Fluctuations of extreme eigenvalues of sparse Erdős-Rényi graphs

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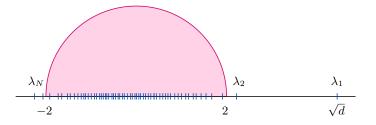
With Yukun He

Erdős-Rényi graph

Let $A = (A_{ij})$ be the adjacency matrix of the Erdős-Rényi graph $\mathbb{G}(N, d/N)$. $0 < d \leq N/2$ is the expected degree.

Denote by $\lambda_1 \ge \lambda_2 \ge \cdots \ge \lambda_N$ the eigenvalues of $A/\sqrt{d(1-d/N)}$.

If $d\to\infty$ as $N\to\infty$ then the empirical measure $\frac{1}{N}\sum_i \delta_{\lambda_i}$ converges to the semicircle law.



Goal: fluctuations of individual eigenvalues near the edges, e.g. λ_2 and λ_N .

Previous results

Notation: write $X \ll Y$ to mean $X = O_{\varepsilon}(N^{-\varepsilon}Y)$ for some $\varepsilon > 0$.

• [Erdős, K, Yau, Yin; 2012]: If $d \gg N^{2/3}$ then λ_2 has GOE Tracy-Widom fluctuations:

$$N^{2/3}(\lambda_2 - \mathbb{E}\lambda_2) \xrightarrow{\mathrm{d}} \mathrm{TW}_1$$

- [Lee, Schnelli; 2016]: The same holds for $d \gg N^{1/3}$.
- [Huang, Landon, Yau; 2017]: If $N^{2/9} \ll d \ll N^{1/3}$ then λ_2 has Gaussian fluctuations:

$$\sqrt{Nd} \left(\lambda_2 - \mathbb{E}\lambda_2\right) \stackrel{\mathrm{d}}{\longrightarrow} \mathcal{N}(0,2).$$

Crossover from Tracy-Widom to Gaussian fluctuations at $d \simeq N^{1/3}$.

Remarks

• The expectation $\mathbb{E}\lambda_2$ undergoes a shift, first observed by [Lee, Schnelli; 2016] and refined by [Huang, Landon, Yau; 2017]

$$\mathbb{E}\lambda_2 = 2 + \frac{1}{d} - \frac{5}{4d^2} + \cdots.$$

• The Gaussian fluctuations are explicitly given by the random variable

$$\begin{split} \mathcal{Z} &:= \frac{\mathcal{D}}{d} - 1 \,, \qquad \mathcal{D} := \frac{1}{N} \sum_{i,j} A_{ij} \quad \text{(average degree)} \,, \\ \text{satisfying } \sqrt{Nd} \mathcal{Z} \overset{\mathrm{d}}{\longrightarrow} \mathcal{N}(0,2). \end{split}$$

Then [Huang, Landon, Yau; 2017] show with very high probability

$$|\lambda_2 - \mathbb{E}\lambda_2 - \mathcal{Z}| \leqslant N^{o(1)} \left(\frac{1}{N^{2/3}} + \frac{1}{d^3}\right).$$

Result

Theorem [He, K; 2020]. If $1 \ll d \ll N^{1/3}$ then

$$\sqrt{Nd} (\lambda_2 - \mathbb{E}\lambda_2) \stackrel{\mathrm{d}}{\longrightarrow} \mathcal{N}(0,2).$$

In fact, we prove a rigidity result for all eigenvalues:

$$\lambda_i = (\mathbb{E}\lambda_i) \left(1 + \frac{\mathcal{Z}}{2} \right) + O\left(\frac{N^{-c}}{\sqrt{Nd}}\right)$$

with very high probability for $2 \leq i \leq N$. (Shown in bulk by [He; 2019].)

