

The Planted Spanning Tree Problem

Jiaming Xu

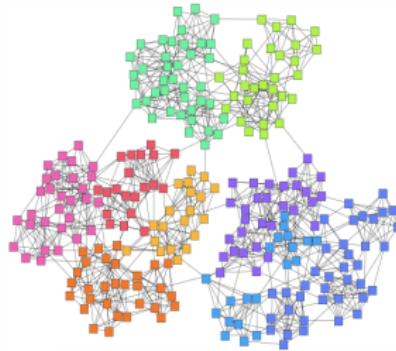
The Fuqua School of Business
Duke University

Joint work with
Mehrdad Moharrami (UIowa) and Cristopher Moore (Santa Fe Institute)

June 30, 2025
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Network model with planted structure

Question: How to recover latent structure from noisy network data?



Classical examples

- Community detection under Stochastic Block Model
- Recovery of planted clique in Erdős-Rényi graphs

An expanding zoo of planted subgraph problems...

- Planted bipartite matching [Chertkov-Kroc-Krzakala-Vergassola-Zdeborová '10, Moharrami-Moore-X. '21, Ding-Wu-X.-Yang '23]
- Planted Hamiltonian cycle problem (TSP) [Bagaria-Ding-Tse-Wu-X. '20]
- Planted trees [Massoulié-Stephan-Towsley '18]
- Planted k -factors [Sicuro-Zdeborová '20, Gaudio-Sandon-X.-Yang '25]
- Planted k -nearest-neighbor graph [Ding-Wu-X.-Yang '21]
- Planted dense cycles [Mao-Wein-Zhang '23]
- Planted general subgraphs [Mossel-Niles-Weed-Sohn-Sun-Zadik '23, Lee-Pernice-Rajaraman-Zadik '25]

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Many fascinating results on phase transitions, statistical vs. computational gaps, optimal algorithms; however, characterizing exact value of **asymptotic overlap** remains formidable mathematical challenge

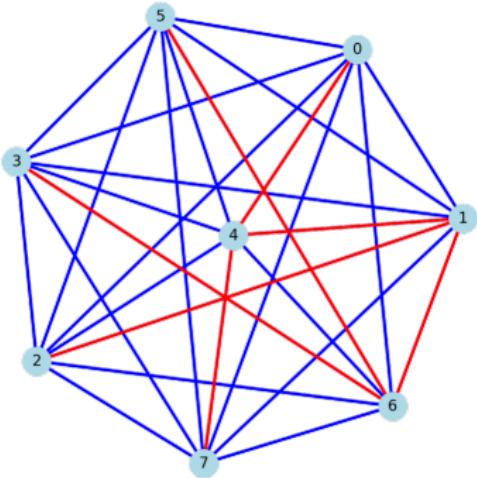
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Today: **Planted spanning tree model** → Exact overlap characterization via local weak convergence theory

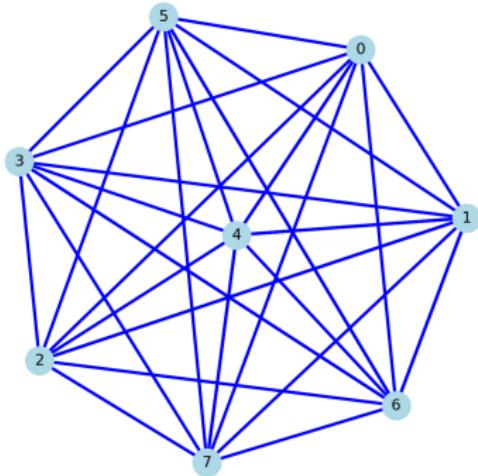
The planted spanning tree model



- A complete graph on n vertices
- A uniform spanning tree T^*
- Non-negative edge weight

$$W_e \stackrel{\text{ind.}}{\sim} \begin{cases} P & e \in T^* \\ Q_n & e \notin T^* \end{cases}$$

