

How fast do quantum walks mix?

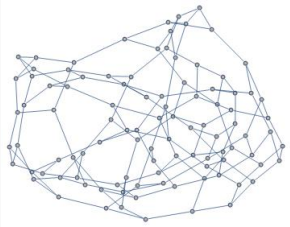
Shantanav Chakraborty¹, Kyle Luh² and Jérémie Roland¹

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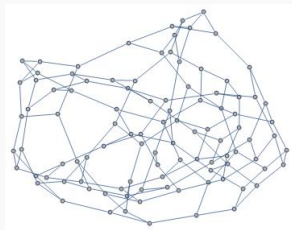
1 QuIC, Université libre de Bruxelles
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IIIT Theory Group Saturday Seminars, IIIT Hyderabad
May 22, 2021

Mixing of Classical random walk on a graph

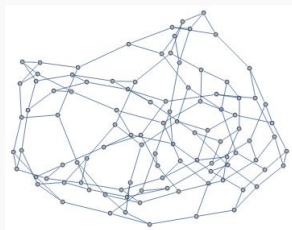


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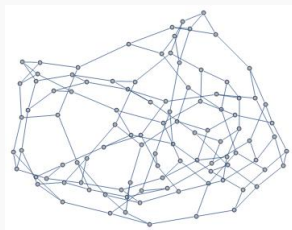
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- Classical random walk on G is defined by an $n \times n$ stochastic matrix P such that $P_{ij} = 1/d_i$, where d_i is the degree of node i .
- If the row-vector v_0 is the initial state of the walker, after t -steps:
 $v_t = v_0 P^t$.

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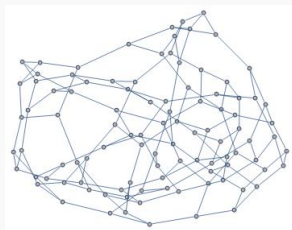
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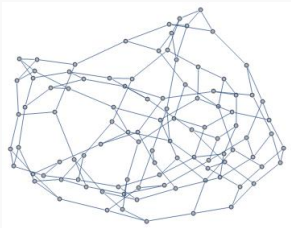
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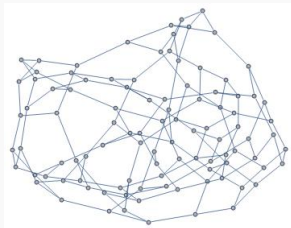


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 - π is independent of v_0 .
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Classical Mixing time

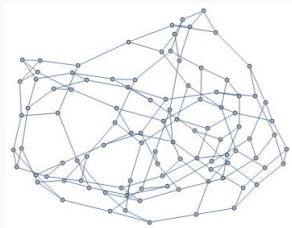


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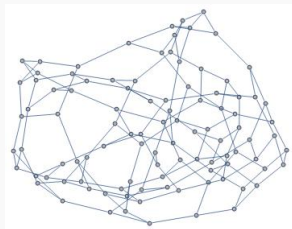
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If P has eigenvalues: $\lambda_n = 1 > \lambda_{n-1} \geq \dots \lambda_1 \geq -1$,

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where $\Delta = 1 - \lambda_{n-1}$.

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- ▶ Depends only on the spectral gap
- ▶ Independent of the initial distribution of the walker.

All the Quantum Mechanics you'll need for this talk:

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In fact, c_j denotes the "*probability amplitude*": probability of $|\psi_0\rangle$ to be in the state $|j\rangle$ is given by the absolute square of the inner-product (or overlap) between the states $|j\rangle$ and $|\psi_0\rangle$:

$$p_j = |\langle j|\psi_0\rangle|^2 = |c_j|^2. \quad [|c_j| \leq 1]$$

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- (iii) Following a *projective* measurement of $|\psi_t\rangle$ in the basis spanned by the states $|k\rangle$, defined by the set of measurement operators $\{M_k = |k\rangle \langle k|\}$, the probability of the system to be in state $|j\rangle$ is given by
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$$|\langle j|M_j|\psi_t\rangle|^2 = |\langle j|\psi_t\rangle|^2.$$

