

# A COUPLING OF ROOTED SPECTRAL MEASURES

T. Espinasse and P. Rochet



1 Spectral measure

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3 Théorème limite

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# Spectral measure

$A = (a_{ij})_{i,j=1,\dots,N}$  real symmetric matrix.

## Definition

The local spectral measure at  $i$  is the unique measure  $\mu_i$  on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  with moments:

$$\int_{\mathbb{R}} x^k d\mu_i(x) = (A^k)_{ii} , \quad k = 0, 1, 2, \dots$$

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For  $A$  the adjacency matrix of a graph  $G$ ,  $\mu_i$ 's  $k$ -th moment counts the length  $k$  closed walks from  $i$  to  $i$  in  $G$ .

In this wase, we call  $\mu_i$  the *rooted spectral measure* at  $i$ .

# Spectral measure

## Proposition

Let  $A = P\Lambda P^\top$  be a spectral decomposition of  $A$  with  $P = (p_{ij})_{i,j=1,\dots,N}$  orthogonal and  $\Lambda = \text{Diag}(\lambda_1, \dots, \lambda_N)$ , alors

$$\mu_i = \sum_{j=1}^N p_{ij}^2 \delta_{\lambda_j}.$$

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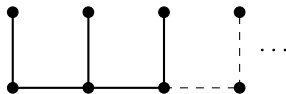
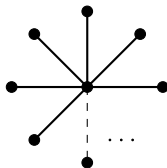
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- $(A^k)_{ii} = (P\Lambda^k P^\top)_{ii} = \sum_{j=1}^N p_{ij}^2 \lambda_j^k$
- Does not depend on the choice of the spectral decomposition
- Spectral measure  $\sum_{i=1}^N \delta_{\lambda_i} = \sum_{i=1}^N \mu_i$  ( $k$ -th moment =  $\text{tr}(A^k)$ )

# In the literature

- Asymptotic distribution of random matrices spectra
- Benjamini-Schramm convergence of rooted random graphs
- Limit theorems on star graphs and comb graphs



- Reconstruction conjecture: Is a (non-regular) graph characterized by its rooted spectral measures?

# Coupling

Goal: Define an "interesting" joint measure  $\mu$  on  $\mathbb{R}^N$  with margins  $\mu_1, \dots, \mu_N$



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Let  $S_N$  be the set of permutations on  $\{1, \dots, N\}$  and  $\pi$  the signed measure on  $S_N$  defined by

$$\pi(\{\sigma\}) = \epsilon(\sigma) \prod_{j=1}^N p_{j\sigma(j)} \text{ , } \sigma \in S_N$$

where  $\epsilon(\cdot)$  is the signature.

- $\pi(\sigma) = \det(P \odot M_\sigma)$  with  $M_\sigma$  the permutation matrix associated to  $\sigma$  and  $\odot$  the Hadamard product.

- $\pi(S_N) = \sum_{\sigma \in S_N} \pi(\{\sigma\}) = \det(P) = 1$  (quasi-probability)

For  $\lambda = (\lambda_1, \dots, \lambda_N)$  the spectrum of  $A$ , let

$$\lambda_\sigma = (\lambda_{\sigma(1)}, \dots, \lambda_{\sigma(N)}) , \quad \sigma \in S_N.$$

### Definition

The *joint spectral measure*  $\mu$  is the pushforward measure of  $\pi$  by  $\sigma \mapsto \lambda_\sigma$ :

$$\mu(\{\lambda_\sigma\}) = \sum_{\tau: \lambda_\tau = \lambda_\sigma} \epsilon(\tau) \prod_{j=1}^N p_{j\tau(j)}$$

- $\text{Supp}(\mu) = \{\lambda_\sigma , \sigma \in S_N\}$
- Distinct eigenvalues:  $\mu(\{\lambda_\sigma\}) = \pi(\{\sigma\}) = \epsilon(\sigma) \prod_{j=1}^N p_{j\sigma(j)}$
- Does not depend on the spectral decomposition
- $\mu$  is a *quasi-probability*:  $\mu(\{\lambda_\sigma , \sigma \in S_N\}) = 1$

# Multivariate moments

let

- $m[k_1, \dots, k_N]$  is the multivariate moment

$$m[k_1, \dots, k_N] = \int_{\mathbb{R}^N} x_1^{k_1} \dots x_N^{k_N} d\mu(x_1, \dots, x_N).$$

- $A[k_1, \dots, k_N]$  is the  $N \times N$  whose  $i$ -th column is the  $i$ -th column of  $A^{k_i}$ ,  $i = 1, \dots, N$ :

$$A[k_1, \dots, k_N] = \left[ \begin{bmatrix} (A^{k_1})_{11} \\ \vdots \\ (A^{k_1})_{N1} \end{bmatrix} \dots \begin{bmatrix} (A^{k_N})_{1N} \\ \vdots \\ (A^{k_N})_{NN} \end{bmatrix} \right]$$