Interpretation: the Gaussian fluctuations vanish to leading order if one replaces the deterministic scaling $A/\sqrt{d(1-d/N)}$ with the random scaling $A/\sqrt{\mathcal{D}(1-d/N)}$: $\sqrt{\frac{d}{\mathcal{D}}}\lambda_i = \mathbb{E}\lambda_i + O\left(\frac{N^{-c}}{\sqrt{Nd}}\right).$

Heurstic picture of eigenvalue fluctuations of sparse matrices

The fluctuations of the eigenvalues of a sparse random matrix consists of two components:

- Random matrix component coinciding with fluctuations of GOE.
 Order √log N/N in bulk ([Gustavsson; 2005], [O'Rourke; 2010])
 Order N^{-2/3} at edge.
- Sparseness component from \mathcal{Z} .

Order $1/\sqrt{Nd}$ everywhere except origin.

Sparseness component dominates over random matrix component for $d \ll N$ in bulk and $d \ll N^{1/3}$ at edge.

Proof

Essence of proof: rigidity bounds for eigenvalue locations with accuracy

$$\frac{N^{-c}}{\sqrt{Nd}}$$

Previously: accuracy

$$\frac{1}{N^{2/3}} + \frac{1}{d^a},$$

with a = 1 [Erdős, K, Yau, Yin; 2012], a = 2 [Lee, Schnelli; 2016], a = 3 [Huang, Landon, Yau; 2017].

Need to go beyond a 1/d-expansion to reach sparseness $d \gg 1$.

Main estimate: upper bound on spectrum of $H := (A - \mathbb{E}A)/\sqrt{d}$ with eigenvalues $\mu_1 \ge \mu_2 \ge \cdots \ge \mu_N$.

Use the Green function and its normalized trace

$$G(z) := (H - z)^{-1}, \qquad \underline{G}(z) := \frac{1}{N} \operatorname{Tr} G(z), \qquad z = E + \mathrm{i}\eta \in \mathbb{C}_+.$$

Simple observation: if

$$\operatorname{Im} \underline{G}(z) = \frac{1}{N} \sum_{i} \frac{\eta}{(E - \mu_i)^2 + \eta^2} \ll \frac{1}{N\eta}$$

then there is no eigenvalue in $[E-\eta,E+\eta].$

Main work: estimate $\operatorname{Im} \underline{G}(z)$ just outside the spectrum.

Starting point: construction of self-consistent polynomial from [Huang, Landon, Yau; 2017], generalizing [Lee, Schnelli; 2016].

Lemma [Huang, Landon, Yau]. There exists a deterministic polynomial

$$P_0(z,x) = 1 + zx + x^2 + \frac{a_2}{d}x^4 + \frac{a_3}{d^2}x^6 + \cdots$$

such that

$$\left|\mathbb{E}P_0(z,\underline{G}(z))\right| \leq N^{o(1)} \left(\frac{\mathbb{E}\operatorname{Im}\underline{G}(z)}{N\eta} + \frac{1}{N}\right)$$

for $\eta \gg 1/N$. Here a_2, a_3, \ldots are universal constants.

Note: this says nothing about the size of $P_0(z, \underline{G}(z))$. In fact, $P_0(z, \underline{G}(z))$ is typically much bigger than the RHS.

Define the random polynomial $P(z, x) := P_0(z, x) + Zx^2$.

Lemma [Huang, Landon, Yau]. There exists a random algebraic function $m : \mathbb{C}_+ \to \mathbb{C}_+$ satisfying

$$P(z,m(z))=0\,,$$

such that m is the Stieltjes transform of a random symmetric probability measure $\varrho.$ We have ${\rm supp}\, \varrho=[-L,L],$ where

$$L = L_0 + \mathcal{Z} + O\left(\frac{N^{o(1)}}{d\sqrt{Nd}}\right)$$

and

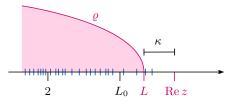
$$L_0 = 2 + \frac{b_1}{d} + \frac{b_2}{d^2} + \cdots$$

Here b_1, b_2, \ldots are universal constants (e.g. $b_1 = 1$, $b_2 = -5/4$).

Interpretation: \boldsymbol{L} is the approximate location of the right spectral edge, and we have to prove that

$$\mu_1 \leqslant L_0 + \mathcal{Z} + \frac{N^{-c}}{\sqrt{Nd}} \,.$$

To that end, choose $z = L_0 + \mathcal{Z} + w$ with $w = \kappa + \mathrm{i}\eta$ deterministic.