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 - ▶ Time-averaged mixing of a quantum walk: Sample from the “limiting distribution” of the quantum walk.
 - ▶ This will be our focus!
 - ▶ We will consider continuous-time quantum walks, but results are valid for its discrete-time counterpart.
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- ▶ We shall consider the normalized adjacency matrix as the quantum walk Hamiltonian: $\bar{A}_G = \gamma A_G$, where $\gamma = 1/\|A_G\|$.
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Average mixing [AAKV 2001]: Pick a time $t \in [0, T]$ uniformly at random, evolve $|\psi_0\rangle$ for this time under \bar{A}_G and measure in the node basis.

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Upper bound on the quantum mixing time:

$$T_{\text{mix}}^G = \mathcal{O}\left(\frac{1}{\epsilon} \sum_{i=1}^{n-1} \sum_{r=1}^{n-i} \frac{|\langle v_i | \psi_0 \rangle| \cdot |\langle \psi_0 | v_{i+r} \rangle|}{|\lambda_{i+r} - \lambda_i|}\right).$$

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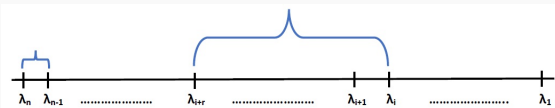
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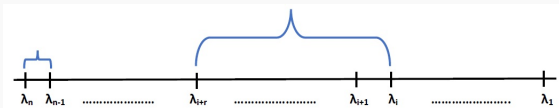
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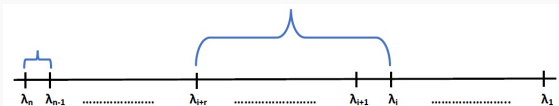
We need to bound

$$\Sigma = \sum_{i=1}^{n-1} \frac{1}{|\lambda_{i+1} - \lambda_i|} + \sum_{i=1}^{n-2} \frac{1}{|\lambda_{i+2} - \lambda_i|} + \dots$$

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If Δ_{\min} is the minimum of all eigenvalue gaps,

$$\frac{1}{\Delta_{\min}} \leq \Sigma \leq \tilde{O} \left(\frac{n}{\Delta_{\min}} \right).$$

So $T_{\text{mix}}^G = \tilde{O}(n/\Delta_{\min})$ is an upper bound for all graphs.

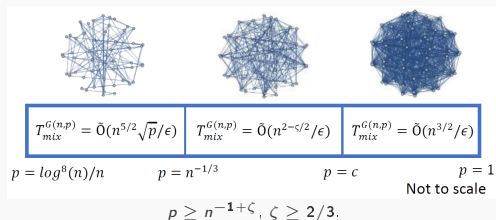
For cycles, the mixing time is in $\tilde{O}(n)$. [AAKV 2001].

Our contributions

Can we obtain a better upper bound for the mixing time for general classes of graphs?

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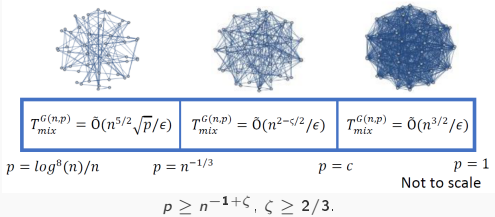
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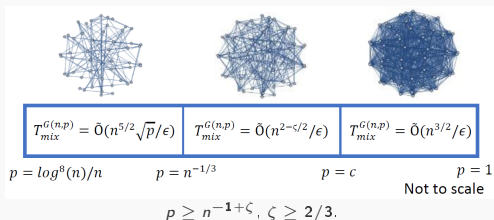
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- ▶ More generally, we obtain a better upper bound on the quantum mixing time for sparse and dense Erdős-Renyi random graphs.

