

Spectra of sparse random graphs

Let $G(n, \frac{d}{n})$ be the Erdős-Rényi random graph on n vertices.

Independently for each pair $i < j$, $P[i \sim j] = \frac{d}{n}$.

Let $A \in \mathbb{R}^{n \times n}$ be the adjacency matrix:

$$A_{ij} = \begin{cases} 1 & \text{if } i \sim j \\ 0 & \text{otherwise.} \end{cases}$$

Then $X_{ij} = \frac{1}{\sqrt{n \cdot \frac{d}{n} (1 - \frac{d}{n})}} (A_{ij} - \frac{d}{n} \mathbb{I}_{\{i \neq j\}})$ defines a Wigner matrix:

$$\bullet \mathbb{E}[X_{ij}] = 0, \mathbb{E}[(\sqrt{n} \cdot X_{ij})^2] = 1$$

$$\bullet \mathbb{E}[|\sqrt{n} X_{ij}|^k] = \left[\frac{d}{n} \cdot \left(\frac{1 - \frac{d}{n}}{\sqrt{\frac{d}{n} (1 - \frac{d}{n})}} \right)^k + \left(1 - \frac{d}{n} \right) \left(\frac{\frac{d}{n}}{\sqrt{\frac{d}{n} (1 - \frac{d}{n})}} \right)^k \right]$$
$$\sim (n/d)^{k/2+1} \quad \text{if } d \equiv d_n \ll n.$$

Previously, we assumed $\mathbb{E}[|\sqrt{n} X_{ij}|^k] \leq C_k$ for a constant $C_k > 0$ independent of n .

Using the types of techniques we've covered, one can show

• If $d \equiv d_n \gg 1$ (i.e. $d \rightarrow \infty$ at any rate with n) then the ESD of X converges (weakly in probability) to the semicircle law.

See Tran, Vu, Wang "Sparse random graphs," 2013.

• If $d \gg \log n$, then $\|X\| \rightarrow 2$ in probability.

See Benaych-Georges, Bordenave, Knowles "Spectral radii of sparse random matrices," 2018.

Q: What happens if d is constant in n ? This is called the dilute regime.

Consider $\mathbb{E}[\frac{1}{n} \text{Tr} X^k] = \frac{1}{n} \sum_{i_1, i_2, \dots, i_k} \mathbb{E}[X_{i_1 i_2} \dots X_{i_{k-1} i_k}]$.

Consider even k , and the path $i_1, i_2, i_1, i_2, \dots, i_1, i_2$.

The contribution from this path is

$$\frac{1}{n} \cdot n(n-1) \cdot \mathbb{E}[X_{ij}^k] \approx \frac{1}{n} \cdot n(n-1) \cdot \frac{d}{n} \left(\frac{1}{\sqrt{d}}\right)^k = d^{1-k/2}.$$

Does not vanish as $n \rightarrow \infty$, even though this visits only two vertices.

Consider $R(z) = (X - zI)^{-1}$. We have

$$R_{ii}(z) = \frac{1}{-z - x_i^T R^{(i)}(z) x_i}.$$

The quantity $x_i^T M x_i \approx \frac{1}{d} (a_i - \frac{d}{n})^T M (a_i - \frac{d}{n})$

does not concentrate around $\mathbb{E}[x_i^T M x_i]$ as $n \rightarrow \infty$:

E.g. for $M = I$, $\mathbb{E}[x_i^T M x_i] = \sum_j \mathbb{E}[x_{ij}^2] = 1$, and

$$\text{Var}[x_i^T M x_i] = \sum_j \mathbb{E}[(x_{ij}^2 - \frac{1}{n})^2] = \sum_j [\mathbb{E}[x_{ij}^4] - \frac{1}{n^2}] \approx \sum_j \frac{1}{n} = \frac{1}{d}.$$

Individual entries of the resolvent remain random in the limit $n \rightarrow \infty$.

Indeed: The ESD of A (and of X) converges to a limit distribution w/ unbounded support, which is not the semicircle law.

