

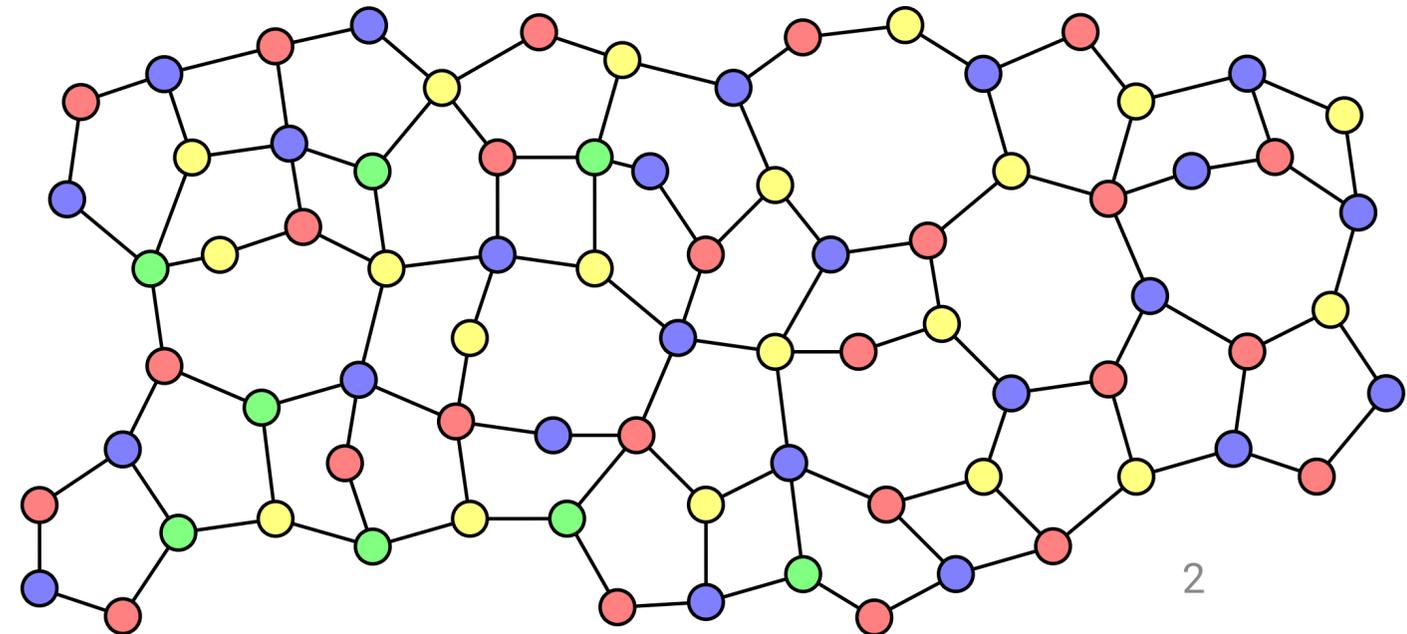
Randomized Coloring & MIS

Dennis Olivetti

University of Freiburg, Germany

Distributed Coloring Problem

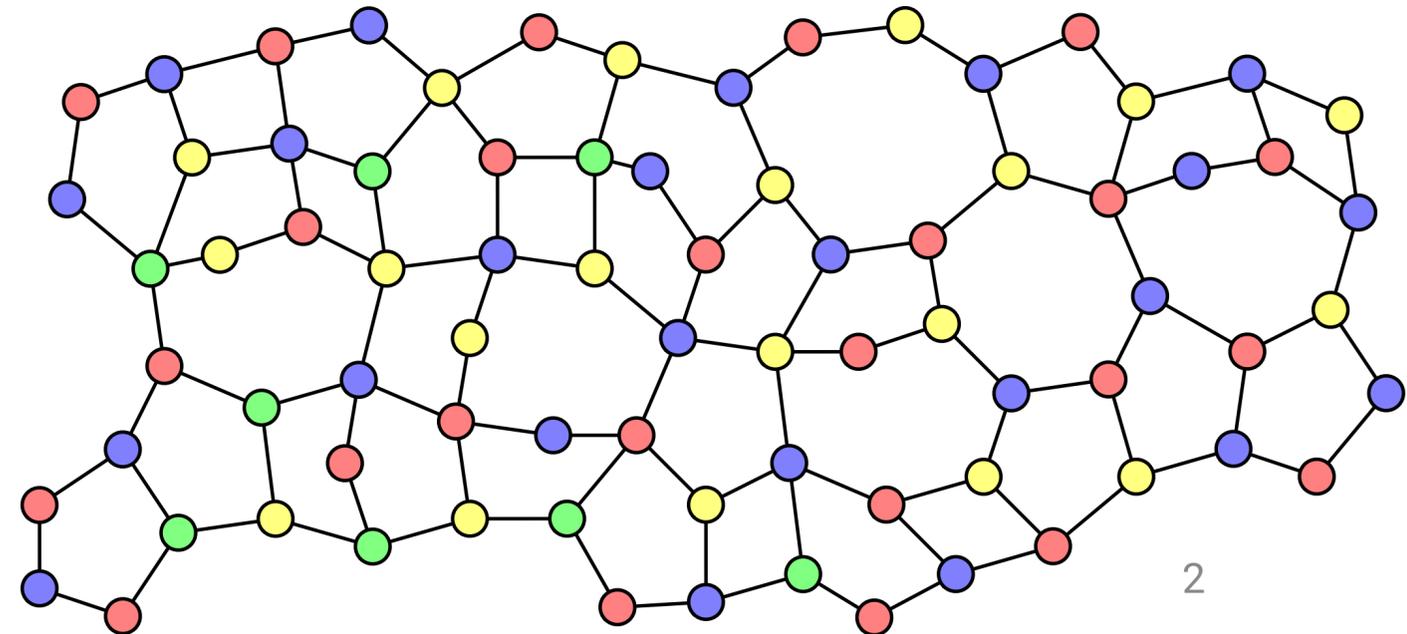
$(\Delta+1)$ -Vertex Coloring



