

# **Heavy-tailed random matrices and the Poisson Weighted Infinite Tree**

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## PART I : RANDOM MATRICES

## SPECTRAL MEASURE

Let  $X = (X_{ij})_{1 \leq i, j \leq n}$  be a  $n \times n$  complex matrix. Let  $\lambda_1, \dots, \lambda_n$  be its eigenvalues, the **spectral measure** of  $X$  is

$$\mu_X = \frac{1}{n} \sum_{k=1}^n \delta_{\lambda_k}.$$

- **(random hermitian)** : the array  $(X_{ij})_{i \geq j \geq 1}$  is i.i.d.,  $X_{ij} = \bar{X}_{ji}$ ,
- **(random non-hermitian)** : the array  $(X_{ij})_{i, j \geq 1}$  is i.i.d..

$\implies$  As  $n$  goes to infinity, does the spectral measure converge ?

## WIGNER'S SEMI-CIRCULAR LAW

**Theorem 1.** *If  $\mathbb{E}X_{11} = 0$ ,  $\mathbb{E}|X_{11}|^2 = 1$  and*

$$A_n = X/\sqrt{n},$$

*then, almost surely,*

$$\mu_{A_n} \Longrightarrow \mu_{sc},$$

*where  $\mu_{sc}(dx) = \frac{1}{2\pi} \sqrt{4 - x^2} dx$ .*

## GIRKO'S CIRCULAR LAW

**Theorem 2** (Tao & Vu (2008)). *If  $\mathbb{E}X_{11} = 0$ ,  $\mathbb{E}|X_{11}|^2 = 1$  and*

$$A_n = X/\sqrt{n},$$

*then, almost surely,*

$$\mu_{A_n} \implies \text{Unif}(D).$$

*where  $\text{Unif}(D)$  is the uniform distribution on the unit complex disc.*

$\implies$  Found by Girko, earlier versions due to Edelman, Bai, Pan & Zhou, Götze & Tikhomirov ...

## HEAVY-TAILED ENTRIES

We now assume that

$$\mathbb{P}(|X_{11}| > t) \sim t^{-\alpha}$$

for some

$$0 < \alpha < 2.$$

Define

$$A_n = X/n^{1/\alpha},$$

$\implies$  In the hermitian and non-hermitian cases, does  $\mu_{A_n}$  converge to a measure  $\mu$  ?

## HERMITIAN CASE

**Theorem 3** (Ben Arous & Guionnet, 2008). *There exists a probability measure  $\mu_{bc}$  depending only on  $\alpha$  such that, with the above assumptions, almost surely,*

$$\mu_{A_n} \Longrightarrow \mu_{bc}.$$

$\Longrightarrow$  Found non-rigourously by Bouchaud-Cizeau (1994).

## PROPERTIES OF THE LIMIT MEASURE

**Theorem 4.** For all  $0 < \alpha < 2$ , the probability measure  $\mu_{bc}$

(i) is symmetric and has a bounded density  $f_{bc}$  on  $\mathbb{R}$ ,

$$(ii) f_{bc}(0) = \frac{1}{\pi} \Gamma\left(1 + \frac{2}{\alpha}\right) \left(\frac{\Gamma(1 - \frac{\alpha}{2})}{\Gamma(1 + \frac{\alpha}{2})}\right)^{\frac{1}{\alpha}},$$

$$(iii) f_{bc}(t) \sim_{t \rightarrow \infty} \frac{\alpha}{2} t^{-\alpha-1}.$$

$\implies$  Summarizes properties obtained by Ben Arous & Guionnet, Belinschi, Dembo & Guionnet, Bordenave, Caputo & Chafaï.



## NON-HERMITIAN CASE

**Theorem 5.** Assume that  $X_{11}$  has a *bounded density* on  $\mathbb{R}$  or  $\mathbb{C}$ , and is *asymptotically radial* :

$$\lim_{t \rightarrow \infty} \mathbb{P} \left( \frac{X_{11}}{|X_{11}|} \in \cdot \mid |X_{11}| \geq t \right) = \theta,$$

for some probability distribution  $\theta$  on  $S^1$ .

Then, there exists a probability measure  $\mu$  depending only on  $\alpha$  such that, with the above assumptions, almost surely,

$$\mu_{A_n} \Longrightarrow \mu.$$

## PROPERTIES OF THE LIMIT SPECTRAL MEASURE

**Theorem 6.** *The measure  $\mu$  has **radial bounded density**  $\mu(dz) = f(|z|)dz$ , where*

$$f(0) = \frac{1}{\pi} \frac{\Gamma(1 + \frac{2}{\alpha})^2 \Gamma(1 + \frac{\alpha}{2})^{\frac{2}{\alpha}}}{\Gamma(1 - \frac{\alpha}{2})^{\frac{2}{\alpha}}},$$

*and as  $r \rightarrow \infty$*

$$f(r) \sim c r^{2(\alpha-1)} e^{-\frac{\alpha}{2} r^\alpha}.$$

## PART II : OBJECTIVE METHOD AND LOCAL OPERATOR CONVERGENCE

## HERMITIAN CASE : SPECTRAL MEASURE AT A VECTOR

There exists a probability measure on  $\mathbb{R}$  such that, for all  $t$  integers,

$$(e_k, A_n^t e_k) = \int x^t d\mu_{A_n}^{(k)},$$

$$\mu_{A_n} = \frac{1}{n} \sum_{k=1}^n \mu_{A_n}^{(k)} \quad \text{and} \quad \mu_{A_n}^{(k)} = \sum_{i=1}^n |(e_k, u_i)|^2 \delta_{\lambda_i}.$$

The spectral measure at a vector is well defined for all self-adjoint operators.