One implication: Consider a stochastic block model $G(n, \frac{a}{n}, \frac{b}{n})$ where

$$P[i \sim j] = \begin{cases} \frac{a}{n} & \text{if } i, j \leq \frac{n}{2} \text{ or } i, j \geq \frac{n}{2} + 1 \\ \frac{b}{n} & \text{if } i \leq \frac{n}{2}, j \geq \frac{n}{2} + 1 \text{ or } i \geq \frac{n}{2} + 1, j \leq \frac{n}{2}. \end{cases}$$

$$\mathbb{E}[A] = \begin{pmatrix} \frac{a}{n} & \frac{b}{n} \\ \frac{b}{n} & \frac{a}{n} \end{pmatrix} \text{ is rank two, with eigenvalues } \frac{a+b}{2} \text{ and } \frac{a-b}{2}.$$

For a dilute graph with a, b constant in n , the "signal eigenvalue" $\frac{a-b}{2}$ is swamped by the bulk of the spectrum as $n \rightarrow \infty$. Applying a spectral decomposition of A fails to uncover the community structure.

Graph spectra and local weak convergence

Def: Let G_n be a deterministic graph on n vertices, for each $n \geq 1$.

G_n converges locally weakly to an infinite, connected random graph

$\mathcal{G}$ with a distinguished root vertex o if:

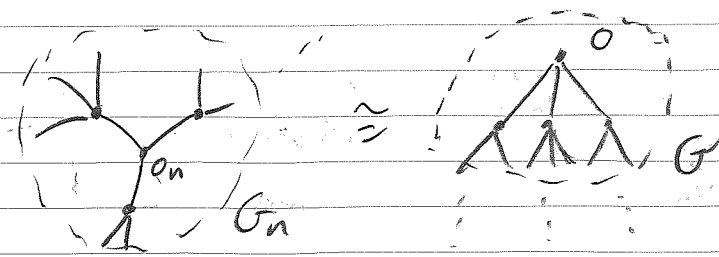
Pick $o_n \in \{1, \dots, n\}$ uniformly at random. Let $B_\ell(G_n, o_n)$ be the radius- ℓ ball around o_n in G_n , and $B_\ell(\mathcal{G}, o)$ be that around o in $\mathcal{G}$. Then for each fixed $\ell \geq 1$ and fixed rooted graph (T, o) ,

$$P_{o_n}[B_\ell(G_n, o_n) \cong (T, o)] \rightarrow P[B_\ell(\mathcal{G}, o) \cong (T, o)]$$

as $n \rightarrow \infty$. We write $G_n \xrightarrow{\text{wl}} (\mathcal{G}, o)$.

Note: $\cong$ denotes a rooted graph isomorphism, i.e. $(G_1, o_1) \cong (G_2, o_2)$ if there is a graph isomorphism mapping G_1 to G_2 and o_1 to o_2 .

The probability on the left is over the random choice of root o_n in G_n , and on the right over the distribution of $(\mathcal{G}, o)$.



Example: Let $G_n \sim G(n, \frac{d}{n})$ (Erdos-Renyi). Consider first $\mathbb{E}[\mathbb{P}_{G_n}[B_e(G_n, o) \cong (T, o)]]$, where outer $\mathbb{E}$ is over G_n .

This is $\mathbb{E}[\frac{1}{n} \sum_{o=1}^n \mathbb{I}\{B_e(G_n, o) \cong (T, o)\}]$
 $= \mathbb{P}[B_e(G_n, 1) \cong (T, o)]$, where now $\mathbb{P}$ is over G_n .

Note that:

- Vertex 1 has degree $D_1 \sim \text{Binom}(n-1, \frac{d}{n}) \xrightarrow{p} \text{Poisson}(d)$.

- Conditional on the D_1 neighbors of vertex 1,

$$\mathbb{P}[\text{no two nbrs of 1 are connected}] = (1 - \frac{d}{n})^{\binom{D_1}{2}} \rightarrow 1.$$

- Conditional on the D_1 neighbors of vertex 1, and the event that no two nbrs of 1 are connected, the first neighbor of 1

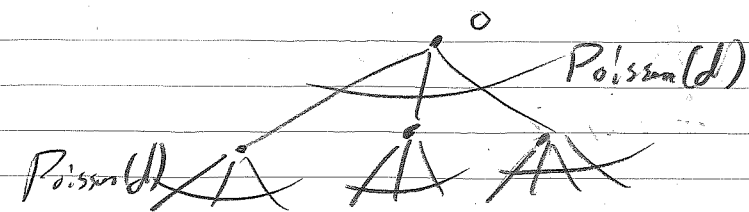
has $D_2 \sim \text{Binom}(n-1-D_1, \frac{d}{n}) \xrightarrow{p} \text{Poisson}(d)$ additional neighbors.