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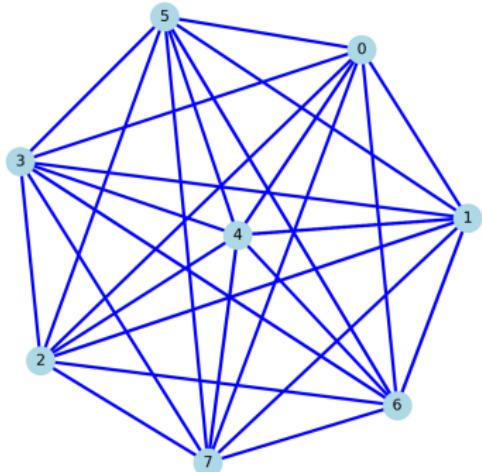


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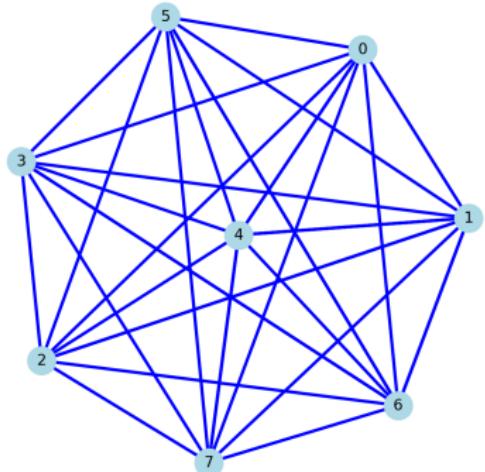
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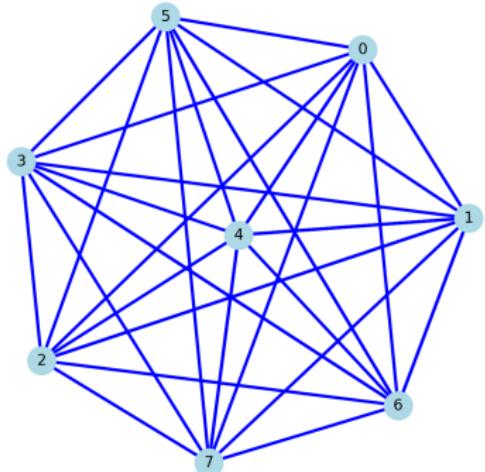
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- When $P = \text{Exp}(1/\mu)$ and $Q_n = \text{Exp}(1/n)$ (mean μ vs n),
min-weight spanning tree T_{\min} is Maximum Likelihood Estimator

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- When $P = \text{Exp}(1/\mu)$ and $Q_n = \text{Exp}(1/n)$ (mean μ vs n), min-weight spanning tree T_{\min} is Maximum Likelihood Estimator
- How much does T_{\min} have in common with T^* ?

Main result: Asymptotic overlap

Theorem (Moharrami-Moore-X. '25)

Let F denote the CDF of planted weight distribution P . Then

$$\lim_{n \rightarrow \infty} \frac{1}{n-1} \mathbb{E}[|T_{\min} \cap T^*|] = \int_0^\infty (1 - p_U(s)p_B(s)) dF(s)$$

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where $p_U(s)$ and $p_B(s)$ are the largest fixed point of

$$1 - p_U(s) = \exp(-sp_U(s) - p_B(s))(1 - F(s)p_U(s))$$

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Remark

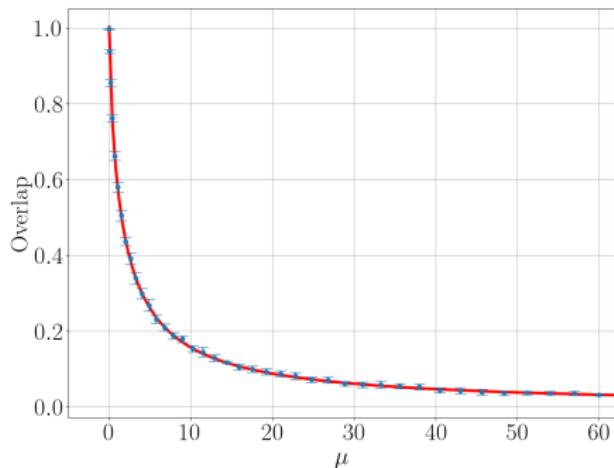
- $p_U(s)$ and $p_B(s)$ are probabilities of certain branching process growing to infinity
- The fixed-point equations have at most two solutions, one of which is zero