## Lemma

For all  $k_1, \dots, k_N \in \mathbb{N}$ ,

$$m[k_1, \dots, k_N] = \det (A[k_1, \dots, k_N]).$$

Proof:

$$\blacksquare \quad m[k_1, \dots, k_N] = \sum_{\sigma \in S_N} \lambda_{\sigma(1)}^{k_1} \dots \lambda_{\sigma(n)}^{k_N} \mu(\{\lambda_\sigma\})$$

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$$\blacksquare \quad A^k P = P \Lambda^k \implies (A^k P)_{ij} = p_{ij} \lambda_j^k \text{ and}$$

$$m[k_1, \dots, k_N] = \sum_{\sigma \in S_N} \epsilon(\sigma) \prod_{i=1}^N (A^{k_i} P)_{i\sigma(i)} = \det(A[k_1, \dots, k_N] P) \quad \square$$

## Corollary

The margins of  $\mu$  are the rooted spectral measures  $\mu_i$ .

Proof:

$\mu_i$  is characterized by its moments, with for all  $k \in \mathbb{N}$ :

$$\int_{\mathbb{R}} x^k d\mu_i(x) = (A^k)_{ii} = \det \left( A[0, \dots, 0, \underset{\substack{\uparrow \\ i\text{-th position}}}{k}, 0, \dots, 0] \right) \quad \square$$

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Let  $X = (X_1, \dots, X_N)$  be a vector with (quasi)-distribution  $\mu$ , we write

$$\mathbb{E}(g(X)) = \int_{\mathbb{R}^N} g(x) d\mu(x) \quad \text{e.g.} \quad m[k_1, \dots, k_N] = \mathbb{E}(X_1^{k_1} \dots X_N^{k_N})$$



## Proposition

If  $A$  is the adjacency matrix of a simple graph,

$$\text{var}(X) = L$$

where  $L$  is the graph Laplacian.

Preuve:

- $\text{var}(X_i) = (A^2)_{ii} = \deg(i)$
- $\text{cov}(X_i, X_j) = \mathbb{E}(X_i X_j) - \mathbb{E}(X_i)\mathbb{E}(X_j)$

$$= \det \left( \begin{bmatrix} 0 & a_{ij} \\ a_{ij} & 0 \end{bmatrix} \right) = -a_{ij}^2 = -a_{ij} \quad (i \neq j) \quad \square$$

More generally:  $\text{cov}(X_i^k, X_j^k) = -(A^k)_{ij}^2$

# Probabilities on subsets

For  $u, v \subset \{1, \dots, N\}$ , let  $M_{uv}$  be the submatrix of  $M$  restricted to  $(i, j), i \in u, j \in v$ .

## Proposition

If the eigenvalues  $\lambda_j$  are distinct, for all subsets  $u, v \subset \{1, \dots, N\}$  of equal sizes,

$$\mathbb{P}(\{X_i, i \in u\} = \{\lambda_j, j \in v\}) = \det(P_{uv})^2.$$

- If  $u, v$  are singletons,

$$\mathbb{P}(X_i = \lambda_j) = p_{ij}^2$$

- The proof uses the Jacobi identity

$$\det(P_{\bar{u}\bar{u}}) = \det(P_{\bar{u}\bar{u}}^{-1}) = \frac{\det(P_{uu})}{\det(P)}$$

where  $\bar{u} = \{1, \dots, N\} \setminus u$ .

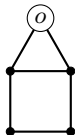
# Weak convergence

Let  $\{G_n, n \in \mathbb{N}\}$  be a sequence of nested graphs built by merging  $n$  copies of a graph  $G$  at a common vertex  $o$ :

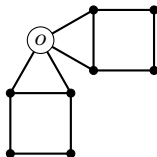
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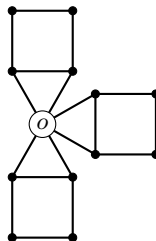
$G_1 = G$



$G_2$



$G_3$



*etc...*

...

# Limit theorem

## Theorem (Obata 2004)

Let  $\mu_o^{(n)}$  be the spectral measure of  $G_n$  rooted at  $o$ ,

$$\sqrt{n}\mu_o^{(n)}(\sqrt{n} \cdot) \xrightarrow{n \rightarrow \infty} \frac{1}{2}(\delta_{-\sqrt{d_o}} + \delta_{\sqrt{d_o}})$$

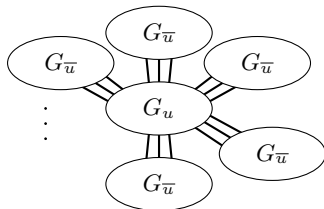
where  $d_o$  is the degree of  $o$  in  $G$ .