- Conditional on these D_2 neighbors,

$$\mathbb{P}[\text{no two of these } D_2 \text{ nbrs are connected}] = (1 - \frac{d}{n})^{\binom{D_2}{2}} \rightarrow 1.$$

- Etc...

So the law of $B_e(G_n, 1)$ converges to an ℓ -layer Galton-Watson tree with degree distribution $\text{Poisson}(d)$:



(Each Poisson d variable is independent.)

Letting (G, o) be the law of this infinite tree, for any fixed ℓ, T

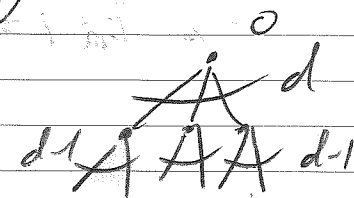
$$\mathbb{E}[\mathbb{P}_{G_n}[B_e(G_n, o) \cong (T, o)]] \rightarrow \mathbb{P}[B_e(G, o) \cong (T, o)].$$

In fact, can show

$$\mathbb{P}_{G_n}[B_e(G_n, o) \cong (T, o)] \rightarrow \mathbb{P}[B_e(G, o) \cong (T, o)]$$

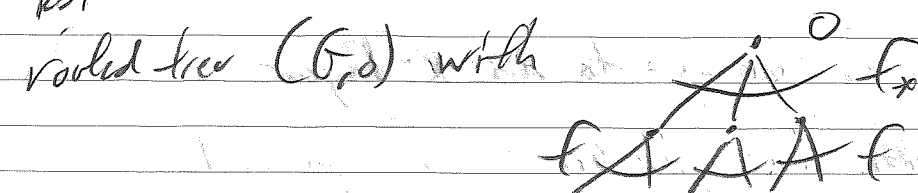
almost surely, and hence $G_n \xrightarrow{ul} (G, o)$ a.s. as $n \rightarrow \infty$.

Example: Let G_n be a d -regular graph chosen uniformly at random (from all d -regular graphs). Then $G_n \xrightarrow{ul} (G, o)$ a.s., the Galton-Watson tree with fixed degree d :



Note: Number of offspring of root o is different from number of offspring of other vertices.

Def: Let f_x be a probability mass function on $\{0, 1, 2, \dots\}$ s.t. $\sum_{k \geq 1} k f_x(k) < \infty$. The GW-tree with degree distribution f_x is the random



rooted tree (G, o) with



where $f(j-1) = j f_x(j) / \sum_{k \geq 1} k f_x(k)$.

Remark: $F_x = \text{Poisson}(d)$ is special:

$$\sum_{k \geq 1} k f_x(k) = \sum_{k \geq 1} k \cdot e^{-d} \cdot \frac{d^k}{k!} = d \cdot \sum_{k \geq 1} e^{-d} \cdot \frac{d^{k-1}}{(k-1)!} = d.$$

$$\text{So } j f_x(j) / \sum_{k \geq 1} k f_x(k) = j \cdot e^{-d} \cdot \frac{d^j}{j!} / d = e^{-d} \cdot \frac{d^{j-1}}{(j-1)!}, \text{ and } f = F_x.$$

In general, $f \neq F_x$. The form of f comes from the following:

Choose o uniformly at random. Conditional on $\deg(o) \geq 1$, choose a random neighbor v of o . Then, letting $N_n = \# \text{ vertices of } G_n \text{ w/ degree } \geq 1$

$$f(k-1) = \lim_{n \rightarrow \infty} \mathbb{P}[\deg(v) = k \mid \deg(o) \geq 1]$$

$$= \lim_{n \rightarrow \infty} \frac{1}{N_n} \sum_{o: \deg(o) \geq 1} \frac{1}{\deg(o)} \cdot \sum_{v \sim o} \mathbb{I}\{\deg(v) = k\}$$

$$= \lim_{n \rightarrow \infty} \underbrace{\frac{N_n}{N_n}}_{1} \cdot \underbrace{\frac{1}{N_n} \sum_{v: \deg(v) \geq 1} \frac{1}{\deg(v)}}_{\rightarrow F_x(k)} \cdot \underbrace{\sum_{o \sim v} \frac{1}{\deg(o)}}_{\rightarrow k \cdot \mathbb{P}_{j \sim f}[\frac{1}{1+j}], \text{ independent of } \mathbb{I}\{\deg(v) = k\}}$$

$$\rightarrow \frac{1}{1 - F_x(0)} \rightarrow F_x(k) \rightarrow k \cdot \mathbb{P}_{j \sim f}[\frac{1}{1+j}], \text{ independent of } \mathbb{I}\{\deg(v) = k\}.$$

$$\propto k \cdot F_x(k).$$

The limit of (G_n, o) is unimodular — “ G looks the same from any choice of root o .”