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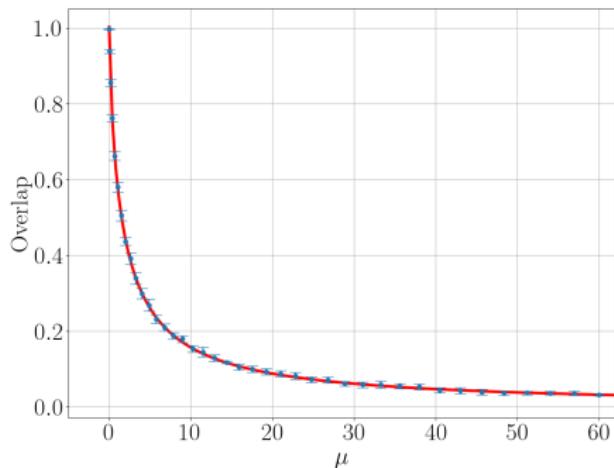
$$\begin{aligned} P &= \exp(1/\mu) \\ Q &= \exp(1/n) \\ \text{mean: } \mu &\text{ vs } n \end{aligned}$$

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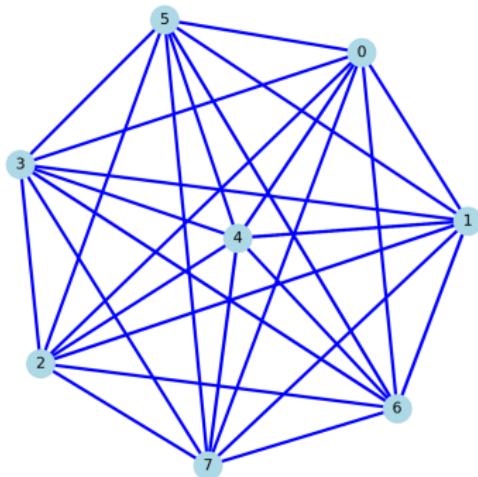


$$P = \exp(1/\mu)$$
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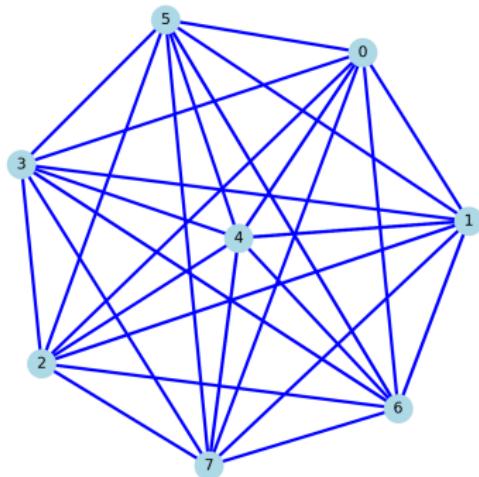
No phase transition!

Warmup: (un-planted) spanning tree problem



- A complete graph on n vertices
- Weights are i.i.d. Q_n with $nQ'_n(0) \rightarrow 1$
- What is the mean weight of the minimum spanning tree,
 $w(T_{\min}) \triangleq \frac{1}{n-1} \sum_{e \in T_{\min}} W_e$?