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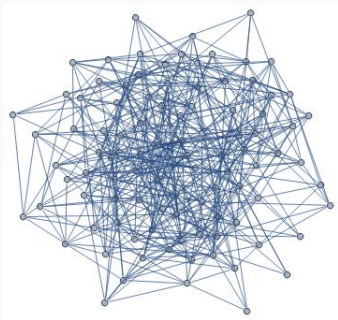
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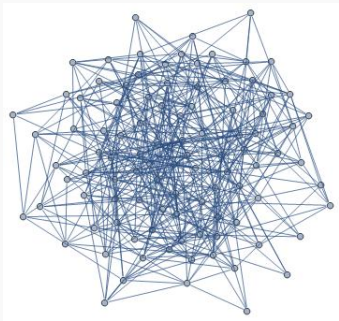
- ▶ More generally, we obtain a better upper bound on the quantum mixing time for sparse and dense Erdős-Renyi random graphs. Bound holds for *almost all* graphs!

Erdős-Renyi random graph (Erdős-Renyi 1959)



A graph where each edge exists with probability p independently of the other edges. Denoted as $G(n, p)$.

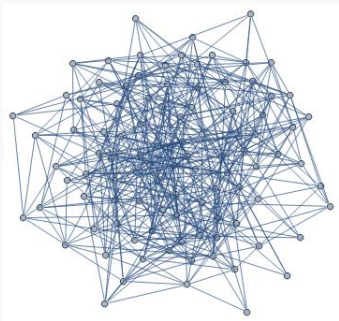
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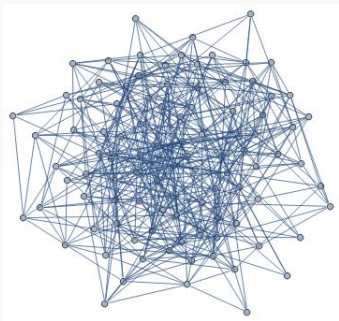


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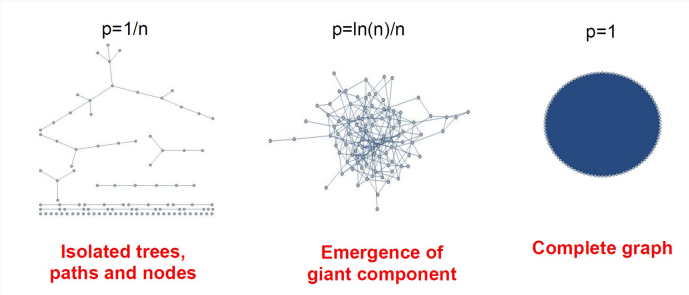


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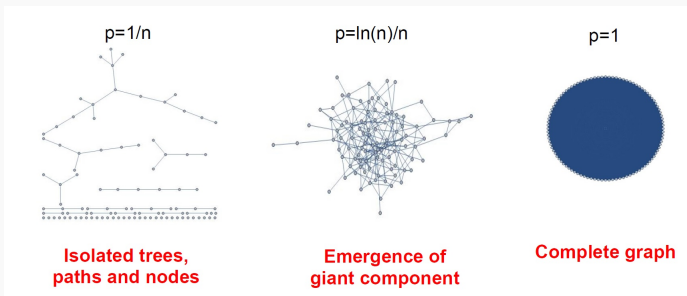
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- ▶ **Almost all** graphs satisfy property P , if $G(n, 1/2)$ satisfies P with probability $1 - o(1)$.
- ▶ Fraction of graphs satisfying P goes to 1 as $n \rightarrow \infty$.

Evolution of Erdős-Renyi random graphs



Evolution of Erdős-Renyi random graphs



$A_{G(n,p)}$ is a symmetric random matrix: Each (non-diagonal) entry is 1 with probability p and 0 with probability $1 - p$.

Upper bound on the quantum mixing time of $G(n, p)$:

We prove that for $p \geq n^{-1/3}$,

$$T_{\text{mix}}^{G(n,p)} = \tilde{O}\left(n^{3/2 - \log(p)/\log(n)} \sqrt{p}\right),$$

almost surely.