The proof follows from the following result.

Proposition. There exists $\eta \gg N^{-1}$ such that for $\kappa = \frac{N^{-c}}{\sqrt{Nd}}$ we have

$$\operatorname{Im} \underline{G}(z) \ll \frac{1}{N\eta} \,.$$

All previous works on eigenvalue rigidity use the following steps:

- (i) $|P(z,\underline{G}(z))| \ll 1$ with very high probability;
- (ii) $|\underline{G}(z) m(z)| \ll 1$ is small with very high probability, by inversion of self-consistent equation associated with P;
- (iii) $\operatorname{Im} \underline{G}(z) \leq \operatorname{Im} m(z) + |\underline{G}(z) m(z)| \ll \frac{1}{N\eta}$.

Problem: if d is small enough there is no choice of η such that both terms in (iii) are small enough.

Instead, in (iii) we estimate $|\text{Im} \underline{G}(z) - \text{Im} m(z)|$.

To that end, we estimate $|\text{Im} P(z, \underline{G}(z))|$ instead of $|P(z, \underline{G}(z))|$. Exploits a crucial cancellation which ensures that $|\text{Im} P(z, \underline{G}(z))| \ll |P(z, \underline{G}(z))|$.

How to invert self-consistent equation associated with $\operatorname{Im} P$? Expand

$$P(z,\underline{G}) = \partial_2 P(z,m)(\underline{G}-m) + \frac{1}{2}\partial_2^2 P(z,m)(\underline{G}-m)^2 + \cdots$$

As $\partial_2^2 P(z,m) \approx 2$, taking the imaginary part and rearranging terms yields

$$\operatorname{Re} \partial_2 P(z,m) \operatorname{Im}(\underline{G}-m) = \operatorname{Im} P(z,\underline{G}) - \operatorname{Im} \partial_2 P(z,m) \operatorname{Re}(\underline{G}-m) - 2 \operatorname{Re}(\underline{G}-m) \operatorname{Im}(\underline{G}-m) + \cdots$$

To solve this in $\operatorname{Im}(\underline{G}-m)$ we need

$$|\underline{G} - m| \ll |\operatorname{Re} \partial_2 P(z, m)| \asymp \sqrt{\kappa}.$$

Thus, the proof splits into two main steps:

(a) Show
$$|\underline{G} - m| \ll \sqrt{\kappa}$$
 by estimating $|P(z, \underline{G}(z))|$.

(b) Estimate $|\text{Im } P(z,\underline{G})|$.

Step (a) is harder because we cannot exploit the cancellation from Im.

Basic strategy for proving (a) and (b): recursive high moment estimates using cumulant expansion.

Recall that

$$P(z,x) = 1 + zx + Q(x) + Z$$
, $Q(x) := x^2 + \frac{a_2}{d}x^4 + \frac{a_3}{d^2}x^6 + \cdots$

Abbreviate $P=P(z,\underline{G}(z))\text{, }Q=Q(\underline{G}(z))$ and write

$$\mathbb{E}|P|^{2n} = \frac{1}{N} \sum_{i,j} \mathbb{E}H_{ij}G_{ji}P^{n-1}P^{*n} + \mathbb{E}(Q+\mathcal{Z})P^{n-1}P^{*n},$$

since

$$1 + z\underline{G} = \underline{H}\underline{G} \,.$$

Then use cumlant expansion (or generalized Stein lemma)

$$\mathbb{E}[h \cdot f(h)] = \sum_{k=0}^{\ell} \frac{1}{k!} \mathcal{C}_{k+1}(h) \mathbb{E}[f^{(k)}(h)] + \mathcal{R}_{\ell+1},$$

where $C_k(h)$ is the *k*th cumulant of *h*.