Distributed Coloring Problem

Objective: properly color the nodes with $\leq \Delta + 1$ colors

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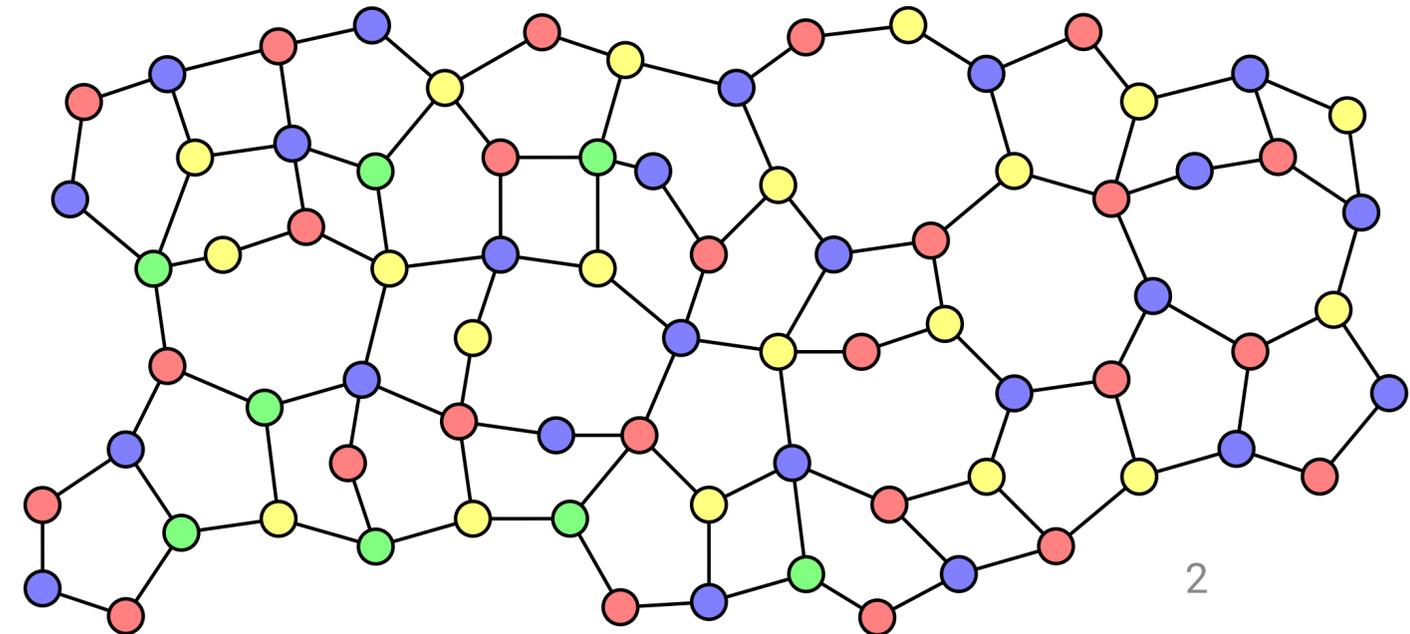