Spectral properties of $A_{G(n,p)}$

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Wigner's semicircle distribution

- ▶ As long as $np \rightarrow \infty$, the bulk of the spectrum of $A_{G(n,p)}$ follows the *Wigner's semicircle distribution*:

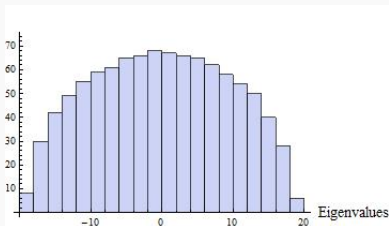
$$\rho_{sc}(x) = \begin{cases} \frac{\sqrt{4np(1-p) - x^2}}{2\pi np(1-p)} & \text{if } |x| < 2\sqrt{np(1-p)} \\ 0 & \text{otherwise} \end{cases} .$$

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- ▶ Radius of the semicircle: $R = 2\sqrt{np(1-p)}$. For $\bar{A}_{G(n,p)}$: $R = 2\sqrt{\frac{1-p}{np}}$.
- ▶ Divide $[-R, R]$ into small bins and count the eigenvalues of $A_{G(n,p)}$ in each bin:

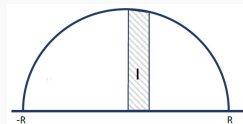


This distribution will converge to a semicircle.

Wigner's semicircle distribution

For any interval $\mathcal{I} \in \mathbb{R}$ if $\mathcal{N}_{\mathcal{I}}$ is the number of eigenvalues of $A_{G(n,p)}$ in \mathcal{I} , then

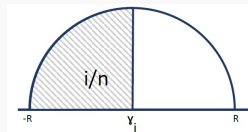
$$\frac{\mathcal{N}_{\mathcal{I}}}{n} = \int_{\mathcal{I}} \rho_{sc}(x) dx + o(1).$$



Classical eigenvalue location:

For any index $1 \leq i \leq n-1$, the *classical location* of each eigenvalue λ_i , denoted by γ_i , is given by

$$\int_{-R}^{\gamma_i} \rho_{sc}(x) dx = \frac{i}{n}.$$



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- ▶ Works only when $|\mathcal{I}| \gg 1$ and not when say $|\mathcal{I}| \sim 1/n$.
- ▶ Useless when we need bounds on consecutive gaps, i.e. $\lambda_{i+1} - \lambda_i$.

What will we need?

Average eigenvalue gap:

$$\bar{\Delta}_{G(n,p)} = \Theta\left(\frac{1}{n^{3/2}\sqrt{p}}\right)$$

Distance between classical eigenvalue locations: For $i \leq n/2$, $r \leq n - 2i$ and some universal constant $c > 0$

$$\gamma_{i+r} - \gamma_i \geq c \frac{r}{n^{7/6} i^{1/3} \sqrt{p}}.$$

Eigenvalue rigidity criterion

Eigenvalues of $\bar{A}_{G(n,p)}$ are concentrated around their classical locations. We present a simplified version (adapted for our analysis):

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Eigenvalue rigidity criterion [EYY2011, EKYY2013]:

For $1 \leq i \leq n-1$, any $\varepsilon \geq 0$ and $p \geq n^{-1/3}$, the eigenvalues of $\bar{A}_{G(n,p)}$ satisfy the inequalities

$$|\lambda_i - \gamma_i| \leq \frac{n^\varepsilon (n^{-2/3} \alpha_i^{-1/3} + n^{-\phi})}{(pn)^{1/2}}$$

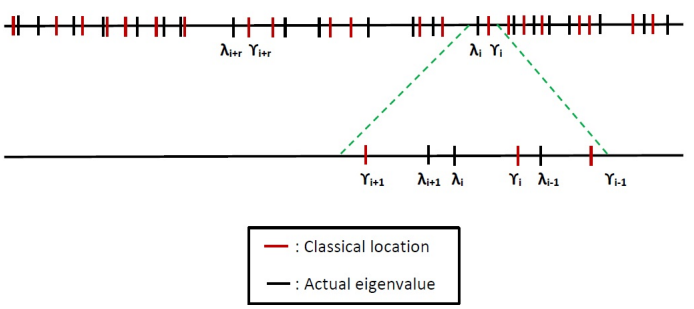
with probability $1 - o(1)$, where

$$\phi := \frac{\log pn}{\log n} \quad \text{and} \quad \alpha_i := \max\{i, n-i\}.$$

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- ▶ Not only this, Nguyen, Tao and Vu [NTV2015] showed that each δ_i is separated.
- ▶ They obtain tail bounds on δ_i : How likely is each δ_i to be δ times less than the average?

