

# Edge universality of sparse Erdős-Rényi directed graphs

Yukun He

City University of Hong Kong

Random Matrices and Related Topics, Jeju, May 2024

## Undirected Erdős-Rényi graphs

$\hat{\mathcal{G}}(N, p)$ :  $N$  vertexes, every edge is connected with probability  $p$  completely independent of other edges.

Denote by  $\hat{A}_{ij} = \mathbf{1}_{i \sim j}$  the adjacency matrices of the Erdős-Rényi graphs.

We consider the normalized matrix and its centering by

$$\hat{A} := \frac{1}{\sqrt{Np(1-p)}} \hat{A}, \quad \hat{H} := \hat{A} - \sqrt{\frac{Np}{1-p}} |\mathbf{e}\rangle\langle \mathbf{e}|$$

so that  $\text{Var}(\hat{A}_{ij}) = \mathbb{E}\hat{H}_{ij}^2 = N^{-1}$ .

The largest eigenvalue of  $\hat{A}$  satisfies  $\hat{\lambda}_1 \approx \sqrt{\frac{Np}{1-p}} \gg 1$ . The key question is to understand the second largest eigenvalue  $\hat{\lambda}_2$ .

## Edge statistics

- [Erdős-Knowles-Yau-Yin, 2011], [Lee-Schnelli, 2016]

When  $p \gg N^{-2/3}$ , the extreme eigenvalues of  $A$  satisfy Tracy-Widom distribution

$$N^{2/3}(\hat{\lambda}_2 - \mathbb{E}\hat{\lambda}_2) \xrightarrow{d} \text{TW}_1.$$

- [Huang-Landon-Yau, 2017], [H.-Knowles 2020]

When  $N^{-1+\varepsilon} \leq p \ll N^{-2/3}$ , one has

$$\frac{\hat{\lambda}_2 - \mathbb{E}\hat{\lambda}_2}{\sqrt{2\mathbb{E}\hat{H}_{12}^4}} \xrightarrow{d} \mathcal{N}(0, 1).$$

The sparse contribution is on the scale

$$\sqrt{\mathbb{E}\hat{H}_{12}^4} \asymp \frac{1}{N\sqrt{p}} \gg N^{-2/3}$$

for  $p \ll N^{-2/3}$ , and it originates from

$$\frac{1}{N}(\text{Tr } \hat{H}^2 - N) \propto \text{Tr } \hat{A}^2 = \sum_{i,j} \hat{A}_{ij} \hat{A}_{ji} = \sum_{i,j} \hat{A}_{ij} = \sum_i d_i.$$

## Edge universality

[Jaehun Lee, 2021] [Huang-Yau, 2022] For  $N^{-1+\varepsilon} \leq p \leq 1/2$ , we have

$$N^{2/3}(\widehat{\lambda}_2 - \mathbb{E}\widehat{\lambda}_2 - \mathcal{X}) \xrightarrow{d} \text{TW}_1.$$

Here

$$\mathbb{E}\widehat{\lambda}_2 = 2 + (Np)^{-1} - \frac{5}{4}(Np)^{-2} + \dots$$

The first two orders of **random term**  $\mathcal{X}$  are

$$\frac{1}{N} \left( \sum_{ij} \widehat{H}_{ij}^2 - N \right)$$

and

$$\frac{1}{N} \left( \sum_{ij} \widehat{H}_{ij}^4 - \frac{1}{Np^2} \right) + \frac{2}{N} \sum_{ijk} \widehat{H}_{ij}^2 \left( \widehat{H}_{ik}^2 - \frac{1}{N} \right).$$

The third order terms of  $\mathcal{X}$  contain e.g. the fluctuation of

$$\sum_{ijkl} \widehat{H}_{ij} \widehat{H}_{jk} \widehat{H}_{kl}.$$

## Edge universality

[Jaehun Lee, 2021] [Huang-Yau, 2022] For  $N^{-1+\varepsilon} \leq p \leq 1/2$ , we have

$$N^{2/3}(\widehat{\lambda}_2 - \mathbb{E}\widehat{\lambda}_2 - \mathcal{X}) \xrightarrow{d} \text{TW}_1.$$

Here

$$\mathbb{E}\widehat{\lambda}_2 = 2 + (Np)^{-1} - \frac{5}{4}(Np)^{-2} + \dots$$

The first two orders of **random term**  $\mathcal{X}$  are

$$\frac{1}{N} \left( \sum_{ij} \widehat{H}_{ij}^2 - N \right) \propto \sum_i d_i$$

and

$$\frac{1}{N} \left( \sum_{ij} \widehat{H}_{ij}^4 - \frac{1}{Np^2} \right) + \frac{2}{N} \sum_{ijk} \widehat{H}_{ij}^2 \left( \widehat{H}_{ik}^2 - \frac{1}{N} \right) \propto \sum_i d_i + \sum_i d_i^2.$$