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- ▶ Δ : maximum degree

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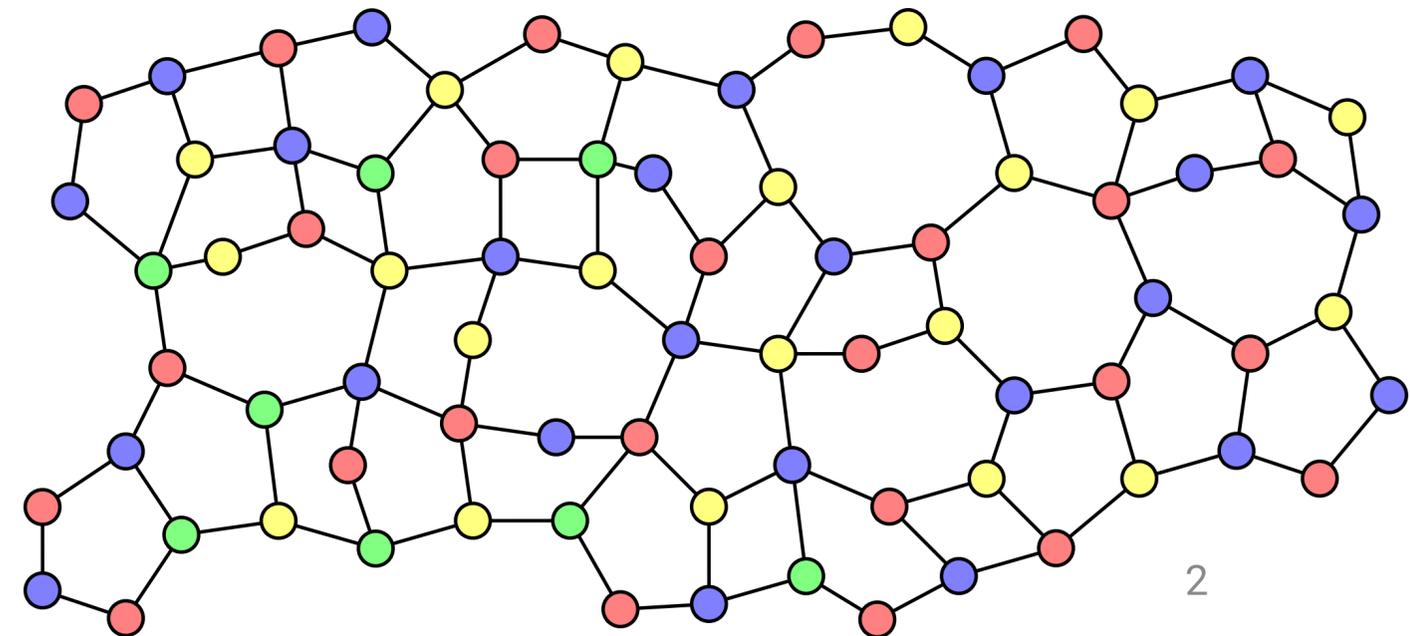