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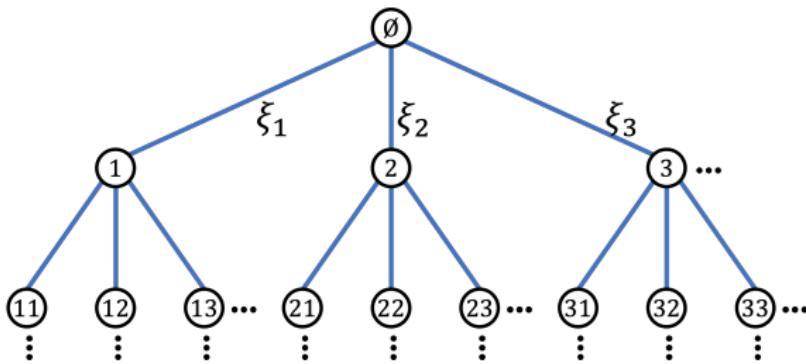
[Frieze '85, Aldous-Steele '04, Addario-Berry'15]

$$\mathbb{E}[w(T_{\min})] \rightarrow \zeta(3) \triangleq \sum_{i=1}^{\infty} i^{-3} \approx 1.202$$

Proved in several distinct and elegant ways

Poisson-weighted infinite tree approximation

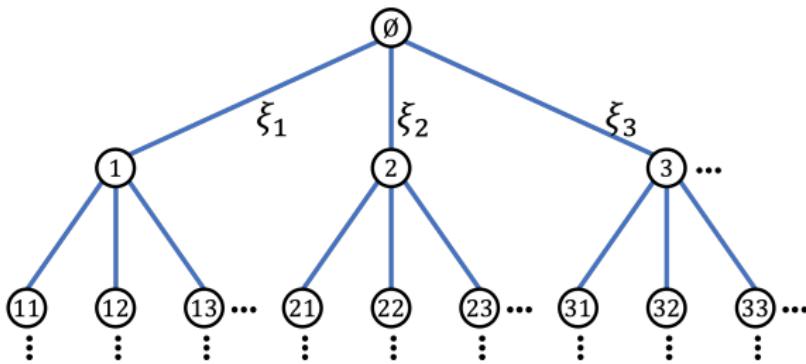
Treat edge weight as distance and model “local” neighborhood as a tree
[Aldous'00, Aldous-Steele '04]



Sort iid Q_n edge weights $W_{\emptyset,1}, W_{\emptyset,2}, \dots$ from smallest to largest:
 $\xrightarrow{n \rightarrow \infty}$ arrival times ξ_1, ξ_2, \dots of a Poisson process with rate 1

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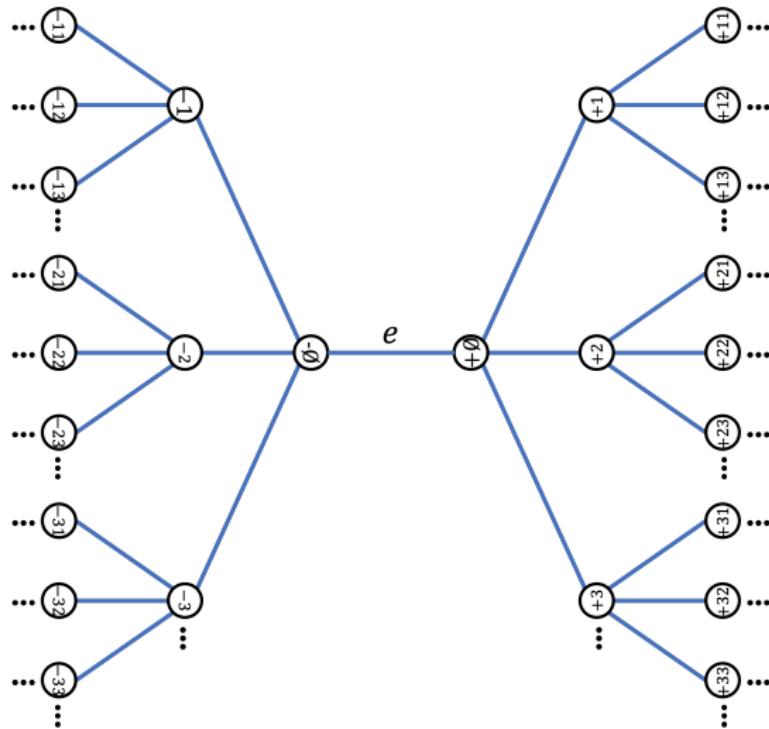