## HERMITIAN CASE : REDUCTION TO LOCAL CONVERGENCE

⇒ By **exchangeability**, we get

$$\mathbb{E}\mu_{A_n} = \mathbb{E}\mu_{A_n}^{(1)}$$

⇒ From basic **concentration inequality**,  $\mu_{A_n} - \mathbb{E}\mu_{A_n}$  converges a.s. to 0.

⇒ It is enough to get the convergence of  $\mu_{A_n}^{(1)}$ .

## HERMITIAN CASE : LOCAL OPERATOR CONVERGENCE

We look for a random operator  $A$  defined in  $L^2(V)$  for some countable set  $V$  such that there exists a sequence of bijections  $\sigma_n : V \rightarrow \mathbb{N}$ ,  $\phi \in V$ ,  $\sigma_n(\phi) = 1$  and, for all  $\phi \in L^2(V)$  with compact support, weakly,

$$\sigma_n^{-1} A_n \sigma_n \phi \rightarrow A\phi.$$

If  $A$  is **self-adjoint**, then it would imply that

$$\mu_{A_n}^{(1)} \rightarrow \mu_A^{(\phi)}.$$

## NON-HERMITIAN CASE

The above strategy does not work. The spectral measure at a vector does not exist for non-normal matrices.

The local operator convergence is **not sufficient**.

$$\begin{pmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & \dots \\ & & \dots & & \\ 0 & \dots & 0 & 0 & 0 \end{pmatrix} \quad \text{vs} \quad \begin{pmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & \dots \\ & & \dots & & \\ 1 & \dots & 0 & 0 & 0 \end{pmatrix} .$$

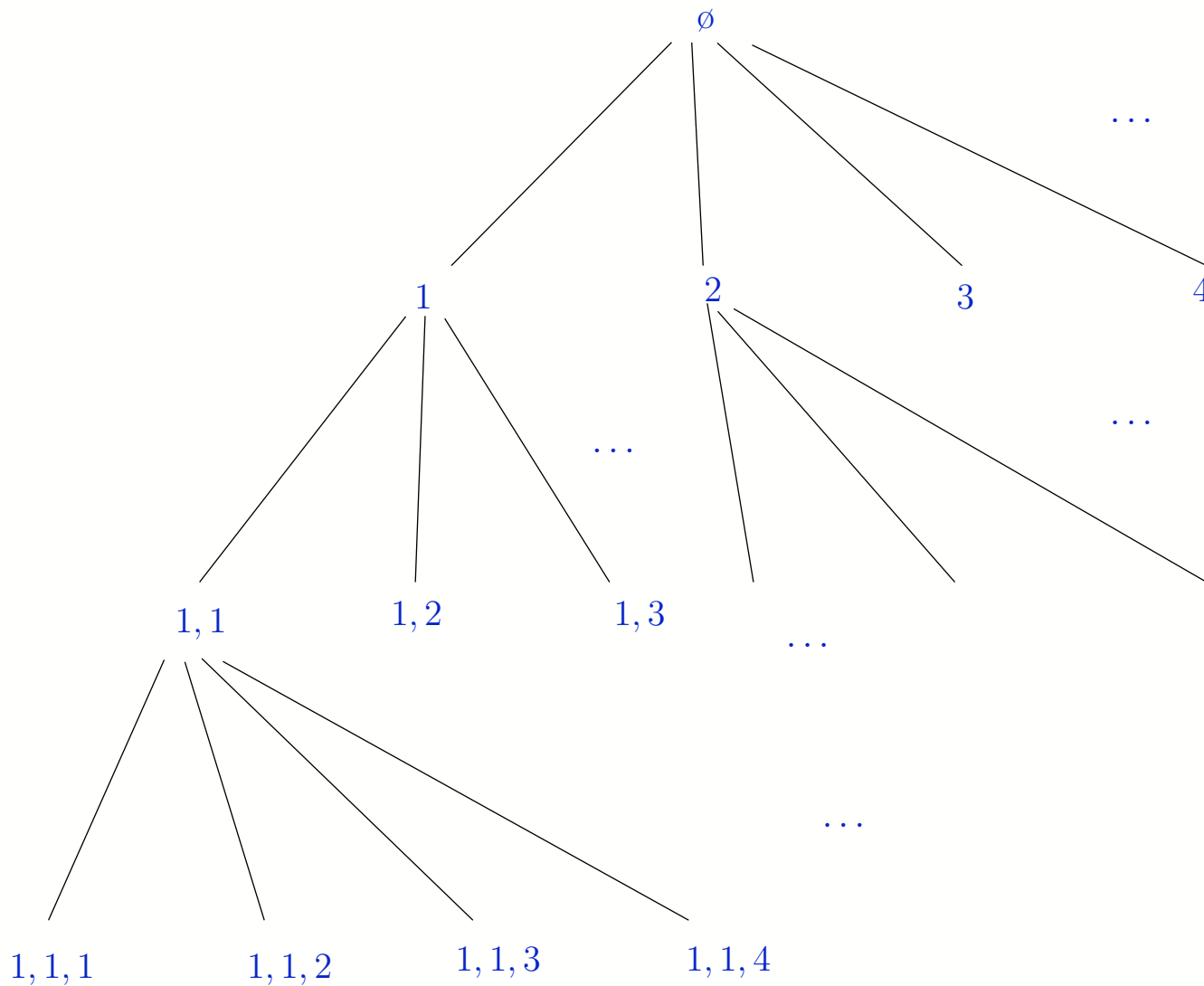
An additional ingredient is needed. (For the moment, we skip this very important part).