Then: (a) Suppose $\{\deg(o_n)\}_{n \geq 1}$ is uniformly integrable, and $G_n \xrightarrow{\text{w.h.p.}} (G, o)$ a.s. as $n \rightarrow \infty$. Then there exists an a.s. limit μ for the ESD of the adjacency matrix of G_n .

(b) Suppose, in addition, that $(G, o) \sim \text{GW tree w/ degree distribution } F_x$, and $\sum_{k \geq 1} k^2 f_x(k) < \infty$. Let $\mathcal{H}$ be the space of analytic functions $f: \mathbb{C}^+ \rightarrow \mathbb{C}^+$ s.t. $|f(z)| \leq \frac{1}{\text{Im } z}$. Then there is a unique

distribution Q over $\mathcal{H}$ s.t. if $Y(z) \sim Q$ then

$$Y(z) \stackrel{d}{=} -\left(z + \sum_{i=1}^N Y_i(z)\right)^{-1}$$

where $N \sim F$, $Y_1, Y_2, \dots \sim Q$ and these are independent. Furthermore, the Stieltjes transform of μ is $m(z) = \mathbb{E}[X(z)]$ where

$$X(z) \stackrel{d}{=} -\left(z + \sum_{i=1}^{N_x} Y_i(z)\right)^{-1},$$

and $N_x \sim F_x$, $Y_1, Y_2, \dots \sim Q$ are independent.

Intuition: Let A be the adjacency matrix of the infinite graph (G, o) , as an operator on $\ell_2(\mathbb{N})$. Let $R(z) = (A - zI)^{-1}$ be its resolvent. Then $m_n(z) \rightarrow \mathbb{E}[o^T R(z) o]$ a.s. where $o = (1, 0, 0, \dots)$ and the first coordinate is the root o . $X(z)$ above is equal in law to $o^T R(z) o$.

Proof of Thm:

① Uniqueness of solution Q : We apply the Banach fixed point theorem:

If (X, d) is a complete metric space and $\Phi: X \rightarrow X$ a contraction ($d(\Phi(x), \Phi(y)) \leq c d(x, y) \forall x, y \in X$, and some $c \in [0, 1)$) then there is a unique

$x_* \in X$ where $\Phi(x_*) = x_*$. Furthermore, $x_* = \lim_{k \rightarrow \infty} \Phi^k(x)$ for any $x \in X$.

Let $X = \mathcal{P}(\mathcal{H})$ be the space of probability distributions on $\mathcal{H}$.

Note that $\mathbb{E}[N] = \sum_{j \geq 1} j f(j) < \infty$ by the condition $\sum_{k \geq 1} k^2 f_x(k) < \infty$.

Define $\mathcal{D} = \{z \in \mathbb{C}^+ : \text{Re } z \in [-1, 1], \text{Im } z \in [\sqrt{\mathbb{E}N+1}, \sqrt{\mathbb{E}N+2}]\}$, and

$$d(P, Q) = \inf_{\text{couplings of } P \text{ and } Q} \mathbb{E}_{(X, Y) \sim (P, Q)} \int_{\mathcal{D}} |X(z) - Y(z)| dz.$$

Note: If $d(P, Q) = 0$, then $\exists$ coupling where $X(z) = Y(z)$ for a.e. $z \in \mathcal{D} \Rightarrow$

$X(z) = Y(z) \forall z \in \mathbb{C}^+$ by analyticity. So $P=Q$. Can check d is a complete metric on $\mathcal{P}(\mathbb{H})$.