Tail-bounds on eigenvalue gaps of $\bar{A}_{G(n,p)}$ [NTV2015, LL2019]:

For $p \geq \log^6(n)/n$ and $1 \leq i \leq n-1$

$$\sup_{1 \leq i \leq n-1} \mathbb{P} \left(\delta_i \leq \delta \frac{1}{n^{3/2} \sqrt{p}} \right) \leq C \delta \log n,$$

with probability $1 - o(1)$, where $\delta \geq n^{-C}$

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Applying union bound gives a lower bound on Δ_{\min} .

Minimum eigenvalue gap $\bar{A}_{G(n,p)}$:

$$\Delta_{\min} \geq \frac{1}{n^{5/2+o(1)} \sqrt{p}},$$

with probability $1 - o(1)$.

Upper bounding Σ_1

Recall:

$$\Sigma = \sum_{i=1}^{n-1} \sum_{r=1}^{n-i} \frac{1}{|\lambda_{i+r} - \lambda_i|},$$

and $1/\Delta_{\min} \leq \Sigma \leq \tilde{O}(n/\Delta_{\min})$.

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First we prove an upper bound on Σ_1 using the tail bounds on δ_i :

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Key idea: Many δ_i 's are close to the average. Count them!

- ▶ Σ_1 is close to $1/\Delta_{\min}$.

Upper bound on Σ

To now obtain an upper bound on Σ , we combine two things:

- (i) Distance between classical eigenvalue locations:

$$\gamma_{i+r} - \gamma_i \geq c \frac{r}{n^{7/6} i^{1/3} \sqrt{p}}.$$

- (ii) Eigenvalue rigidity - Eigenvalues are close to the classical locations

$$|\lambda_i - \gamma_i| \leq \frac{n^\epsilon (n^{-2/3} \alpha_i^{-1/3} + n^{-\phi})}{(pn)^{1/2}}$$

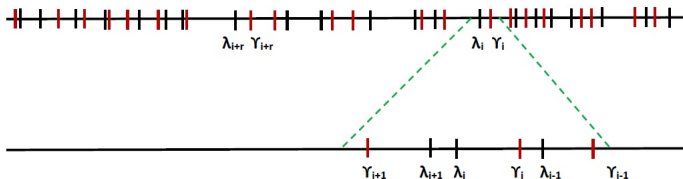
Idea:

For small values of r use tailbounds of δ_i while for larger values of r make use of eigenvalue rigidity!

Upper bound on Σ

Idea:

- ▶ For small values of r use tailbounds of δ_i while for larger values of r make use of eigenvalue rigidity!
- ▶ Exploit eigenvalue rigidity for large enough r such that $(\gamma_{i+r} - \gamma_i)$ is larger than the error due to $|\lambda_{i+r} - \gamma_{i+r}| + |\lambda_i - \gamma_i|$.
- ▶ Critical value: $r_*(i) \leq n^{\varepsilon - \log p / \log n}$.



— : Classical location
— : Actual eigenvalue

Upper bound on Σ

Upper bound for Σ

For $p \geq n^{-1/3}$,

$$\Sigma = \sum_{i=1}^{n-1} \sum_{r=1}^{n-i} \frac{1}{|\lambda_{i+r} - \lambda_i|} \leq \mathcal{O} \left(n^{5/2 - \frac{\log p}{\log n} + o(1)} \sqrt{p} \right),$$

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Proof idea: Split Σ into two parts.

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Σ is close to $1/\Delta_{\min}$ for dense $G(n, p)$.