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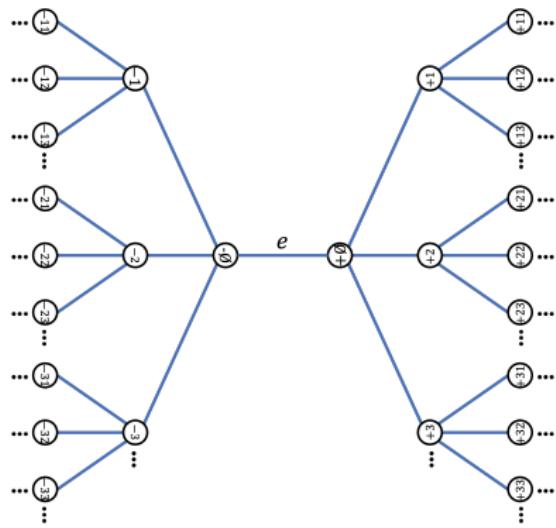
Removing all edges with weights s or greater, we get Galton-Watson tree with $\text{Pois}(s)$ offspring distribution, denoted by $\text{PGW}(s)$

From the perspective of an edge

Pick an edge and model its local neighborhood as a two-sided infinite tree

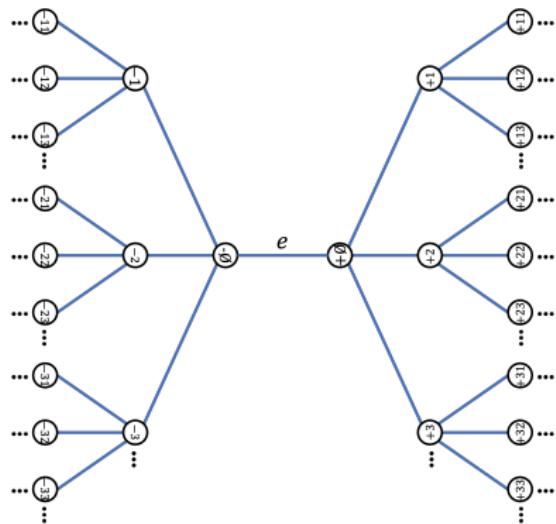


From the perspective of an edge, cont'd



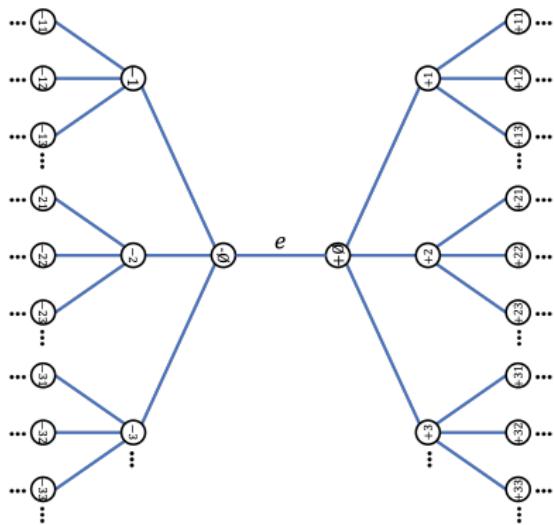
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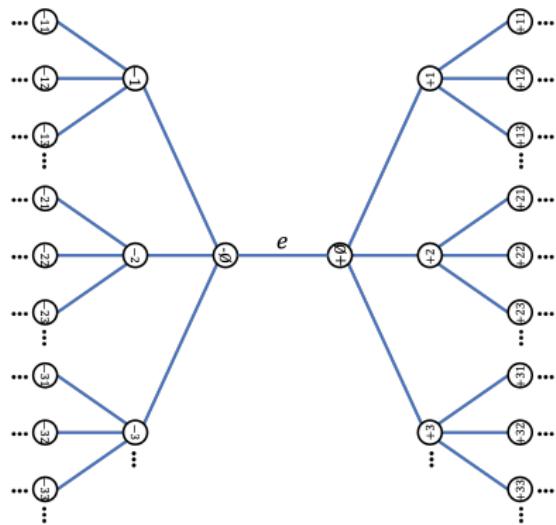
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- e in MST iff removing all edges with weight s or greater, no path exists between its two endpoints
 \Leftrightarrow at least one side is finite