Define $\Psi: \mathcal{P}(\mathbb{H}) \rightarrow \mathcal{P}(\mathbb{H})$ s.t. $\Psi(Q)$ is the law of $-(z + \sum_{i=1}^N Y_i(z))^{-1}$, where $N \sim F$ and $Y_1, Y_2, \dots \sim Q$. For any $P, Q \in \mathcal{P}(\mathbb{H})$, let (X_i, Y_i) have the coupled law s.t. $d(P, Q) = \mathbb{E} \int_D |X(z) - Y(z)| dz$. Then

$$\begin{aligned} d(\Psi(P), \Psi(Q)) &\leq \mathbb{E} \int_D \left| -\frac{1}{z + \sum_{i=1}^N X_i(z)} + \frac{1}{z + \sum_{i=1}^N Y_i(z)} \right| dz \\ &= \mathbb{E} \int_D \left| \frac{\sum_{i=1}^N X_i(z) - Y_i(z)}{(z + \sum_{i=1}^N X_i(z))(z + \sum_{i=1}^N Y_i(z))} \right| dz \\ &\leq \int_D \frac{1}{(\operatorname{Im} z)^2} \cdot \mathbb{E} \left[\sum_{i=1}^N |X_i(z) - Y_i(z)| \right] dz \\ &\leq \int_D \frac{\mathbb{E}[N]}{(\operatorname{Im} z)^2} \cdot d(P, Q) dz \leq \frac{\mathbb{E}[N]}{\mathbb{E}[N]+1} \cdot d(P, Q). \end{aligned}$$

So unique Q where $\Psi(Q) = Q$, by Banach Fixed point theorem.

② Degree truncation: Fix a constant $D > 0$. Let $\tilde{G}_n$ be the pruned graph removing all edges adjacent to vertices w/ degree $\geq D$. Let $(\tilde{G}, o)$ be the rooted graph s.t.

$$\tilde{G} = \begin{cases} \text{single vertex } o & \text{if } \deg(o) \geq D \text{ in } G \\ \text{pruned version of } G & \text{otherwise.} \end{cases}$$

Then $\tilde{G}_n \xrightarrow{w.e.} (\tilde{G}, o)$. Furthermore, in (b), $(\tilde{G}, o)$ is a GW-tree w/ some offspring distribution $\tilde{F}_x$ for the root and $\tilde{F}$ for other vertices. By previous remark, $\tilde{F}(j+1) = j \tilde{F}_x(j) / \sum_{k \geq 1} k \tilde{F}_x(k)$.

Suppose $\text{ESD of } \tilde{A}_n \xrightarrow{L} \tilde{\mu}(D)$ a.s. $\tilde{A}_n$ differs from A_n in at most $\sum_{i=1}^n \mathbb{1}\{\deg(i) \geq D\} (\deg(i)+1)$ rows, and

$$\frac{1}{n} \sum_{i=1}^n \mathbb{1}\{\deg(i) \geq D\} (\deg(i)+1) \rightarrow \mathbb{E}[\mathbb{1}\{\deg(o) \geq D\} (\deg(o)+1)] = \varepsilon_D \text{ a.s. by}$$

$G_n \xrightarrow{w.e.} (G, o)$ and the given uniform integrability. We have

$$\text{Lévy-distance}(\text{ESD of } \tilde{A}_n, \text{ESD of } A_n) \leq \frac{1}{n} \text{rank}(A_n - \tilde{A}_n), \text{ so}$$

$$\limsup_{n \rightarrow \infty} \text{Lévy-distance}(\text{ESD of } A_n, \tilde{\mu}(D)) \leq \varepsilon_D.$$

Since $\varepsilon_D \rightarrow 0$ as $D \rightarrow \infty$, this implies $\tilde{\mu}(D) \xrightarrow{L} \tilde{\mu}$ as $D \rightarrow \infty$ for some $\tilde{\mu}$, and

$$\text{ESD of } A_n \xrightarrow{L} \tilde{\mu} \text{ a.s.}$$

Furthermore, if $\tilde{X}^{(D)}(z)$ is the law of $X(z)$ corresponding to $\tilde{F}_x$, then