The third order terms of  $\mathcal{X}$  contain e.g. the fluctuation of

$$\sum_{ijkl} \widehat{H}_{ij} \widehat{H}_{jk} \widehat{H}_{kl} \propto \sum_{jk} d_j \widehat{A}_{jk} d_k.$$

## Random regular graphs

**$d$ -regular graph:** (undirected, simple) graphs on  $\{1, \dots, N\}$ , each vertex has  $d$  neighbours.

**Random  $d$ -regular graph:** uniform probability measure on the set of  $d$ -regular graphs. **Adjacency matrix**  $\tilde{A}_{ij} := \mathbf{1}_{i \sim j}$ .

The second largest eigenvalue of  $\tilde{A}$  satisfies

$$N^{2/3} \left( \frac{\tilde{\lambda}_2 + d/N}{\sqrt{(d-1)(N-d)/N}} - 2 \right) \xrightarrow{d} \text{TW}_1$$

- For  $N^{2/9} \ll d \ll N^{1/3}$  in [Bauerschmidt-Huang-Knowles-Yau, 2019]
- For  $N^\varepsilon \leq d \ll N^{1/3}$  in [Huang-Yau, 2023]
- For  $d \gg N^{2/3}$  in [H., 2022].

## Erdős-Rényi digraphs

$\mathcal{G}(N, p)$ :  $N$  vertexes, every **directed edge**  $(i, j)$  is connected with probability  $p$  completely **independent** of other edges.

The **adjacency matrix**  $\mathcal{A}$  has i.i.d. entries with

$$\mathcal{A}_{ij} = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1 - p. \end{cases}$$

We normalize

$$A = \frac{\mathcal{A}}{\sqrt{Np(1-p)}}.$$

Eigenvalues of  $A$  satisfies (global) **circular law**:

$$\frac{1}{N} \sum_i \delta_{\lambda_i} \longrightarrow \frac{1}{\pi} \mathbf{1}_{\{z \in \mathbb{C}: |z| \leq 1\}}$$

as  $N \rightarrow \infty$ .

# Local eigenvalue statistics

For **dense** non-Hermitian matrices, we have

- [Tao-Vu, 2012]: Under **four-moment matching** condition, we have **universality** of the local statistics in the bulk and near the edge.

Under **two-moment** matching

- [Bourgade-Yau-Yin, 2012]: Circular law on all mesoscopic scale.
- [Alt-Erdős-Krüger, 2016]: Spectral radius location.
- [Cipolloni-Erdős-Schröder, 2019]: Universality near the edge.
- [Cipolloni-Erdős-Xu, 2023]: Spectral radius distribution.
- [Maltsev-Osman, 2023; Dubova-Yang 2024]: Bulk universality.

There is **no previous results** on the local statistics of **sparse** non-Hermitian matrices.



## Main results

For a square matrix  $S \in \mathbb{C}^{N \times N}$  with eigenvalues  $\lambda_1^S, \dots, \lambda_N^S$ , we define its  $k$ -point correlation function  $p_k^S$  through

$$\int_{\mathbb{C}^k} F(z_1, \dots, z_k) p_k^S(z_1, \dots, z_k) dz_1 \dots dz_k = \binom{N}{k}^{-1} \mathbb{E} \sum_{i_1, \dots, i_k=1}^N F(\lambda_{i_1}^S, \dots, \lambda_{i_k}^S).$$

[H. 2023] (Edge Universality)

Fix  $k \in \mathbb{N}_+$  and let  $w_1, \dots, w_k \in \mathbb{T} = \{w \in \mathbb{C}, |w| = 1\}$ . For  $N^{-1+\varepsilon} \leq p \leq 1/2$

$$\lim_{N \rightarrow \infty} \int_{\mathbb{C}^k} F(z_1, \dots, z_k) \left[ p_k^A \left( w_1 + \frac{z_1}{N^{1/2}}, \dots, w_k + \frac{z_k}{N^{1/2}} \right) - p_k^W \left( w_1 + \frac{z_1}{N^{1/2}}, \dots, w_k + \frac{z_k}{N^{1/2}} \right) \right] dz_1 \dots dz_k = 0$$

Here  $W \in \mathbb{R}^{N \times N}$  is the **real Ginibre ensemble**, i.e. it has i.i.d. Gaussian entries with mean 0 and variance  $1/N$ .