## PART III : CONVERGENCE TO POISSON WEIGHTED INFINITE TREE



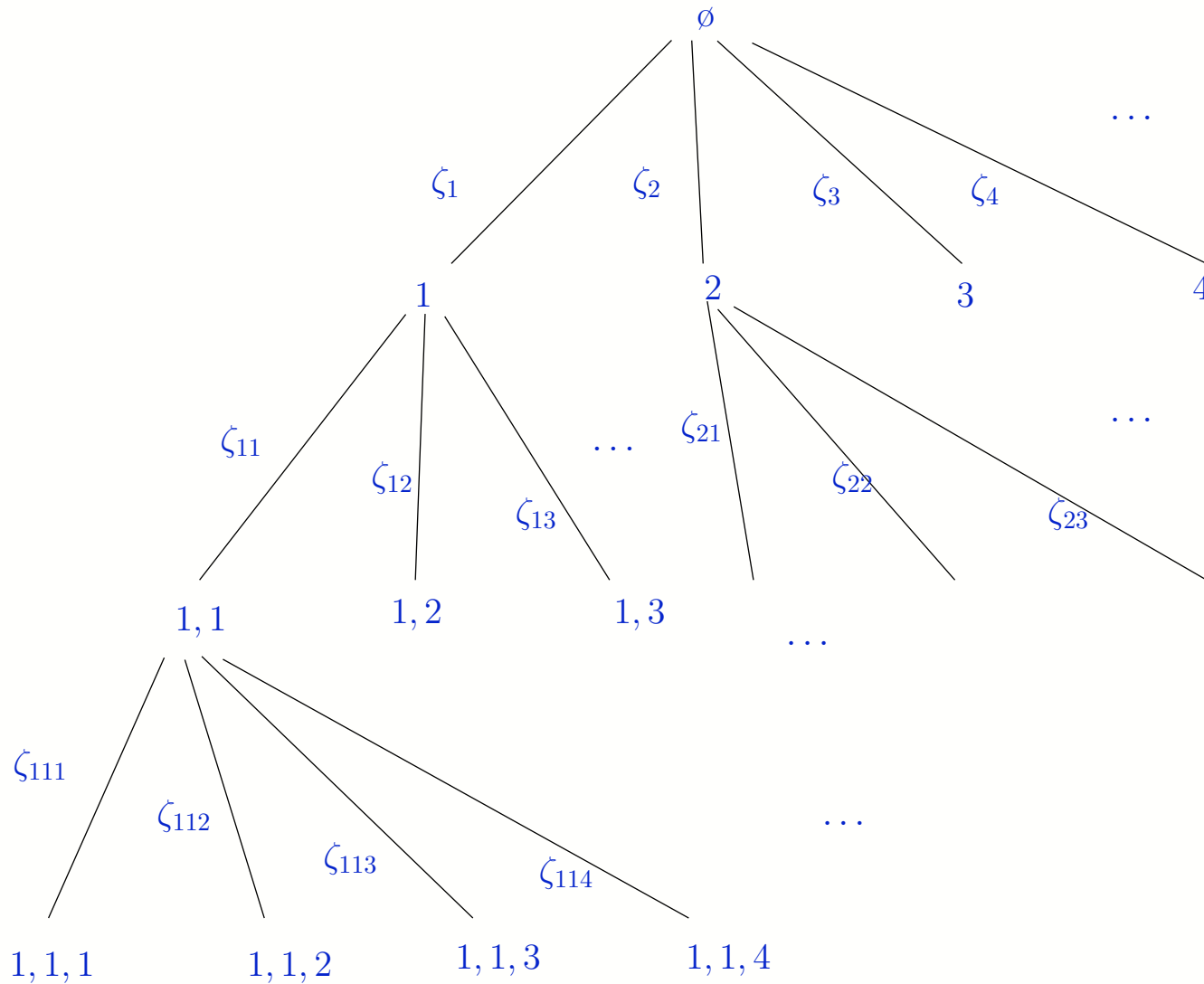
## ALDOUS' PWIT

Let  $V = \bigcup_{k \in \mathbb{N}} \mathbb{N}^k$  with  $\mathbb{N}^0 = \emptyset$ . Consider the infinite tree on  $V$ :