The cumulant expansion yields

$$\begin{split} \mathbb{E}|P|^{2n} &= \frac{1}{N} \sum_{k=1}^{\ell} \frac{1}{k!} \sum_{s=1}^{k} \binom{k}{s} \sum_{i,j} \mathcal{C}_{k+1}(H_{ij}) \mathbb{E}\left[\frac{\partial^{s}(P^{n-1}P^{*n})}{\partial H_{ij}^{s}} \frac{\partial^{k-s}G_{ij}}{\partial H_{ij}^{k-s}}\right] \\ &+ \frac{1}{N} \sum_{k=1}^{\ell} \frac{1}{k!} \sum_{i,j} \mathcal{C}_{k+1}(H_{ij}) \mathbb{E}\left[\frac{\partial^{k}G_{ij}}{\partial H_{ij}^{k}} P^{n-1}P^{*n}\right] + \mathbb{E}(Q+\mathcal{Z})P^{n-1}P^{*n} + \text{error} \,. \end{split}$$

The polynomial P is designed so that

$$\begin{split} \mathbb{E}P &= \frac{1}{N} \sum_{i,j} \mathbb{E}H_{ij} G_{ji} + \mathbb{E}(Q + \mathcal{Z}) \\ &= \frac{1}{N} \sum_{k=1}^{\ell} \frac{1}{k!} \sum_{i,j} \mathcal{C}_{k+1}(H_{ij}) \mathbb{E}\left[\frac{\partial^k G_{ij}}{\partial H_{ij}^k}\right] + \mathbb{E}(Q + \mathcal{Z}) + \text{error} \approx 0 \,, \end{split}$$

and for the same reason there is a near-exact cancellation between the blue terms above.

Dangerous terms: red ones, which capture the fluctuations of P.

Example of dangerous term (k = 3 and s = 2):

$$T := \frac{1}{N} \sum_{i,j} \mathcal{C}_4(H_{ij}) \mathbb{E} \Big[(\partial_2 P^*) N^{-1} (G^{*2})_{ii} G^*_{jj} G_{ii} G_{jj} |P|^{2n-2} \Big].$$

Unlike in [Erdős, K, Yau, Yin], [Lee, Schnelli], [Huang, Landon, Yau]: taking absolute value inside expectation is not affordable.

Instead, we have to exploit higher-order cancellations arising from construction of ${\cal P}.$ Use key identity

$$(\partial_2 P^*)\underline{G^{*2}} = \partial_{w^*} P(z^*, \underline{G}^*) - \underline{G^*}$$

and approximations (to be justified)

$$(G^{*2})_{ii} \approx \underline{G^{*2}}, \qquad G^*_{jj} \approx \underline{G^*}, \qquad G_{ii}, G_{jj} \approx \underline{G}$$

to write

$$(\partial_2 P^*) N^{-1}(G^{*2})_{ii} G^*_{jj} G_{ii} G_{jj} = N^{-1} \partial_{w^*} P(z^*, \underline{G}^*) \underline{G^*} \underline{G}^2 + \operatorname{error}.$$

Thus, up to error terms,

$$T = \frac{1}{N^2} \sum_{i,j} \mathcal{C}_4(H_{ij}) \mathbb{E} \Big[\partial_{w^*} P(z^*, \underline{G}^*) \underline{G}^* \underline{G}^2 |P|^{2n-2} \Big]$$

$$= \frac{1}{N^3} \sum_{i,j,k,l} \mathcal{C}_4(H_{ij}) \mathbb{E} \Big[\partial_{w^*} (H_{kl} G_{lk}^*) \underline{G}^* \underline{G}^2 |P|^{2n-2} \Big]$$

$$+ \frac{1}{N^2} \sum_{i,j} \mathcal{C}_4(H_{ij}) \mathbb{E} \Big[\partial_{w^*} (Q^* + \mathcal{Z}) \underline{G}^* \underline{G}^2 |P|^{2n-2} \Big]$$

Now apply cumulant expansion again, using that ∂_{w^*} and $\partial/\partial H_{ij}$ commute. Thus, we can exploit the cancellation built into P.

Need a systematic machinery to track the algebraic structure of all terms generated by this procedure and the corresponding higher-order cancellations. This is done by constructing a hierarchy of Schwinger-Dyson equations for a sufficiently large class of polynomials in $(G_{ij})_{i,i=1}^N$.