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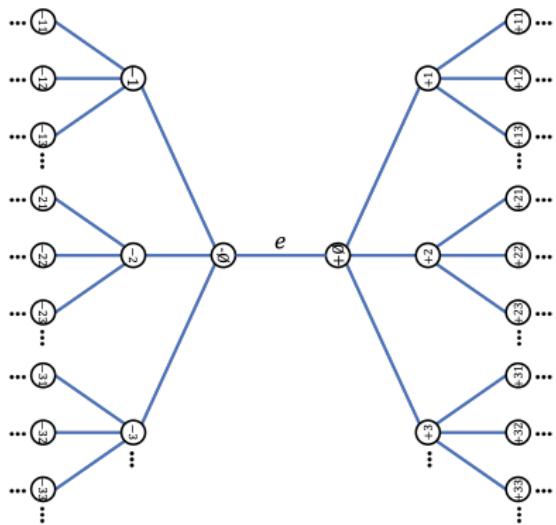


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Let $p(s) = \mathbb{P} [\text{PGW}(s) \text{ is infinite}]$. Then $1 - p(s) = \exp(-sp(s))$ and

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$$\mathbb{P}[e \in T_{\min}] = 1 - p^2(s) \Rightarrow \mathbb{E}[w(T_{\min})] \rightarrow \frac{1}{2} \int_0^{\infty} s(1 - p^2(s))ds = \zeta(3)$$

Back to planted spanning tree

Question: How does the planted uniform spanning tree look like locally?

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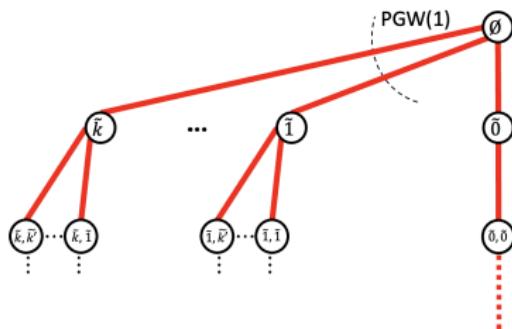


- Start with the infinite path from the root to infinity

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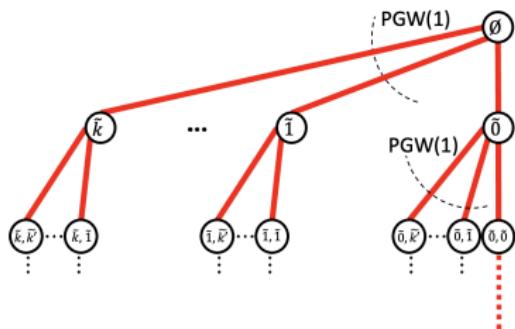
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- Start with the infinite path from the root to infinity
- To each vertex in the infinite path, attach an independent $\text{PGW}(1)$

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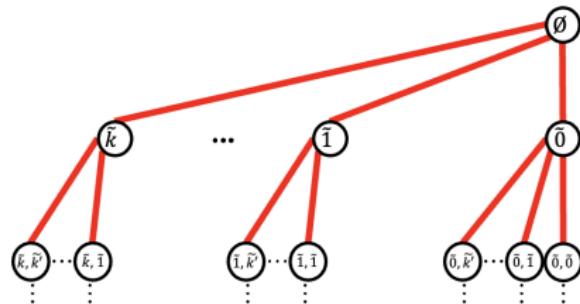
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Putting together: Planted Poisson-weighted infinite tree