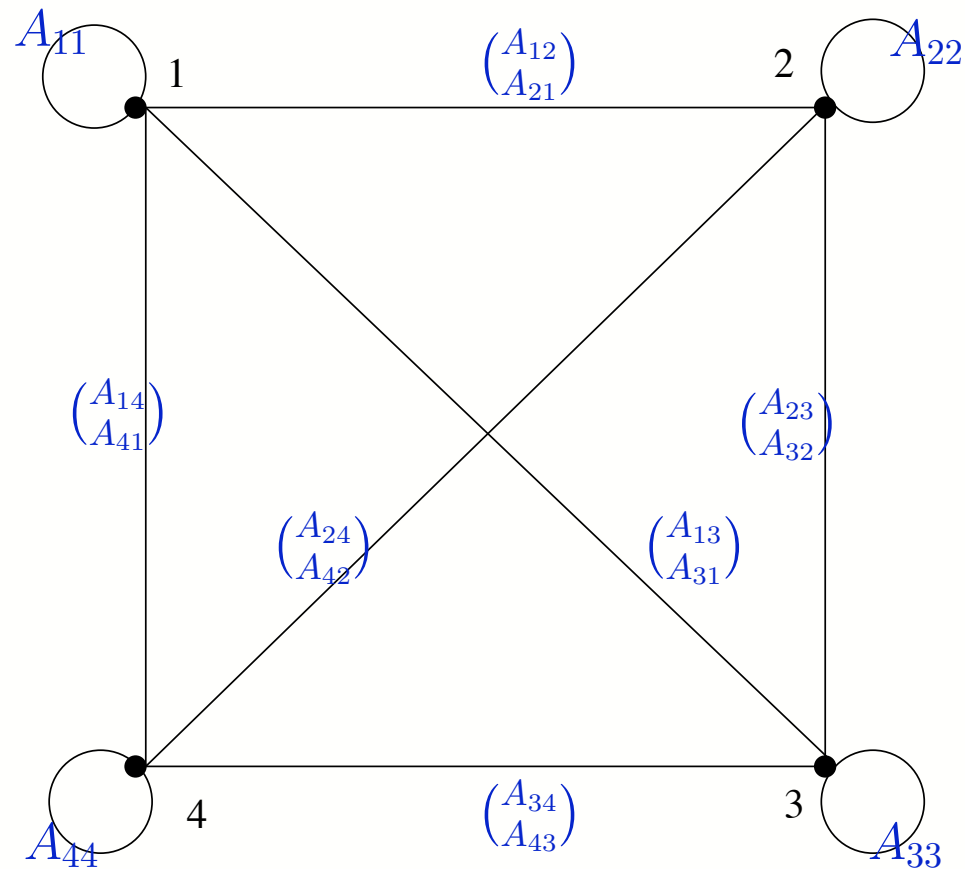
# ALDOUS' PWIT

Let  $(Z_v)_{v \in V}$  be iid Poisson processes of intensity  $\lambda$  on  $\mathbb{R}_+$ ,  
 $Z_v = \{0 \leq \zeta_{v1} \leq \zeta_{v2} \leq \dots\}$



## GRAPHIC REPRESENTATION OF A MATRIX

We think of the matrix  $A_n$  as an **oriented weighted graph** on  $n$  vertices.



## ORDERED STATISTICS

The vector  $\left( \begin{pmatrix} A_{11} \\ A_{11} \end{pmatrix}, \begin{pmatrix} A_{12} \\ A_{21} \end{pmatrix}, \dots, \begin{pmatrix} A_{1n} \\ A_{n1} \end{pmatrix} \right)$  is reordered non-increasingly in

$$\left( \begin{pmatrix} A_{1\sigma(1)} \\ A_{\sigma(1)1} \end{pmatrix}, \begin{pmatrix} A_{1\sigma_1(2)} \\ A_{\sigma(2)1} \end{pmatrix}, \dots, \begin{pmatrix} A_{1\sigma(n)} \\ A_{\sigma(n)1} \end{pmatrix} \right)$$

with

$$\left\| \begin{pmatrix} A_{1\sigma(1)} \\ A_{\sigma(1)1} \end{pmatrix} \right\|_1 \geq \left\| \begin{pmatrix} A_{1\sigma(2)} \\ A_{\sigma(2)1} \end{pmatrix} \right\|_1 \geq \dots .$$

$\implies$  We restrict ourselves to **non-hermetian case** and **non-negative random variables**.

## CONVERGENCE OF ORDERED STATISTICS

$$\left( \left( \begin{array}{c} A_{1\sigma(1)} \\ A_{\sigma(1)1} \end{array} \right), \left( \begin{array}{c} A_{1\sigma(2)} \\ A_{\sigma(2)1} \end{array} \right), \dots, \left( \begin{array}{c} A_{1\sigma(n)} \\ A_{\sigma(n)1} \end{array} \right) \right)$$

converges to

$$\left( \left( \begin{array}{c} \varepsilon_1 \\ 1 - \varepsilon_1 \end{array} \right) \zeta_1^{-\frac{1}{\alpha}}, \left( \begin{array}{c} \varepsilon_2 \\ 1 - \varepsilon_2 \end{array} \right) \zeta_2^{-\frac{1}{\alpha}}, \dots \right),$$

where  $(\zeta_k)_{k \geq 1}$ ,  $\zeta_1 \leq \zeta_2 \leq \dots$ , is a Poisson point process of intensity

$$\Lambda(dx) = 2\mathbb{I}_{x>0}dx$$

and  $(\varepsilon_k)$  iid  $\text{Ber}(1/2)$  random variables.

## CONVERGENCE OF ORDERED STATISTICS

For fixed  $i$ , the vector

$$\left( \left( \begin{array}{c} A_{j\sigma(i)} \\ A_{\sigma(i)j} \end{array} \right) \right)_{j \neq 1}$$

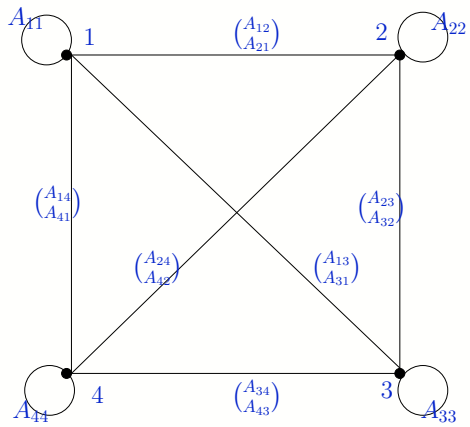
is reordered non-increasingly.

