $\{f_k\}_{k \geq 1} \subset \mathcal{F}$  such that  $f_k$  converges to  $f$  a.s. as  $k \rightarrow \infty$  and  $f_k \geq 0$  for all  $k$ . Moreover, we have*

$$\mu_{f_k} \rightharpoonup \mu_f \text{ as } k \rightarrow \infty.$$

# Main Theorem I

Let  $(\mathcal{A}, \phi_{\mathcal{A}}) = (\mathcal{B}, \phi_{\mathcal{B}}) = (\mathcal{R}([0, 1]), dx)$  be two commutative probability spaces. For  $f(x, y) = \sqrt{\sum r_k(x) \tilde{r}_k(y)}$ , i.e.  $f^2 \in \mathcal{F}$ , we define random variables  $a_k = r_k \in \mathcal{A}$ ,  $b_k = \tilde{r}_k \in \mathcal{B}$ , and

$$\eta_f = \sum_{k=1}^N a_k b_k \in \mathcal{A} * \mathcal{B}.$$

# Main Theorem I

Given a non-crossing partition  $\pi \in NC(n)$ , we denote

$$m_\pi(\eta_f) := \phi(\eta_f^{|V_1|})\phi(\eta_f^{|V_2|}) \cdots \phi(\eta_f^{|V_r|}),$$

where  $\pi = \{V_1, V_2, \dots, V_r\}$ , is a multiplicative function on  $NC(n)$ .

For example, for  $\pi = \{(1, 4, 8), (2, 3), (5), (6, 7)\} \in NC(8)$ , the moment  $m_\pi$  is defined by

$$m_\pi(\eta_f) = \phi(\eta_f^3)\phi(\eta_f^2)^2\phi(\eta_f).$$

# Main Theorem I

## Theorem 3.1 (M. Hsiao, H-W. Huang, 2023)

Let  $f^2 \in \mathcal{F}$  be a nonnegative SRIF and  $\eta_f$  be defined as above. Then

$$m_{2n}(\mu_f) = \sum_{\pi \in NC(n)} m_{\pi}(\eta_f),$$

for all  $n \geq 0$ . In particular, if  $\eta_f$  has the distribution on  $\mathbb{R}$ , then we have  $\mu_f^2 = \eta_f \boxtimes \sigma$ , where  $\sigma$  is the Free Poisson distribution with rate 1.

# Main Theorem II

## Theorem 3.2 (M. Hsiao, H-W. Huang, 2023)

Let  $g \in \mathcal{F}$  be a nonnegative Riemann integrable function. Define a SRIF

$$f(x, y) = \sqrt{2g(2x - 1, 2y)\mathbf{1}_{[\frac{1}{2}, 1] \times [0, \frac{1}{2}]}(x, y) + 2g(2y - 1, 2x)\mathbf{1}_{[0, \frac{1}{2}] \times [\frac{1}{2}, 1]}(x, y)},$$

then for any  $n \geq 1$ , we have

$$m_{2n}(\mu_f) = \sum_{\pi \in NC(n)} m_{\pi}(\eta_{\sqrt{g}}),$$

Moreover, if  $\eta_{\sqrt{g}}$  has a distribution  $\mu$  on  $\mathbb{R}$ , then  $\mu_f = \sqrt{\mu \boxtimes \sigma}$ .

**Remark:** Choose  $g(x, y) = r(x)$ , then we obtain that  $\mu_f = \sqrt{r(U) \boxtimes \sigma}$ .

# Remark

- This generalizes Chakrabarty's results. ( $N = 1$ )
- In general,  $\eta_f$  may not have the distribution. ( $\eta_f$  is not a self-adjoint element, not even a normal element.)
- Let  $\mathcal{H}$  be the set of all  $\mu_f$  with SRIF  $f$ . Then we show that  $\mathcal{H}$  is a convex set. Moreover, we show that

$$\left\{ \sqrt{\mu \boxtimes \sigma} : \mu \in \mathbf{P}(\mathbb{R}_{\geq 0}) \right\} \subsetneq \mathcal{H} \subsetneq \mathbf{P}_{\text{sym}}(\mathbb{R}).$$



# Corollary I

Given two SRIF's  $f_1, f_2$ , the property of normal distribution leads to

$$Z_{f_1, N} + \tilde{Z}_{f_2, N} \stackrel{d}{=} \hat{Z}_{\sqrt{f_1^2 + f_2^2}, N},$$

for all  $N \geq 1$ . However, in general, as  $N \rightarrow \infty$ , Main Theorem I shows that we might not have  $\mu_{f_1} \boxplus \mu_{f_2} = \mu_{\sqrt{f_1^2 + f_2^2}}$ . But, since  $\mathbb{1}$  is freely independent of any random variable., if we choose  $f_1 \equiv c$  to be a constant, we get that

$$\mu_{\sqrt{f^2 + c^2}} = \mu_f \boxplus w_c,$$

for all SRIF  $f$ .