- Only depends of the immediate neighborhood in  $G$  (through  $d_o$ )
- Letting  $X_o^{(n)} \sim \mu_o^{(n)}$ :  $X_o^{(n)}/\sqrt{n} \xrightarrow[n \rightarrow \infty]{loi} \sqrt{d_o}B$  with  $B \sim \text{Rad}(1/2)$

# Generalization

The  $G_n, n \in \mathbb{N}$  are built by merging  $n$  copies of  $G$  rooted at a common subgraph  $G_u$  of  $p$  vertices  $u = \{1, \dots, p\}$ :

Graph  $G_n$



Adjacency matrix  $A^{(n)}$

$$\begin{bmatrix} A_{uu} & A_{u\bar{u}} & \dots & A_{u\bar{u}} \\ A_{\bar{u}u} & A_{\bar{u}\bar{u}} & 0 & 0 \\ \vdots & 0 & \ddots & 0 \\ A_{\bar{u}u} & 0 & 0 & A_{\bar{u}\bar{u}} \end{bmatrix}$$

## Theorem

Let  $\mu^{(n)}$  be the joint spectral measure  $G_n$  and  $(X_1^{(n)}, \dots, X_p^{(n)})$  a vector of the  $p$  quasi-random variables rooted at  $u$ ,

$$\frac{(X_1^{(n)}, \dots, X_p^{(n)})}{\sqrt{n}} \xrightarrow[n \rightarrow \infty]{d} (B_1 \sqrt{Y_1}, \dots, B_p \sqrt{Y_p}),$$

where

- $(Y_1, \dots, Y_p)$  is a quasi-random vector with distribution the joint spectral measure of  $D = A_{u\bar{u}} A_{\bar{u}u}$
- $B_1, \dots, B_p$  are iid Rademacher  $\text{Rad}(1/2)$ , independent from the  $Y_i$ 's

$$D_{ij} = \#\{\text{neighbors in } G_{\bar{u}} \text{ to both } i \text{ and } j\}, i, j \in u$$

Proof: (convergence of normalized moments)

$$\blacksquare \quad m_u^{(n)}[k_1, \dots, k_p] = \mathbb{E} \left[ \left( \frac{X_1^{(n)}}{\sqrt{n}} \right)^{k_1} \dots \left( \frac{X_p^{(n)}}{\sqrt{n}} \right)^{k_p} \right] = \det \left( \frac{A^{(n)}}{\sqrt{n}} [k_1, \dots, k_p, 0, \dots] \right)$$

$\implies$  we study the convergence of  $\left[ (A^{(n)}/\sqrt{n})^k \right]_{uu}$



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$$\blacksquare I + zA^{(n)} + z^2A^{(n)2} + \dots = \begin{bmatrix} I - zA_{uu} & A_{u\bar{u}} & \dots & A_{u\bar{u}} \\ A_{\bar{u}u} & I - zA_{\bar{u}\bar{u}} & 0 & 0 \\ \vdots & 0 & \ddots & 0 \\ A_{\bar{u}u} & 0 & 0 & I - zA_{\bar{u}\bar{u}} \end{bmatrix}^{-1}$$

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$\blacksquare$  Schur complement:

$$\begin{bmatrix} A & B \\ B^\top & C \end{bmatrix}^{-1} = \begin{bmatrix} (A - BC^{-1}B^\top)^{-1} & \cdot \\ \cdot & \cdot \end{bmatrix}$$

$$\blacksquare \left( (I - zA^{(n)})^{-1} \right)_{uu} = \left( I - zA_{uu} - nz^2 A_{u\bar{u}} (I - zA_{\bar{u}\bar{u}})^{-1} A_{\bar{u}u} \right)^{-1}$$

- $\left((I - zA^{(n)})^{-1}\right)_{uu} = \left(I - zA_{uu} - nz^2A_{u\bar{u}}(I - zA_{\bar{u}\bar{u}})^{-1}A_{\bar{u}u}\right)^{-1}$
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- $\left((I - \frac{z}{\sqrt{n}}A^{(n)})^{-1}\right)_{uu} \xrightarrow{n \rightarrow \infty} (I - z^2D)^{-1}$
- $m_u^{(n)}[k_1, \dots, k_p] \xrightarrow{n \rightarrow \infty} \begin{cases} \det\left(D[k_1/2, \dots, k_p/2]\right) & \text{if } k_1, \dots, k_p \text{ are even} \\ 0 & \text{sinon} \end{cases}$

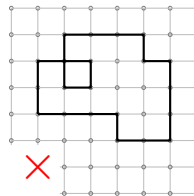
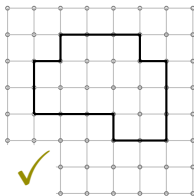
# Some combinatorics properties