It converges again to

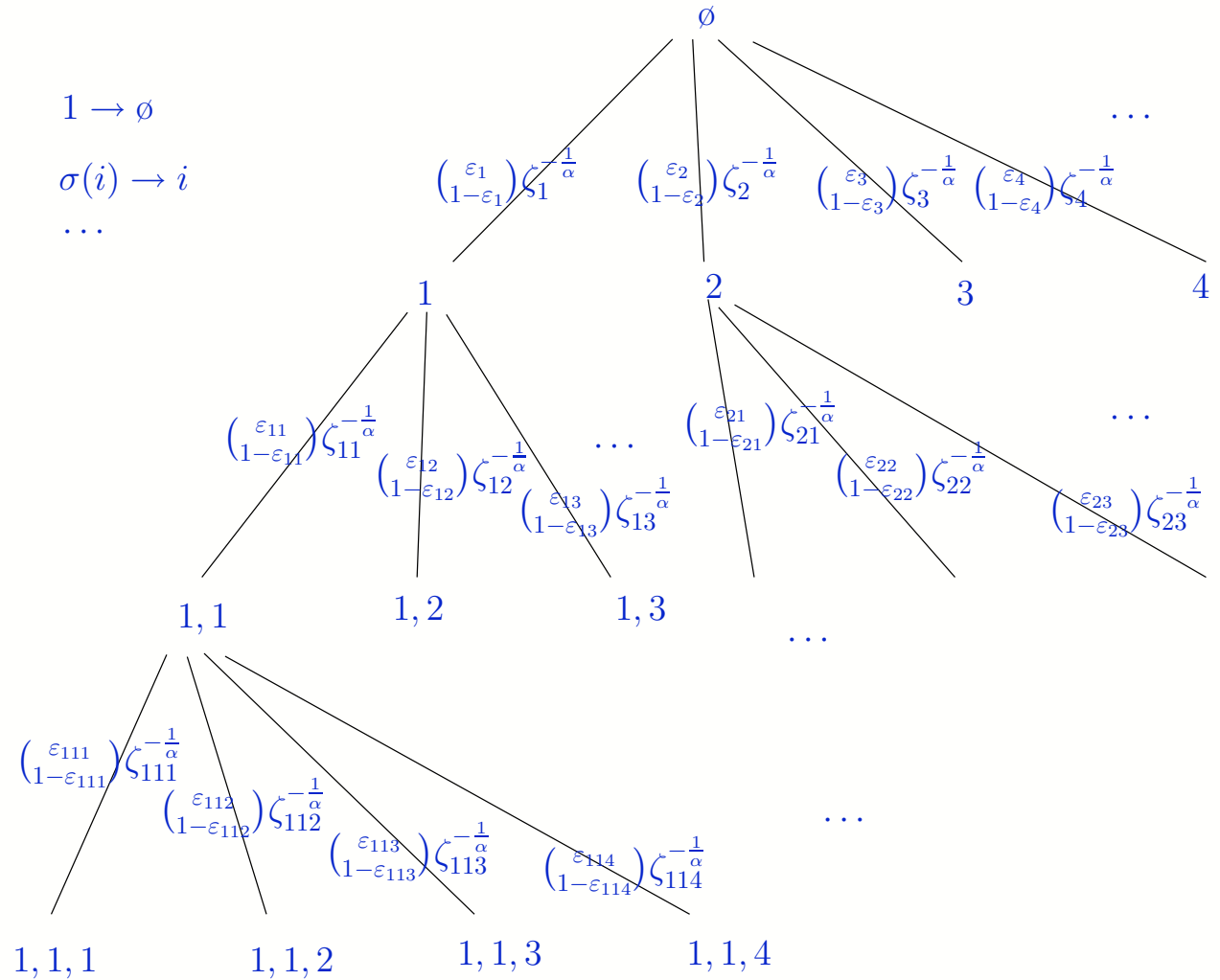
$$\left( \left( \begin{array}{c} \varepsilon_{i1} \\ 1 - \varepsilon_{i1} \end{array} \right) \zeta_{i1}^{-\frac{1}{\alpha}}, \left( \begin{array}{c} \varepsilon_{i2} \\ 1 - \varepsilon_{i2} \end{array} \right) \zeta_{i2}^{-\frac{1}{\alpha}}, \dots \right),$$

where  $(\zeta_{ik})_{k \geq 1}$  are **independent** Poisson processes of intensity  $\Lambda$  and  $(\varepsilon_{ik})_{i,k}$  iid  $\text{Ber}(1/2)$  r.v.

# LOCAL CONVERGENCE TO ALDOUS' PWIT



$\longrightarrow$   
 $n \rightarrow \infty$



## OPERATOR ON THE PWIT

Define the operator on compactly supported function of  $L^2(V)$ ,

$$A\delta_v = \sum_{k \geq 1} (1 - \varepsilon_{vk}) \zeta_{vk}^{-\frac{1}{\alpha}} \delta_{vk} + \varepsilon_v \zeta_v^{-\frac{1}{\alpha}} \delta_{a(v)},$$

where  $a(v)$  is the ancestor of  $v \neq \emptyset$ .

$\implies$  There exists a sequence of bijections  $\sigma_n : V \rightarrow \mathbb{N}$ ,  $\emptyset \in V$ ,  $\sigma_n(\emptyset) = 1$  and, for all  $\phi \in L^2(V)$  with compact support, weakly,

$$\sigma_n^{-1} A_n \sigma_n \phi \rightarrow A\phi.$$



## HERMITIAN CASE : OPERATOR ON THE PWIT

In the hermitian case, the operator is defined similarly, we simply forget about, the  $\varepsilon'_v$ 's :

$$A\delta_v = \sum_{k \geq 1} \zeta_{vk}^{-\frac{1}{\alpha}} \delta_{vk} + \zeta_v^{-\frac{1}{\alpha}} \delta_{a(v)}.$$

—→ Again, for all  $\phi \in L^2(V)$  with compact support, weakly,  $\sigma_n^{-1} A_n \sigma_n \phi \rightarrow A\phi$ .

**Theorem 7.** *With probability one, the operator  $A$  is (essentially) self-adjoint.*

⇒ As a corollary, we obtain the convergence of  $\mu_{A_n}$  to  $\mu_{bc} := \mathbb{E} \mu_A^{(\emptyset)}$ .