$\tilde{X}^{(D)}(z) \xrightarrow{L} X(z)$. Letting $Q, \tilde{Q}$ be fixed points of Ψ corresponding to $N \sim F$ and $\tilde{N} \sim \tilde{F}$, and optimally coupling $(N, \tilde{N})$ and $(X_i, \tilde{Y}_i) \sim (Q, \tilde{Q})$ as before,

$$\begin{aligned} d(Q, \tilde{Q}) &\leq \mathbb{E} \int_D \mathbb{1}\{N \neq \tilde{N}\} \cdot \left| -\frac{1}{z + \sum_{i=1}^N X_i(z)} + \frac{1}{z + \sum_{i=1}^{\tilde{N}} Y_i(z)} \right| dz \\ &\quad + \mathbb{E} \int_D \mathbb{1}\{N = \tilde{N}\} \cdot \left| -\frac{1}{z + \sum_{i=1}^N X_i(z)} + \frac{1}{z + \sum_{i=1}^{\tilde{N}} Y_i(z)} \right| dz \\ &\leq \frac{C}{\operatorname{Im} z} \cdot \mathbb{P}[N \neq \tilde{N}] + \frac{1}{(\operatorname{Im} z)^2} \mathbb{E}[N \cdot \mathbb{1}\{N = \tilde{N}\}] \cdot d(Q, \tilde{Q}). \end{aligned}$$

So $d(Q, \tilde{Q}) \leq C \cdot \mathbb{P}[N \neq \tilde{N}]$ for fixed z . Similarly, letting $P, \tilde{P}$ be laws of $X(z)$ and $\tilde{X}(z)$, $d(P, \tilde{P}) \leq C \cdot \mathbb{P}[N_x \neq \tilde{N}_x] + C' \cdot d(Q, \tilde{Q})$.

So as $D \rightarrow \infty$, $\mathbb{P}[N \neq \tilde{N}] \rightarrow 0$ and $\mathbb{P}[N_x \neq \tilde{N}_x] \rightarrow 0$, implying $d(P, \tilde{P}) \rightarrow 0$, i.e., $\tilde{X}^{(D)}(z) \xrightarrow{L} X(z)$.

Thus: It suffices to show both (a) and (b) assuming G_n has max degree D .

③ Argument for bounded degree: By Skorokhod Representation, construct $\{G_n\}_{n \geq 1}$ and (G, o) on a common probability space, s.t. for all fixed l , $B_l(G_n, o_n) \cong B_l(G, o)$ a.s. for all large n .

Then also $B_{L_n}(G_n, o_n) \cong B_{L_n}(G, o)$ for some sequence $L_n \rightarrow \infty$, a.s. for all large n . This implies

$$A_n e_k \rightarrow A e_k \text{ a.s. for each fixed } k,$$

where A is the adjacency matrix of G w/ first vertex o , and

A_n is the adjacency matrix of G_n w/ first vertex o_n .

When G_n, G have bounded max degree, A_n, A are bounded self-adjoint operators on $\ell_2(N)$. Then also

$$(A_n - zI)^{-1} e_k \rightarrow (A - zI)^{-1} e_k \text{ a.s. for each fixed } k.$$

In particular, $(A_n - zI)^{-1}_{o_n o_n} \rightarrow (A - zI)^{-1}_{oo}$ a.s. Since both sides are bounded by $\frac{1}{\text{Im } z}$, dominated convergence yields

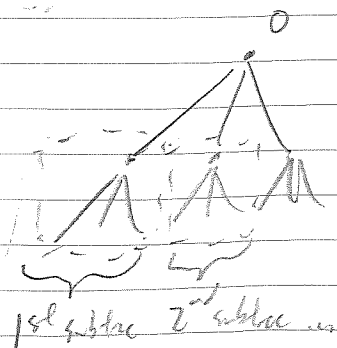
$$m_n(z) = \mathbb{E}_{o_n}[(A_n - zI)^{-1}_{o_n o_n}] \rightarrow \mathbb{E}[(A - zI)^{-1}_{oo}] \equiv m(z).$$

Here $m: \mathbb{C}^+ \rightarrow \mathbb{C}^+$ and $|m(z)| \leq \frac{1}{\text{Im } z}$, so m is the Stieltjes transform of a measure μ . This shows (a).

For (b), consider (G, o) truncated to level L from o . Call this $(G^{(L)}, o)$, and the associated adjacency matrix $A^{(L)}$. Write

$$A^{(L)} - zI = \begin{pmatrix} -z & 1 & 0 & \dots & 0 & 1 & 0 & \dots & 0 \\ 0 & A_1^{(L)} - zI & 0 & & & & & & \\ 0 & & A_2^{(L)} - zI & & & & & & \\ 0 & & & \ddots & & & & & \\ 0 & & & & 0 & A_{L-1}^{(L)} - zI & & & \\ 0 & & & & & & & & \ddots \end{pmatrix}$$

first subtree second subtree



Then by Schur complement,

$$(A^{(L)} - zI)^{-1}_{oo} = \frac{1}{-z - \sum_{i=1}^{N_o} (A_i^{(L)} - zI)^{-1}_{o_i o_i}}$$

where o_i is the root of the i -th subtree, and $N_o = \deg(o)$. Similarly,

$$(A_i^{(L)} - zI)^{-1} = \frac{1}{-z - \sum_{j=1}^{N_i} (A_{ij}^{(L)} - zI)^{-1}_{o_i o_j}}$$

where o_j is the root of the (ij) -th subtree, and $N_i = \# \text{offspring of } o_i$.