Let  $A$  be the adjacency matrix of a graph  $G$

- $(A^k)_{ij}$  is the number of length  $k$  walks from  $i$  to  $j$  in  $G$
- Matrix generating function of walks in  $G$ :

$$z \mapsto (I - zA)^{-1} = I + zA + z^2A^2 + \dots$$

Let  $\mathcal{C}$  be the set of simple cycles on  $G$



Let  $V(c)$  the set of vertices met by  $c$  and  $|c|$  its length,

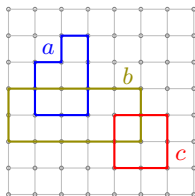
$$c \in \mathcal{C} \iff \#(V(c)) = |c|$$

## Characteristic polynomial

$$\det(I - zA) = \sum_{p \geq 0} (-1)^p \sum_{\substack{c_1, \dots, c_p \in \mathcal{C} \\ V(c_i) \cap V(c_j) = \emptyset, i \neq j}} z^{|c_1| + \dots + |c_p|}$$

We denote by  $\mathcal{H}$  the set of "words" formed by simple cycles, under the commutation rule

$$cc' = c'c \iff V(c) \cap V(c') = \emptyset.$$



$$abc \neq acb$$

$$acb = cab$$

$$bca = bac$$

$$etc...$$



We call *hike* an element of  $\mathcal{H}$ .

### Proposition

The function

$$\zeta(z) := \frac{1}{\det(I - zA)} = \sum_{h \in \mathcal{H}} z^{|h|}$$

is the hikes generating function.

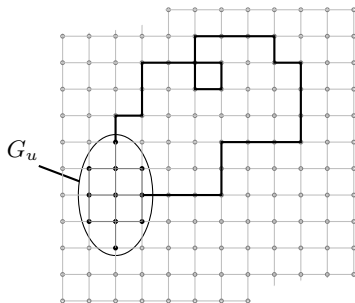
- $\zeta(\cdot)$  is the zeta function on  $\mathcal{H}$
- $z \mapsto \det(I - zA)$  is the Möbius function

### Remark

The closed walks on  $G$  are elements of  $\mathcal{H}$  with a unique right simple cycle.

## Definition

An excursion on  $u$  in a graph  $G$  is a walk that starts and ends in  $u$  but avoids  $u$  in between.

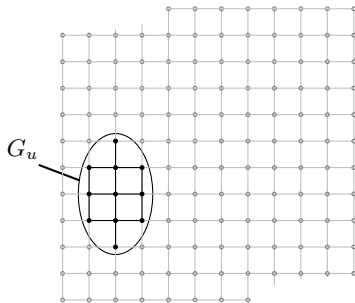


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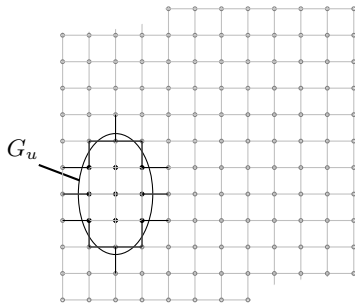


Generating function:

$$E_u(z) = zA_{uu} \dots$$

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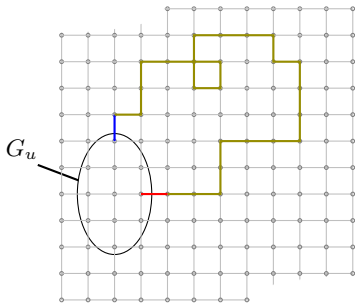
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Generating function:

$$E_u(z) = zA_{uu} + z^2A_{u\bar{u}}A_{\bar{u}u} \dots$$

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$$+ \sum_{k \geq 3} z^k A_{u\bar{u}} A_{\bar{u}u}^{k-2} A_{\bar{u}u}$$

Generating function of the excursions on  $u$

$$E_u(z) = zA_{uu} + z^2A_{u\bar{u}}(I - zA_{\bar{u}\bar{u}})^{-1}A_{\bar{u}u}$$

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$$I - E_u(z) = (I - zA)_{uu} - z^2 A_{u\bar{u}}((I - zA)_{\bar{u}\bar{u}})^{-1} A_{\bar{u}u}$$

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Unique factorisation of a walk from  $u$  to  $u$  as a product of excursions

$$I + E_u(z) + E_u(z)^2 + \dots = (I + zA + z^2A^2 + \dots)_{uu}$$

## Proposition

The function

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is the generating function of hikes whose right simple cycles all intersect  $u$ .

Or

$$\zeta_u(z) = \det \left( ((I - zA)^{-1})_{uu} \right) = \mathbb{E} \left( \prod_{i \in u} \frac{1}{1 - zX_i} \right)$$

Combinatorial expression on the homogenous  $k$ -moments of  $X_i, i \in u$ .