## HERMITIAN CASE : RECURSIVE DISTRIBUTIONAL EQUATION

The **resolvent formula** and the **recursive structure of the PWIT** implies a RDE for,  
 $z \in \mathbb{C}_+ = \{z \in \mathbb{C} : \Im(z) > 0\}$ ,

$$g_\emptyset(z) := \langle \delta_\emptyset, (A - z)^{-1} \delta_\emptyset \rangle$$

$$g_\emptyset \stackrel{d}{=} - \left( z + \sum_{k \in \mathbb{N}} \xi_k g_k \right)^{-1},$$

where  $g_\emptyset, (g_k)_{k \in \mathbb{N}}$  are i.i.d. independent of  $\{\xi_k\}_{k \in \mathbb{N}}$ , a independent Poisson point process of  $\mathbb{R}_+$  with intensity  $\frac{\alpha}{2} x^{-\frac{\alpha}{2}-1} dx$ .

## RECURSIVE DISTRIBUTIONAL EQUATION

If  $S$  is a positive  $\alpha/2$ -stable random variable,

$$\sum_{k \in \mathbb{N}} \xi_k g_k \stackrel{d}{=} \mathbb{E}[g_\emptyset^{\frac{\alpha}{2}}]^{\frac{2}{\alpha}} S.$$

$\implies$  The RDE can be solved in terms of a scalar fixed point equation for  $\mathbb{E}[g_\emptyset^{\frac{\alpha}{2}}]^{\frac{2}{\alpha}}$ .

$\implies$  Since  $g$  is the **Cauchy-Stieltjes transform** of  $\mu_\emptyset$ , we deduce the properties of  $\mu_{bc} = \mathbb{E}\mu_A^{(\emptyset)}$ .

## PART IV : CONVERGENCE IN THE NON-HERMITIAN CASE

## SINGULAR VALUES

$0 \leq \sigma_n \leq \dots \leq \sigma_1$  : square roots of the eigenvalues of  $A^*A$ .

Since  $|\det A| = \sqrt{\det(A^*A)}$ ,

$$\prod_{k=1}^n |\lambda_k| = \prod_{k=1}^n \sigma_k$$

For  $z \in \mathbb{C}$ , let  $\sigma_n(z) \leq \dots \leq \sigma_1(z)$  be the singular values of  $A - z$  and

$$\nu_A(z) = \frac{1}{n} \sum_{k=1}^n \delta_{\sigma_k(z)}.$$

$\implies$  for all  $z \in \mathbb{C} \setminus \text{supp}(\mu_A)$ ,

$$U_{\mu_A}(z) = \int_{\mathbb{C}} \ln |\lambda - z| \mu_A(d\lambda) = \int_{\mathbb{R}_+} \ln(x) \nu_A(z, dx)$$

## LOGARITHMIC POTENTIAL

$$U_\mu(z) = \int_{\mathbb{C}} \ln |\lambda - z| \mu(d\lambda).$$

Define

$$\partial = \frac{1}{2} \left( \frac{\partial}{\partial x} - i \frac{\partial}{\partial y} \right) \quad \text{and} \quad \bar{\partial} = \frac{1}{2} \left( \frac{\partial}{\partial x} + i \frac{\partial}{\partial y} \right).$$

Laplacian differential operator

$$\Delta = \bar{\partial}\partial = \frac{1}{4} \left( \frac{\partial^2}{\partial^2 x} + \frac{\partial^2}{\partial^2 y} \right).$$

In  $\mathcal{D}'(\mathbb{C})$ ,

$$\Delta U_\mu = \pi \mu.$$

$\implies$  The logarithmic potential characterizes the measure.

## GIRKO'S METHOD

**Theorem 8** (Criterion of convergence). *Let  $(A_n)$  be a sequence of matrices. Assume that for almost all  $z \in \mathbb{C}$ ,*

(i)  $\nu_{A_n}(z)$  converges weakly to  $\nu(z)$ , a probability measure on  $\mathbb{R}_+$ .

(ii)  $\int \ln(x) \nu_{A_n}(z, dx)$  is uniformly integrable.

Then there exists a probability measure  $\mu$  on  $\mathbb{C}$ , such that for almost all  $z \in \mathbb{C}$ ,

$U_\mu(z) = \int \ln(x) \nu(z, dx)$  and  $\mu_{A_n}$  converges weakly to  $\mu$ .

## CONVERGENCE OF THE SINGULAR VALUES

**Theorem 9.** *For all  $z \in \mathbb{C}$ , there exists a measure  $\nu(z, \cdot)$ , depending only on  $\alpha$  and  $z$ , such that almost surely*

$$\lim_n \nu_{A_n}(z) = \nu(z).$$

(For  $z = 0$ , Belinschi, Dembo & Guionnet (2009). For an explanation of this result, wait a few slides)



## UNIFORM INTEGRABILITY

Bai (1999) has developed the first method to prove the uniform integrability of

$$\int \ln(x) \nu_{A_n}(z, dx) = \frac{1}{n} \sum_{i=1}^n \ln \sigma_i(z).$$

Here, we adapt the argument of Tao & Vu (2008) for the circular law.

For the **large singular values**, we use the inequality, for any  $0 < p \leq 2$ ,

$$\sum_{i=1}^n \sigma_i^p \leq \sum_{i=1}^n \|R_i\|_2^p = \sum_{i=1}^n \left( \sum_{j=1}^n |A_{ij}|^2 \right)^{\frac{p}{2}},$$

where  $R_1, \dots, R_n$  are rows of  $A$ .

## UNIFORM INTEGRABILITY

For the **smallest singular value**, the bounded density assumption implies easily, for some  $p \geq 1$ , almost surely,

$$\sigma_n(z) = \Omega(n^{-p}).$$

$\implies$  we may lower bound  $\sigma_{n-i}(z)$  by  $n^{-p}$  for  $0 \leq i \leq n^{1-\gamma}$ .