Apply this recursively: Let $Q_0 \in \mathcal{P}(\mathbb{R})$ be the (deterministic) law of $Y(z) \equiv -\frac{1}{z}$.

Then $(A_i^{(L)} - zI)^{-1} \equiv \Psi^{L-2}(Q_0)$. As $L \rightarrow \infty$, this converges in law to the fixed-point Q of Ψ . Then $(A^{(L)} - zI)^{-1}_{oo} \rightrightarrows X(z)$. By some argument as in part (a), $(A^{(L)} - zI)^{-1}_{oo} \xrightarrow{\text{a.s.}} (A - zI)^{-1}_{oo}$, since $A^{(L)}, A$ are bounded self-adjoint operators. So $(A - zI)^{-1}_{oo} \equiv X(z)$, and $m(z) = \mathbb{E}[X(z)]$.

Example: Let G_n be a uniform d -regular graph. Then $G_n \xrightarrow{\text{w.h.p.}} (G, o)$, where $\mathbb{P}_x(d) = 1$. The fixed-point equation is

$$Y(z) \equiv -\left(z + \sum_{i=1}^{d-1} Y_i(z)\right)^{-1}$$

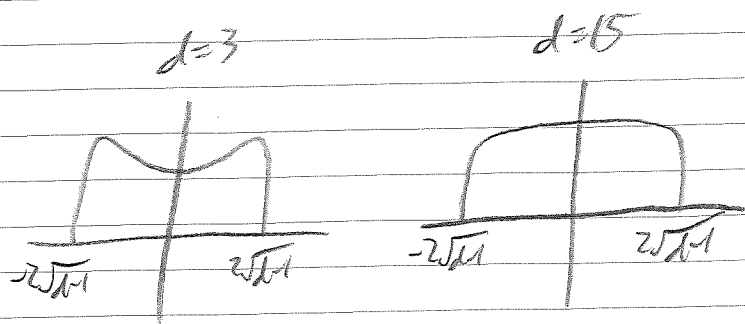
This has a deterministic solution solving $Y(z) = -(z + (d-1)Y(z))^{-1}$.

In fact $Y(z)$ is the Stieltjes transform of the semi-circle law w/ support $[-2\sqrt{d-1}, 2\sqrt{d-1}]$. Then $X(z)$ is also deterministic, given by

$$X(z) = -\frac{1}{z + dY(z)} = -\frac{2(d-1)}{(d-2)z + 2\sqrt{z^2 - 4(d-1)}}.$$

Then $\lim_{\eta \rightarrow 0} \frac{1}{\pi} \text{Im } X(\lambda + i\eta) = \frac{d}{2\pi} \frac{\sqrt{4(d-1) - \lambda^2}}{d^2 - \lambda^2}$ is the limiting density.

This is called the Kesten-McKay Law.



As $d \rightarrow \infty$, this approaches the semi-circle law.

pk: $G_n \sim G(n, \frac{d}{n})$ Erdős-Rényi. Here, $f_{sp} = f = \text{Poisson}(d)$.

The fixed point $Y(z)$ and correspondingly $X(z)$ are random. This corresponds to individual resonant entries $R(z)$ being random in the large- n limit. There is no simple form for the law of $Y(z)$.

It is possible to write a fixed-point equation for the moment-generating function $\ell(u, z) = \mathbb{E}[e^{uY(z)}]$:

$$\ell(u, z) = 1 - \sqrt{u} \int_0^\infty \frac{J_1(2\sqrt{ut})}{\sqrt{t}} \cdot \exp(-itz + d(\ell(t, z) - 1)) dt,$$

for $J_1(t) = \frac{t}{2} \sum_{k=0}^\infty \frac{(-t^2/4)^k}{k!(k+1)!}$ the Bessel function of the first kind,