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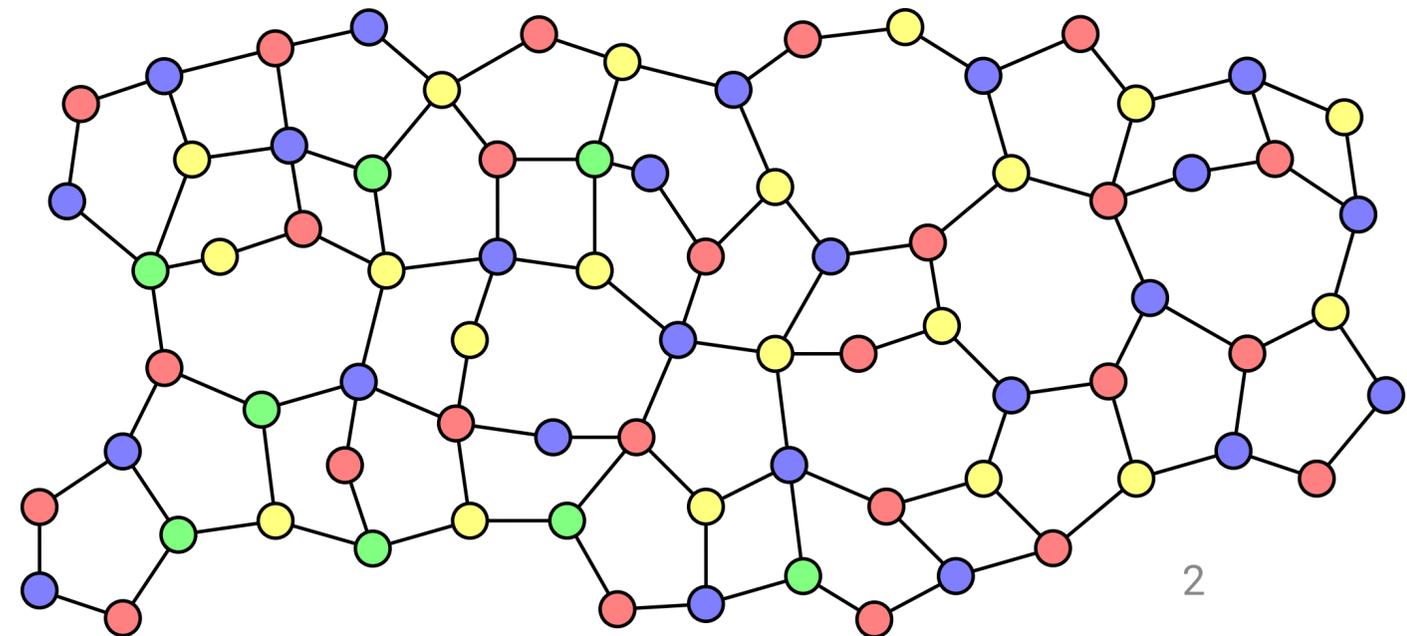


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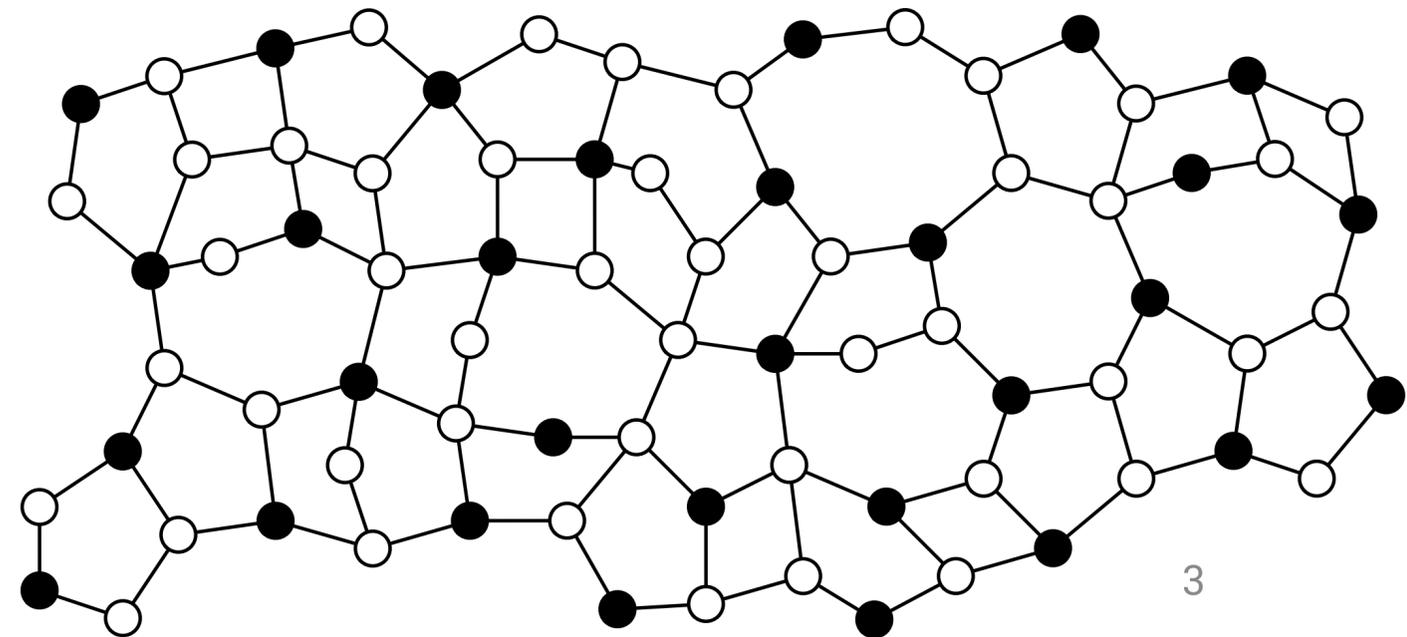
- ▶ Δ : maximum degree
- ▶ $\Delta + 1$: what a simple sequential **greedy** algorithm achieves