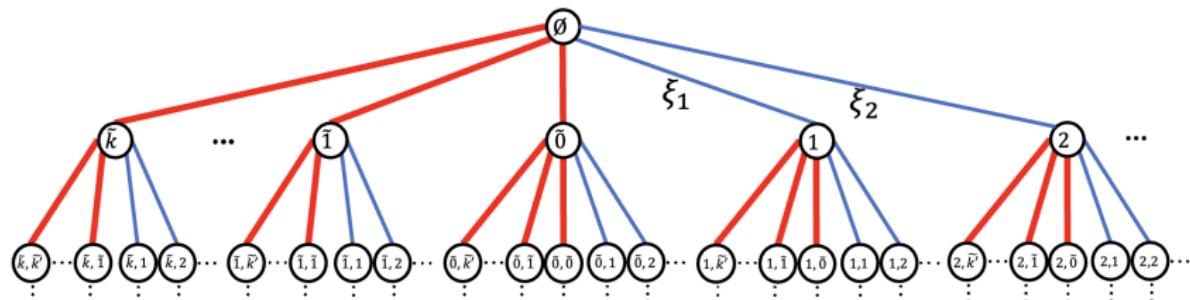
The local view of graph G with planted uniform spanning tree



- The planted (red) edges form **skeleton trees**

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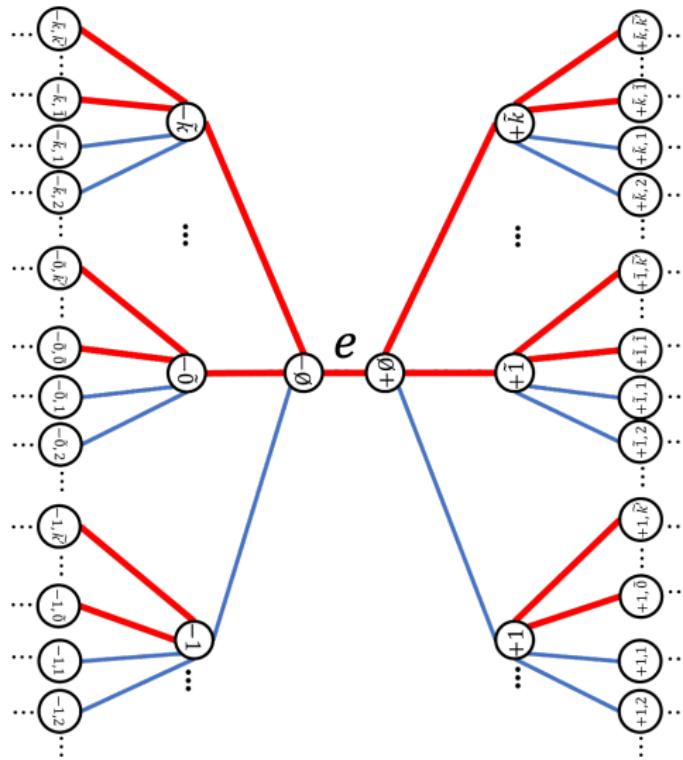
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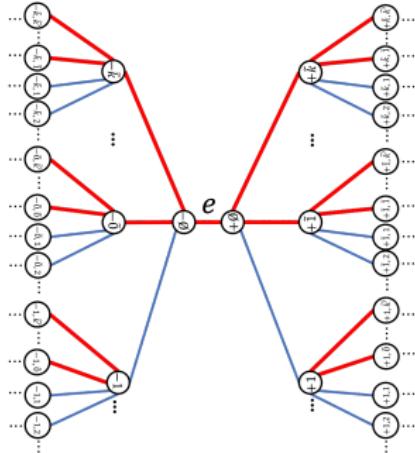
- The planted (red) edges form **skeleton trees**
- The unplanted (blue) edges form **Poisson-weighted infinite trees**