For the **moderately small singular values**, Tao & Vu prove that almost surely, for all  $n^{1-\gamma} \leq i \leq n$ ,

$$\sigma_{n-i}(z) = \Omega\left(\frac{i}{n}\right).$$

We only prove that, in a weaker sense, that

$$\sigma_{n-i}(z) = \Omega\left(\frac{i}{n}\right)^{\frac{1}{\alpha} + \frac{1}{2}}.$$

## A NEW LOOK AT THE SPECTRAL MEASURE

We want to study the limit spectral measure  $\mu$ .

The measure  $\nu(z)$  is not explicit and we only know that

$$U_\mu(z) = \int \ln(x) \nu(z, dx)$$

and

$$\mu = \frac{1}{\pi} \Delta U_\mu.$$

$\implies$  We need another characterization of the spectral measure : some have appeared in the physics literature, Feinberg & Zee, Jarosz & Nowak, Rogers & Castillo...

## BIPARTIZATION

Define

$$B_{ij} = \begin{pmatrix} 0 & A_{ij} \\ \bar{A}_{ji} & 0 \end{pmatrix} \in \mathcal{M}_2(\mathbb{C}).$$

$\implies B = (B_{ij})_{1 \leq i, j \leq n}$  is an **hermitian matrix** in  $\mathcal{M}_n(\mathcal{M}_2(\mathbb{C})) \simeq \mathcal{M}_{2n}(\mathbb{C})$ .

# BIPARTIZATION

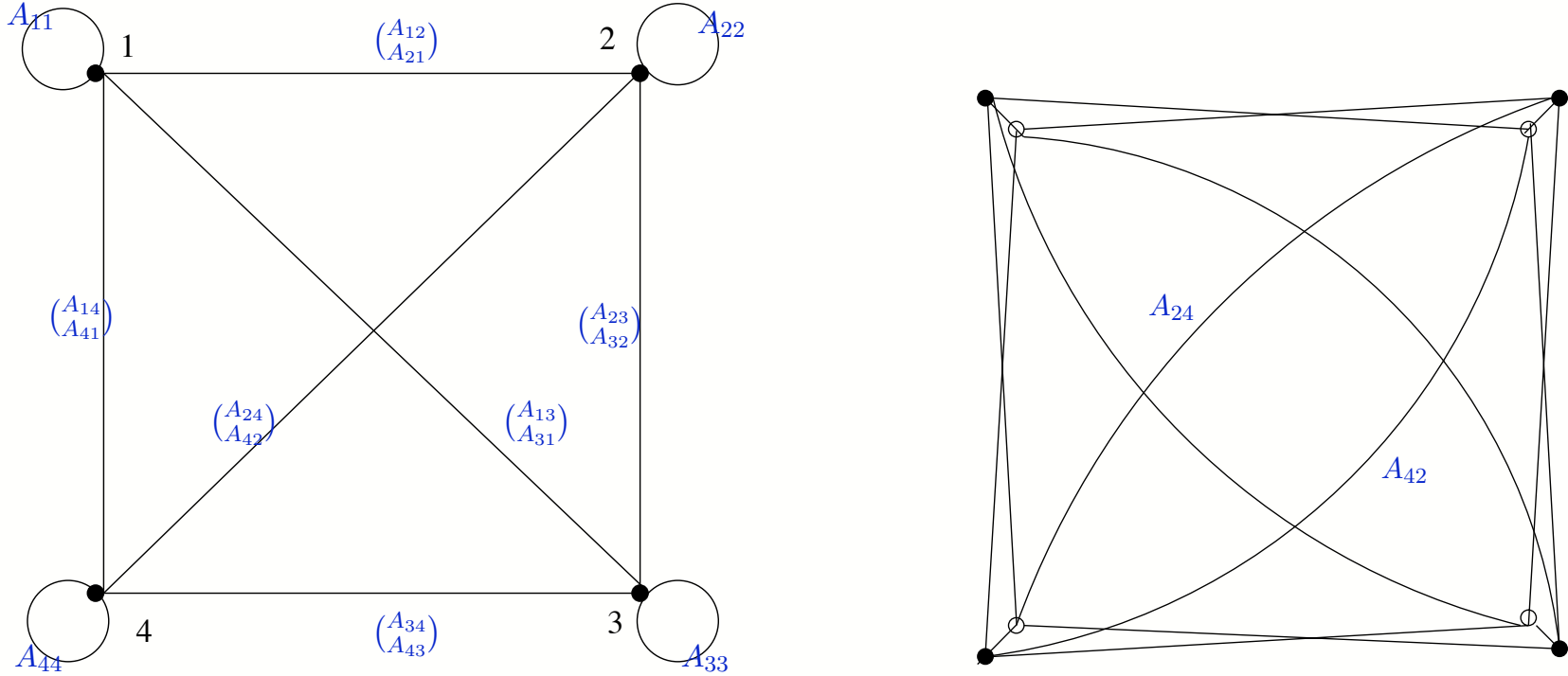


Figure 1: Graphical interpretation of bipartization.

## RESOLVENT

Define the quaternionic set  $\mathbb{H}_+ = \left\{ U = \begin{pmatrix} \eta & z \\ \bar{z} & \eta \end{pmatrix}, \eta \in \mathbb{C}_+, z \in \mathbb{C} \right\}$ .