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Distributed MIS Problem

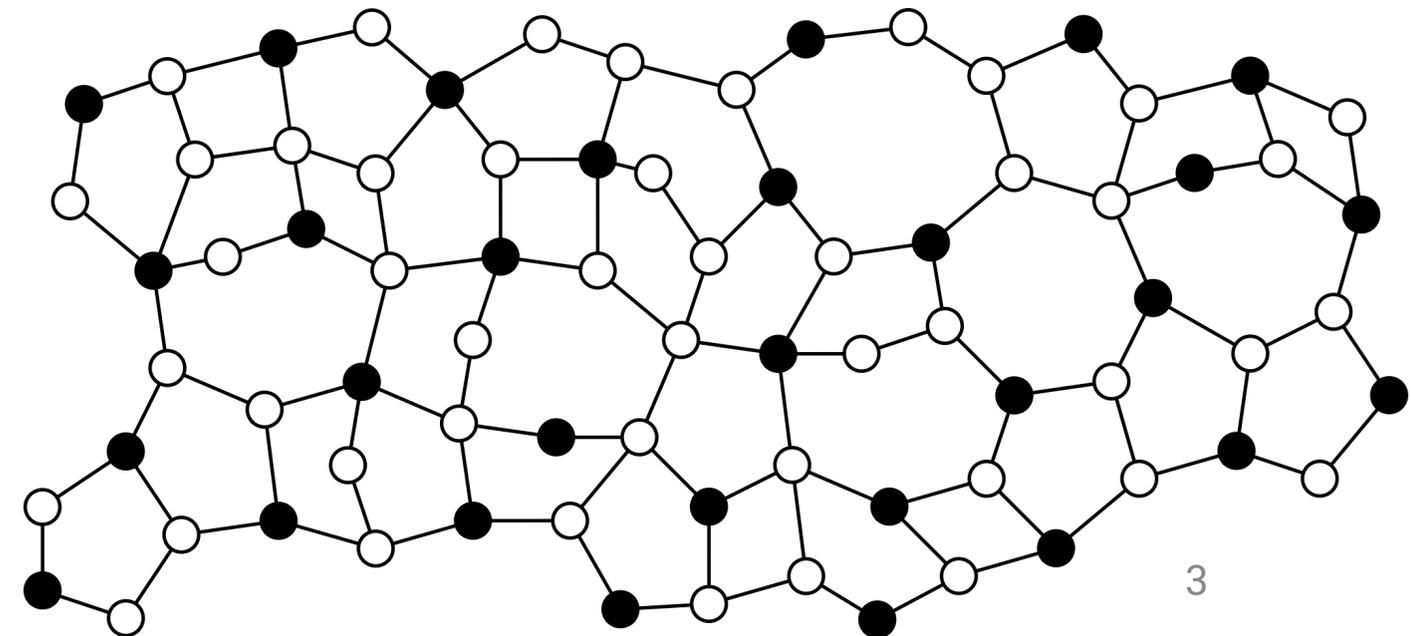
Maximal Independent Set



Distributed MIS Problem

Objective: compute a maximal independent set (MIS)