## Main results (Cont)

[H. 2023] Assume  $N^{-1+\varepsilon} \leq p \leq 1/2$  and let  $\lambda_1, \lambda_2, \dots, \lambda_N$  be the eigenvalues of  $A$  with  $|\lambda_1| = \max_i |\lambda_i| \approx \sqrt{\frac{Np}{1-p}} \gg 1$ .

1. (Spectral radius) We have

$$\max_{2 \leq i \leq N} |\lambda_i| = 1 + O(N^{-1/2+\varepsilon}) \quad (1)$$

with very high probability.

2. (Delocalization) Suppose  $\mathbf{u} \in \mathbb{C}^N$  satisfies  $A\mathbf{u} = \lambda\mathbf{u}$  for some  $\lambda \in \mathbb{C}$  with  $|\lambda| \leq 2$ . Then

$$\|\mathbf{u}\|_\infty = O(N^{-1/2+\varepsilon} \|\mathbf{u}\|)$$

with very high probability.

Observation. **No phase transition!** The edge statistics of **non-Hermitian** random matrices are more robust than those of **Hermitian** random matrices.

Simple explanation: the **sparse contribution** is on the scale

$$\sqrt{\mathbb{E}\widehat{H}_{12}^4} \asymp \sqrt{\mathbb{E}(A_{12} - \mathbb{E}A_{12})^4} \asymp \frac{1}{N\sqrt{p}}$$

for both cases, while **Tracy-Widom** is on scale  $N^{-2/3}$ , and the extreme eigenvalues of  $A$  fluctuates on scale  $N^{-1/2}$ .

Detailed explanation: proof outline.

# Proof outline

## Girko's Hermitization

$$\frac{1}{N} \sum_i f(\sqrt{N}(\lambda_i - w_*)) = \frac{i}{4\pi N} \int_{\mathbb{C}} \int_0^\infty \nabla_w^2 f(\sqrt{N}(w - w_*)) \operatorname{Tr}(H_w - i\eta)^{-1} d\eta d^2w$$

with  $|w_*| = 1$ , and

$$H_w := \begin{pmatrix} 0 & A - w \\ A^* - \bar{w} & 0 \end{pmatrix} \in \mathbb{C}^{2N \times 2N}.$$

The **key** is to study  $g(w, i\eta) := \frac{1}{2N} \operatorname{Tr}(H_w - i\eta)^{-1}$ . It is close to a deterministic  $m \equiv m(w, i\eta)$ , which solves

$$P(m) := m^3 + 2i\eta m^2 + (1 - \eta^2 - |w|^2)m + i\eta = 0.$$

Estimate  $g$  through

$$g - m = O\left(\frac{P(g)}{P'(m)}\right),$$

and near the edge it is greatly affected by the **cuspid singularity**

$$P'(m) = 3m^2 + 4i\eta m + (1 - \eta^2 - |w|^2) \asymp |1 - |w|| + \eta^{2/3}.$$

It originates from the smallness of  $m$

$$m = i \operatorname{Im} m = O(|1 - |w||^{1/2} + \eta^{1/3}).$$

This requires very precise estimates of  $P(g)$ .

**Observation:** cusp singularity is an **advantage** on estimating higher order terms in the sparse regime. For instance

$$\mathbb{E}P(g) = O\left(\frac{\mathbb{E}g^3}{Np} + \frac{\mathbb{E}g^5}{N^2p^2} + \dots\right)$$

Due to the smallness of  $m$

$$\frac{\mathbb{E}g^3}{Np} \approx \frac{m^3}{Np} = O\left(\frac{|1 - |w||^{3/2} + \eta}{Np}\right) = O\left(P'(m)\frac{|1 - |w||^{1/2} + \eta^{1/3}}{Np}\right).$$

Sparsity **does not** affect the estimate!

**Observation:** cusp singularity is an **advantage** on estimating higher order terms in the sparse regime. For instance

$$\mathbb{E}P(g) = O\left(\frac{\mathbb{E}g^3}{Np} + \frac{\mathbb{E}g^5}{N^2p^2} + \dots\right)$$

Due to the smallness of  $m$

$$\frac{\mathbb{E}g^3}{Np} \approx \frac{m^3}{Np} = O\left(\frac{|1 - |w||^{3/2} + \eta}{Np}\right) = O\left(P'(m) \frac{|1 - |w||^{1/2} + \eta^{1/3}}{Np}\right).$$

Sparsity **does not** affect the estimate!