Resolvent matrix :

$$R = (B - U \otimes I_n)^{-1} \in \mathcal{M}_n(\mathcal{M}_2(\mathbb{C})),$$

$$B - U \otimes I_n = \begin{pmatrix} B_{11} - U & B_{12} & \cdots \\ B_{12}^* & B_{22} - U & \cdots \\ \cdots & \cdots & \cdots \end{pmatrix}.$$

## FROM THE RESOLVENT TO THE SINGULAR VALUES

$$R_{kk}(U) = \begin{pmatrix} a_k & b_k \\ b'_k & c_k \end{pmatrix}, \quad \text{with } U = \begin{pmatrix} \eta & z \\ \bar{z} & \eta \end{pmatrix}.$$

For  $\nu_A(z)$  : the trace of  $R$  is the Cauchy-Stieltjes transform of  $\check{\nu}_A(z)$ ,

$$\frac{1}{n} \sum_{k=1}^n \frac{1}{2} (a_k + c_k) = \int \frac{1}{x - \eta} \check{\nu}_A(z, dx),$$

$$\check{\nu}_A(z) = \frac{1}{2n} \sum_{i=1}^n \delta_{\sigma_i(z)} + \delta_{-\sigma_i(z)}$$

## FROM THE RESOLVENT TO THE SPECTRAL MEASURE

For  $\mu_A$  : in  $\mathcal{D}'(\mathbb{C})$ ,

$$\mu_A = -\frac{1}{\pi n} \sum_{k=1}^n \partial b_k(\cdot, 0) = \lim_{\eta \downarrow 0} -\frac{1}{\pi n} \sum_{k=1}^n \partial b_k(\cdot, \eta).$$

(Similar computation in Rogers & Costillo (2009))



## LOCAL OPERATOR CONVERGENCE

⇒ By exchangeability, we get

$$\mathbb{E}\mu_{A_n} = \lim_{\eta \downarrow 0} -\frac{1}{\pi} \partial \mathbb{E}b_1(\cdot, \eta).$$

⇒ It is enough to get the convergence of  $R_{11}(U)$ .

The local convergence of  $A_n$  to an operator  $A$

$$\sigma_n^{-1} A_n \sigma_n \phi \rightarrow A\phi$$

implies the local convergence of  $B_n$  to  $B$ , the bipartized operator of  $A$ .

## LOCAL OPERATOR CONVERGENCE

We show that  $B$  is (essentially) self-adjoint  $\implies$  convergence of  $\nu_{A_n}(z)$ .

+ **Uniform integrability**, we have  $\mu = \lim_{t \downarrow 0} -\frac{1}{\pi} \partial \mathbb{E} b(\cdot, it)$ , where

$$R_{\emptyset\emptyset}(U) = \begin{pmatrix} a & b \\ b' & c \end{pmatrix},$$

and

$$R = (B - U \otimes I)^{-1}.$$

## RECURSIVE DISTRIBUTIONAL EQUATION

The **resolvent formula** and the **recursive structure of the PWIT** implies a RDE for

$$R_{\emptyset\emptyset}(U) = \begin{pmatrix} a & b \\ \bar{b} & c \end{pmatrix}.$$

$$\begin{pmatrix} a & b \\ \bar{b} & c \end{pmatrix} \stackrel{d}{=} - \left( U + \sum_{k \in \mathbb{N}} \begin{pmatrix} \xi'_k c_k & 0 \\ 0 & \xi_k a_k \end{pmatrix} \right)^{-1},$$

where  $a, c, (a_k)_{k \in \mathbb{N}}, (c_k)_{k \in \mathbb{N}}$  are i.i.d. independent of  $\{\xi_k\}_{k \in \mathbb{N}}, \{\xi'_k\}_{k \in \mathbb{N}}$  two independent Poisson point processes of  $\mathbb{R}_+$  with intensity  $\frac{\alpha}{2} x^{-\frac{\alpha}{2}-1} dx$ .

## RECURSIVE DISTRIBUTIONAL EQUATION

For  $\eta = it$ ,  $a = ih$  is pure imaginary and

$$h \stackrel{d}{=} \frac{t + \sum_{k \in \mathbb{N}} \xi_k h_k}{|z|^2 + \left(t + \sum_{k \in \mathbb{N}} \xi_k h_k\right) \left(t + \sum_{k \in \mathbb{N}} \xi'_k h'_k\right)}.$$

If  $S$  is a positive  $\alpha/2$ -stable random variable,

$$\sum_{k \in \mathbb{N}} \xi_k h_k \stackrel{d}{=} \mathbb{E}[h_1^{\frac{\alpha}{2}}]^{\frac{2}{\alpha}} S.$$

$\implies$  The RDE can be solved in terms of a scalar fixed point equation for  $\mathbb{E}[h_1^{\frac{\alpha}{2}}]^{\frac{2}{\alpha}}$ .

$\implies$  From  $\mu = \lim_{t \downarrow 0} -\frac{1}{\pi} \partial \mathbb{E} b_1(\cdot, it)$ , we get the properties of  $\mu$ .

## IN SUMMARY

- The **objective method** is an efficient framework to deal with sparse random matrices.
- Dependencies in the entries are allowed : all computations are done in the limit operator.
- In other sparse cases : how to prove the uniform integrability ?
- What about eigenvectors ? analogs of local Wigner's theorem ?

## OPEN PROBLEM

Let  $G_n$  be a  $k$ -oriented regular graph on  $\{1, \dots, n\}$ , drawn uniformly.

Consider its adjacency matrix

$$(A_n)_{ij} = \mathbb{I}(i \rightarrow j).$$

The limit operator is the adjacency operator of the  $k$ -oriented regular infinite tree. The computation on the  $2 \times 2$  resolvent shows that

$$\mu(dz) = \frac{1}{\pi} \frac{k^2(k-1)}{(k^2 - |z|^2)^2} \mathbb{I}_{|z| < \sqrt{k}} dz.$$

(= Brown's measure of the free sum of  $k$  Haar unitary, Haagerup and Larsen (2000))

$\implies$  How to prove the uniform integrability of the spectral measure in this case ?