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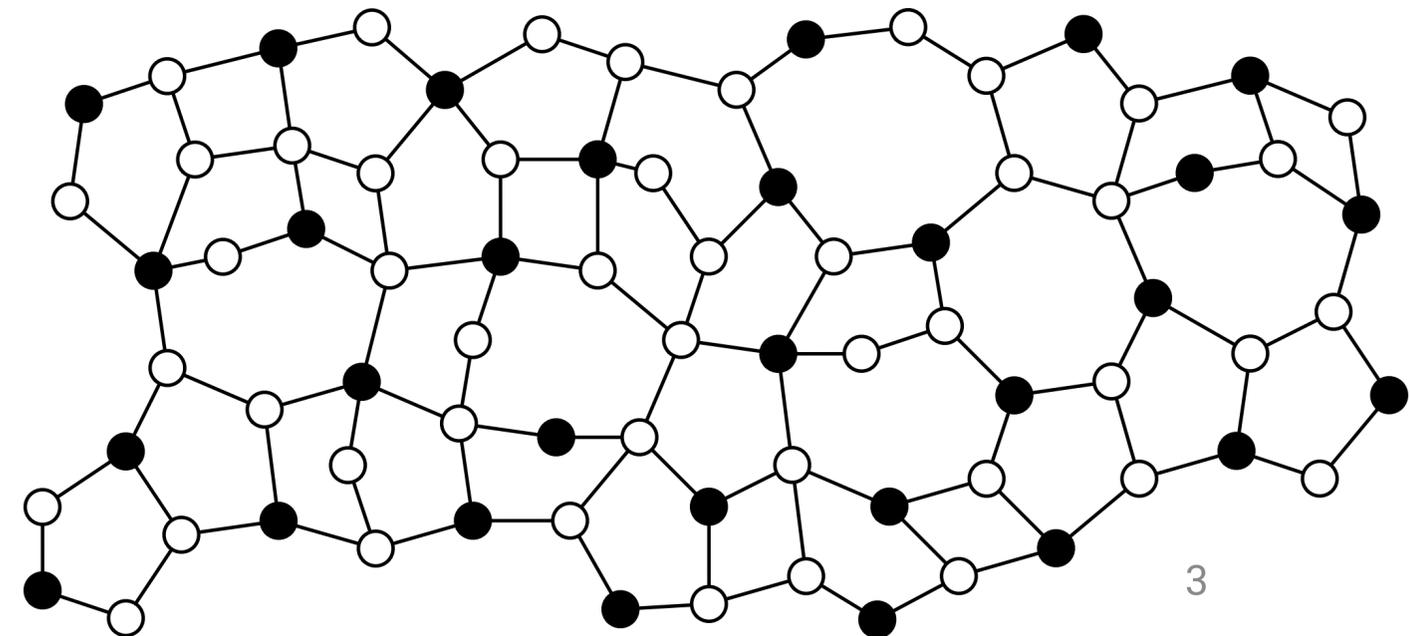


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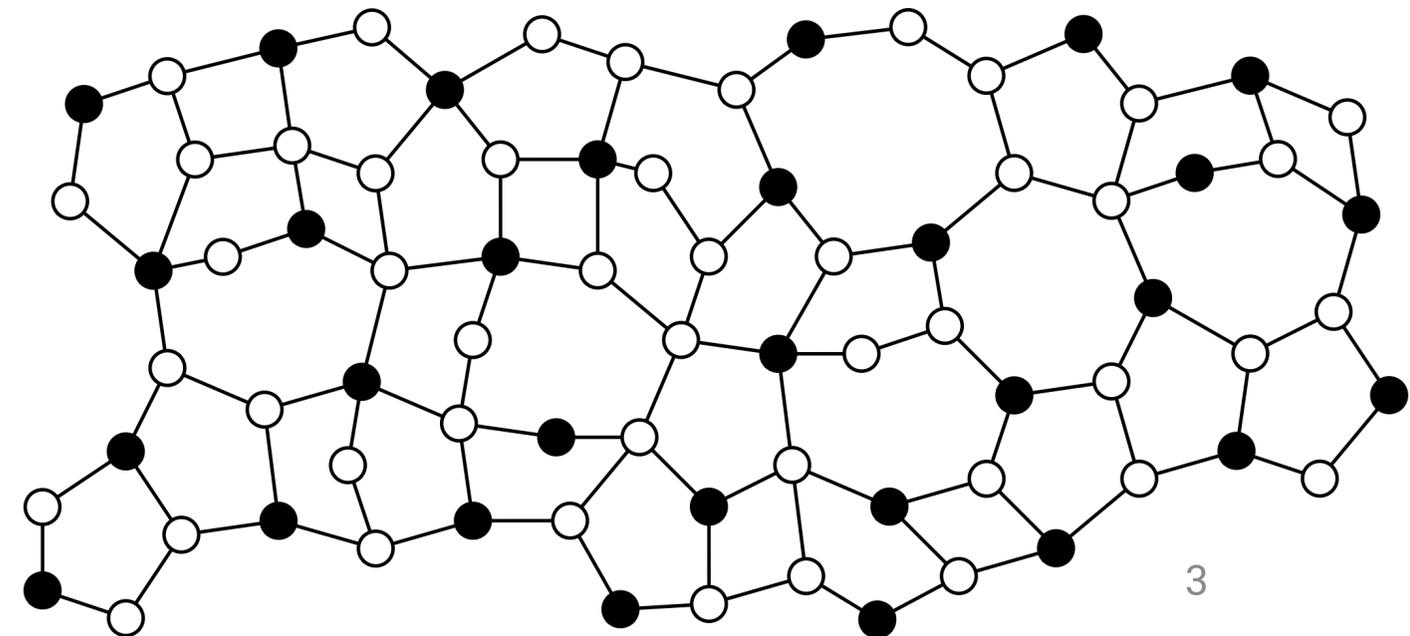