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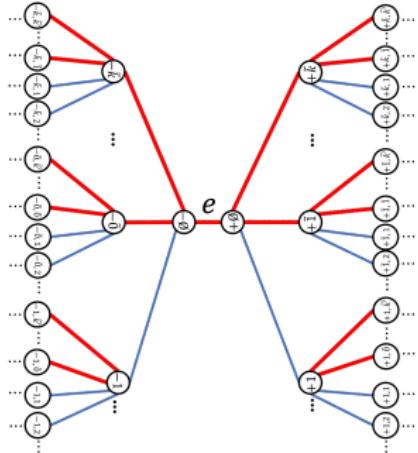


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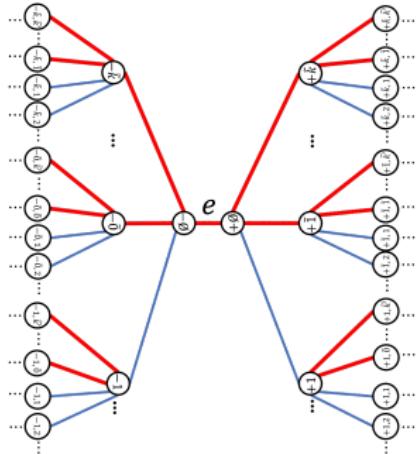
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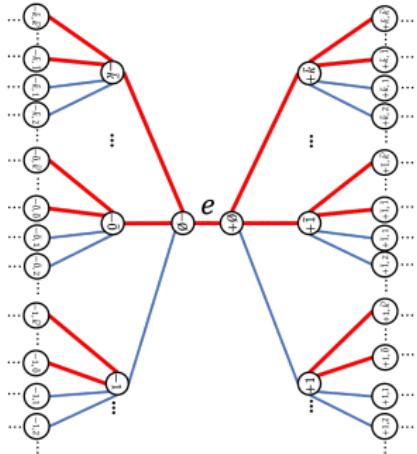
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$$1 - p_U(s) = \underbrace{\exp(-s p_U(s))}_{T_{-i} \text{ 's are finite}} \times \underbrace{\exp(-p_B(s))}_{T_{-\tilde{k}} \text{ 's are finite}} \times \underbrace{(1 - F(s)p_U(s))}_{T_{-\tilde{0}} \text{ is finite}}$$

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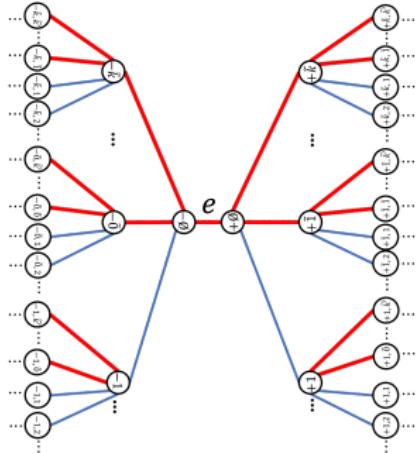


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 1 - p_B(s) &= \exp(-sp_U(s)) \times \exp(-p_B(s)) \\
 \Rightarrow \frac{1}{n-1} \mathbb{E}[|T_{\min} \cap T^*|] &\rightarrow \int_0^\infty (1 - p_U(s)p_B(s)) dF(s)
 \end{aligned}$$

- Planted graph G_n converges to planted Poisson-weighted infinite trees T_∞
 - ▶ Local treelikeness of light edges
 - ▶ Uniform spanning tree converges to skeleton trees [Grimmett '80]
- MST of G_n converges to *minimum spanning forest* of T_∞ [Aldous '91]

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- Exactly characterize overlap between MST and the planted one
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- Characterize the overlap of MST for planted Hamiltonian path/cycle model [\[Bagaria-Ding-Tse-Wu-X. '20\]](#)
- Determine the weight of MST under the planted model, extending Frieze's $\zeta(3)$ result
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Open problems

- ① Determine the information-theoretically optimal overlap
- ② Other planted structures, such as spanning k -regular graphs [Sicuro-Zdeborová '21, Gaudio-Sandon-X.-Yang '25]

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Reference

M. Moharrami, C. Moore, & J. X., *The planted spanning tree problem.*
[arXiv: 2502.08790](#). Conference on Learning Theory 2025