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Mixing time of random graphs

Recall:

Limiting distribution:

$$P_f(T \rightarrow \infty) = \lim_{T \rightarrow \infty} P_f(T) = \sum_{i=1}^n |\langle f | v_i \rangle \langle v_i | \psi_0 \rangle|^2.$$

- ▶ Our results are independent of $|\psi_0\rangle$.
- ▶ All eigenstates of $\bar{A}_{G(n,p)}$ are delocalized, i.e. $\|v_i\|_\infty \leq n^{-1/2+o(1)}$, with probability $1 - o(1/n)$ [EKYY 2013, HKM 2018].

Limiting distribution for $G(n, p)$:

Immediately we obtain:

$$P_f(T \rightarrow \infty) \leq \tilde{O}(1/n),$$

almost surely irrespective of $|\psi_0\rangle$. (Close to uniform!)

Mixing time of random graphs

Limiting distribution:

$$P_f(T \rightarrow \infty) \leq \tilde{O}(1/n).$$

Upper bound on the quantum mixing time:

$$T_{\text{mix}}^G = \mathcal{O}\left(\frac{1}{\epsilon} \sum_{i=1}^{n-1} \sum_{r=1}^{n-i} \frac{|\langle v_i | \psi_0 \rangle| \cdot |\langle \psi_0 | v_{i+r} \rangle|}{|\lambda_{i+r} - \lambda_i|}\right).$$

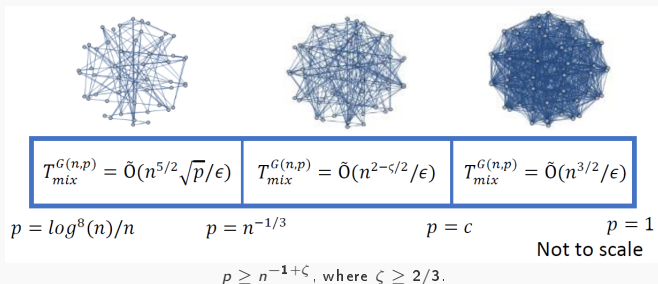
Quantum mixing time for $G(n, p)$

From the upper bound on Σ_1 and from the delocalization of the eigenstates we have

$$T_{\text{mix}}^{G(n,p)} = \tilde{O}\left(n^{3/2 - \log(p)/\log(n)} \sqrt{p/\epsilon}\right),$$

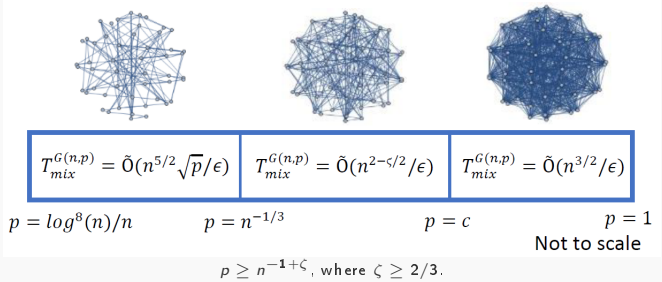
almost surely for any $p \geq n^{-1/3}$.

Summary of results



- Improved upper bound on the quantum mixing time for almost all graphs.

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- ▶ Improved upper bound on the quantum mixing time for almost all graphs.
- ▶ For sparser graphs, i.e. when $p = \log^D(n)/n$, $D \geq 8$, eigenvalue rigidity breaks down.
- ▶ In that case we have a weaker upper bound of

$$T_{mix}^{G(n,p)} = \tilde{O}(n^{5/2}\sqrt{p}/\epsilon).$$

Ongoing and Future work

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[C and Y. Atia, *arXiv:2005.04062 (2020)*].
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- ▶ Continuous-time fast forwarding of Markov chains: For any Markov chain P , initial state v , there exists a continuous-time quantum procedure that outputs $|e^{Pt}v\rangle = \sum_j (e^{Pt}v)_j |j\rangle / \|e^{Pt}v\|$ in time $\tilde{O}\left(\sqrt{t} \|e^{Pt}v\|^{-1}\right)$.
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- ▶ Discrete-time quantum walks W using continuous-time quantum walks H : Express

$$W^K \approx \sum_{t=-\mathcal{O}(K)}^{+\mathcal{O}(K)} c_t e^{-iHt}.$$

Thank you for your attention!