# Corollary I

## Corollary 3.3 (M. Hsiao, H-W. Huang, 2023)

Let  $g_1, g_2 : [0, 1] \rightarrow [0, \infty)$  be two Riemann integrable functions, and define

$$f_1(x, y) = \sqrt{2g_1(2y)\mathbf{1}_{[\frac{1}{2}, 1] \times [0, \frac{1}{2}]}(x, y) + 2g_1(2x)\mathbf{1}_{[0, \frac{1}{2}] \times [\frac{1}{2}, 1]}(x, y)},$$

and

$$f_2(x, y) = \sqrt{2g_2(2x - 1)\mathbf{1}_{[\frac{1}{2}, 1] \times [0, \frac{1}{2}]}(x, y) + 2g_2(2y - 1)\mathbf{1}_{[0, \frac{1}{2}] \times [\frac{1}{2}, 1]}(x, y)}.$$

Then we have  $\mu_{f_1} \boxplus \mu_{f_2} = \mu_{\sqrt{f_1^2 + f_2^2}}$  and

$$\sqrt{\lambda_1 \boxtimes \sigma} \boxplus \sqrt{\lambda_2 \boxtimes \sigma} = \sqrt{(\lambda_1 \boxplus \lambda_2) \boxtimes \sigma},$$

where  $\lambda_i$  is the distribution of the random variable  $g_i(U)$  for  $i = 1, 2$ .

## Corollary II

Therefore, if  $\text{supp}(\mu^{\boxtimes 2}) \subset [\alpha, M]$  holds for some  $\alpha, M > 0$ , consider  $\lambda_1 := \delta_\alpha$  and  $\lambda_2 := \mu^{\boxtimes 2} \boxplus \delta_{-\alpha}$ , we get that

$$\mu \boxtimes w = \sqrt{\mu^{\boxtimes 2} \boxtimes \sigma} = \sqrt{\lambda_1 \boxtimes \sigma} \boxplus \sqrt{\lambda_2 \boxtimes \sigma} = w_\alpha \boxplus \sqrt{\lambda_2 \boxtimes \sigma}.$$

Moreover, since  $\lambda_2$  is a distribution on  $\mathbb{R}_{\geq 0}$ , we obtain

$\nu := \sqrt{\lambda_2 \boxtimes \sigma} \in I_{r,+}$ , that is, for any positive integer  $m \geq 1$ , there exists  $\nu_m \in \mathbf{P}(\mathbb{R}^+)$  such that  $\nu = \nu_m^{\boxplus m}$ .

# Corollary II

## Corollary 3.4 (M. Hsiao, H-W. Huang, 2023)

Let  $\mu \in \mathbf{P}(\mathbb{R}_{\geq 0})$  be compactly supported and  $\alpha > 0$  be a constant.  
TFAE.

- 1  $\text{supp}(\mu^{\boxtimes 2}) \subset [\alpha, \infty)$ .
- 2 There exists a distribution  $\nu \in \mathbf{P}_{\text{sym}}(\mathbb{R})$  with compact support such that  $\nu \boxplus w_\alpha = \mu \boxtimes w$  and  $\nu^2 \in I_{r,+}$ .

**Remark:** Using some Complex Analysis tools, we can remove the bounded support assumption.

# Corollary III

## Corollary 3.5 (M. Hsiao, H-W. Huang, 2023)

Let  $f : [0, 1]^2 \rightarrow \mathbb{R}_{\geq 0}$  be a SRIF. Suppose that there is a constant  $M_f > 0$  such that

$$\int_0^1 f^2(x, y) dy = M_f^2,$$

for all  $x \in [0, 1]$ . Then  $\mu_f = w_{M_f}$ , the semi-circular distribution with mean zero and variance  $M_f$ .

# Proof of Corollary III

*Sketch the proof of Corollary III.*

By approximation lemma, we may assume that  $f^2 \in \mathcal{F}$ . If the function  $f$  is of the form  $f^2(x, y) = M^2 + \sum_{k=1}^N a_k(x)b_k(y)$  with  $\int_0^1 a_k(x)dx = \int_0^1 b_k(y)dy = 0$  for all  $1 \leq k \leq N$ , then by Main Theorem 1., we can show that  $\mu_f = w_M$ . Therefore, for  $f^2 \in \mathcal{F}$  satisfies the assumption, we have the identity

$$\begin{aligned} f^2(x, y) &= M_f^2 + f^2(x, y) - \int_0^1 f^2(x, y)dx - \int_0^1 f^2(x, y)dy + \int_{[0,1]^2} f^2(x, y) \\ &= M_f^2 + \sum_{k=1}^N \left( a_k(x) - \int_0^1 a_k(x)dx \right) \left( b_k(y) - \int_0^1 b_k(y)dy \right). \end{aligned}$$

Hence, we conclude that  $\mu_f = w_{M_f}$ , this completes the proof. □

# Future works

- If  $f$  is only Riemann integrable on  $(0, 1)^2$ ?
- Relation between  $\mu_{\sqrt{f_1^2 + f_2^2}}$  and  $\mu_{f_1}, \mu_{f_2}$ .
- $\boxplus$ -infinitely divisible and Hadamard random matrix.
- Asymptotically freeness?
- Is  $\mathcal{H}$  weakly closed?
- More applications.

# Reference

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