In the **Hermitian case**,  $\hat{g} = N^{-1} \text{Tr}(\hat{A} - z)^{-1}$  is close to the solution of  $1 + z\hat{m} + \hat{m}^2 = 0$ , while

$$\mathbb{E}(1 + z\hat{g} + \hat{g}^2) = O\left(\frac{\mathbb{E}\hat{g}^4}{Np} + \frac{\mathbb{E}\hat{g}^6}{N^2p^2} + \dots\right)$$

and

$$\frac{\mathbb{E}\hat{g}^4}{Np} \approx \frac{\hat{m}^4}{Np} = O\left(\frac{1}{Np}\right), \quad \text{as } |\hat{m}| \approx 1!$$

We can prove that

$$|g - m| \lesssim \frac{1}{N\eta} + \frac{\eta^{1/3}}{Np} \quad (*)$$

with very high probability. For  $\eta \leq N^{-3/4}$ , we get  $|g - m| \lesssim \frac{1}{N\eta}$ .

In Hermitization

$$\frac{1}{N} \sum_i f(\sqrt{N}(\lambda_i - w_*)) = \frac{i}{2\pi} \int_{\mathbb{C}} \int_0^\infty \nabla_w^2 f(\sqrt{N}(w - w_*)) g(w, i\eta) d\eta d^2w,$$

we still need to treat

$$\frac{i}{2\pi} \int_{\mathbb{C}} \int_{N^{-3/4}}^\infty \nabla_w^2 f(\sqrt{N}(w - w_*)) g(w, i\eta) d\eta d^2w,$$

while (\*) is **unimprovable**.



**Idea:** Integration by-parts for the shift.

$$\begin{aligned} & \int_{\mathbb{C}} \int_{N^{-3/4}}^{\infty} \nabla_w^2 f(\sqrt{N}(w - w_*)) g(w, i\eta) d\eta d^2w \\ &= -4 \int_{\mathbb{C}} \int_{N^{-3/4}}^{\infty} \partial_{\bar{w}} f(\sqrt{N}(w - w_*)) \partial_w g(w, i\eta) d\eta d^2w \\ &= -\frac{2i}{N} \int_{\mathbb{C}} \int_{N^{-3/4}}^{\infty} \partial_{\bar{w}} f(\sqrt{N}(w - w_*)) \sum_{i=1}^N (G^2)_{i+N,i} d\eta d^2w \quad G := (H_w - i\eta)^{-1} \\ &= \frac{2i}{N} \int_{\mathbb{C}} \partial_{\bar{w}} f_{w_*} \sum_{i=1}^N G_{i+N,i}(iN^{-3/4}) d^2w. \end{aligned}$$

**Idea:** Integration by-parts for the shift.

$$\begin{aligned} & \int_{\mathbb{C}} \int_{N^{-3/4}}^{\infty} \nabla_w^2 f(\sqrt{N}(w - w_*)) g(w, i\eta) d\eta d^2w \\ &= -4 \int_{\mathbb{C}} \int_{N^{-3/4}}^{\infty} \partial_{\bar{w}} f(\sqrt{N}(w - w_*)) \partial_w g(w, i\eta) d\eta d^2w \\ &= -\frac{2i}{N} \int_{\mathbb{C}} \int_{N^{-3/4}}^{\infty} \partial_{\bar{w}} f(\sqrt{N}(w - w_*)) \sum_{i=1}^N (G^2)_{i+N,i} d\eta d^2w \quad G := (H_w - i\eta)^{-1} \\ &= \frac{2i}{N} \int_{\mathbb{C}} \partial_{\bar{w}} f_{w_*} \sum_{i=1}^N G_{i+N,i} (iN^{-3/4}) d^2w. \end{aligned}$$

From  $\nabla_w^2 f(\sqrt{N}(w - w_*))g(w, i\eta)$  to  $\partial_{\bar{w}} f(\sqrt{N}(w - w_*))\frac{1}{N} \sum_{i=1}^N (G^2)_{i+N,i}$ , the (naive) scale change is

$$\frac{1}{\sqrt{N}} \times \frac{1}{\eta} = N^{1/4} \quad \text{for } \eta = N^{-3/4}.$$

However, we can show that

$$\frac{1}{N} \sum_i G_{i+N,i} + \frac{1+m^2}{w} \lesssim \frac{1}{N^2\eta^2} \quad \left( \text{instead of } g - m \lesssim \frac{1}{N\eta} \right).$$

Hence the **actual** scale change is

$$\frac{1}{\sqrt{N}} \times \frac{1}{\eta} \times \frac{1}{N\eta} = 1 \quad \text{for } \eta = N^{-3/4}.$$

Thank you!