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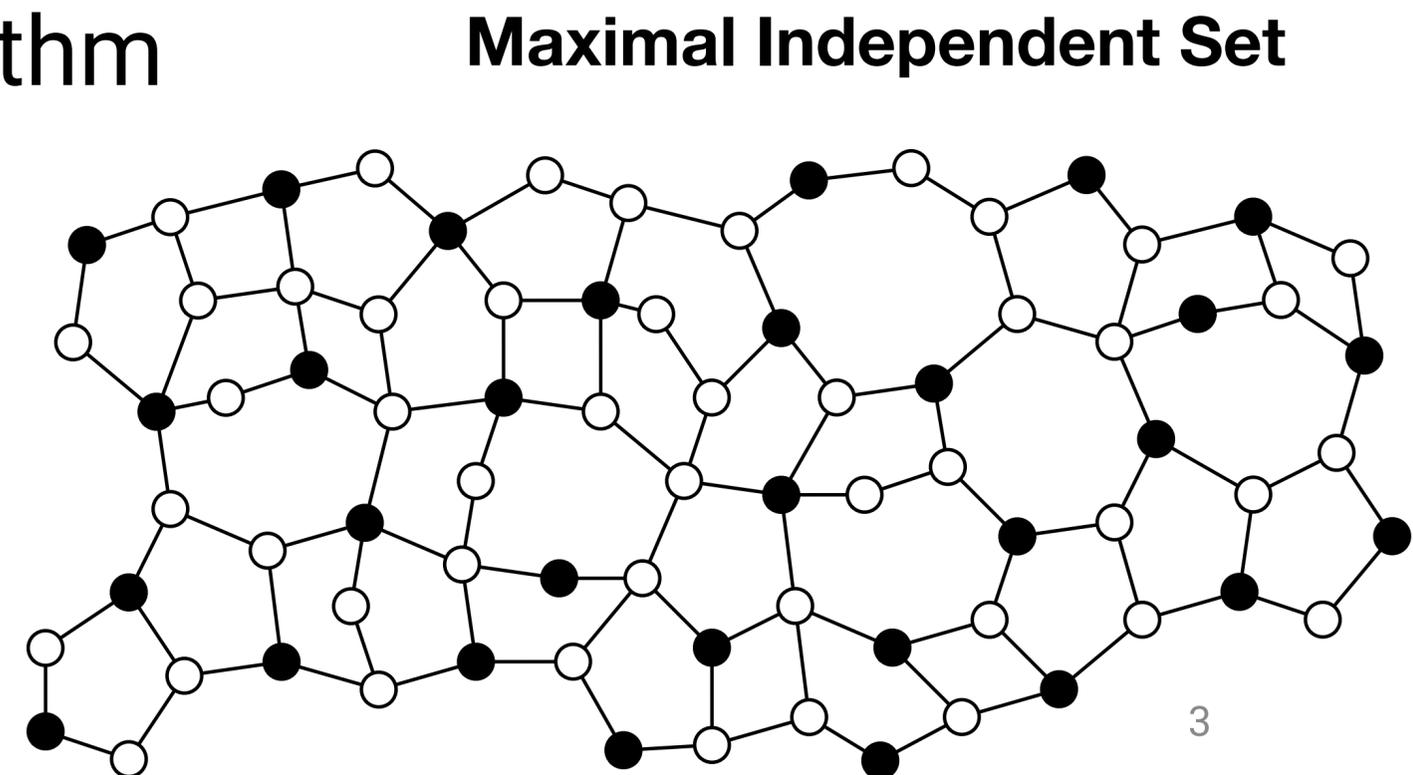
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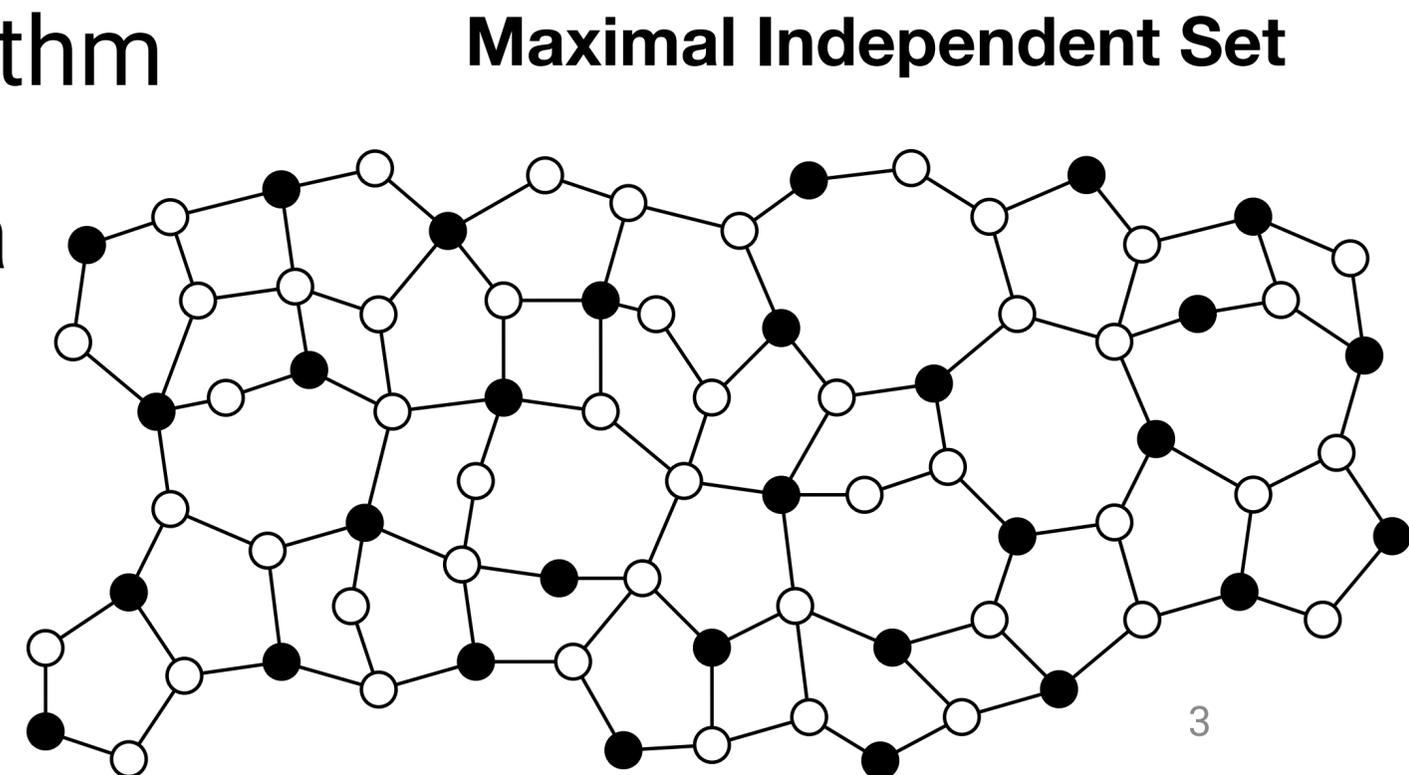
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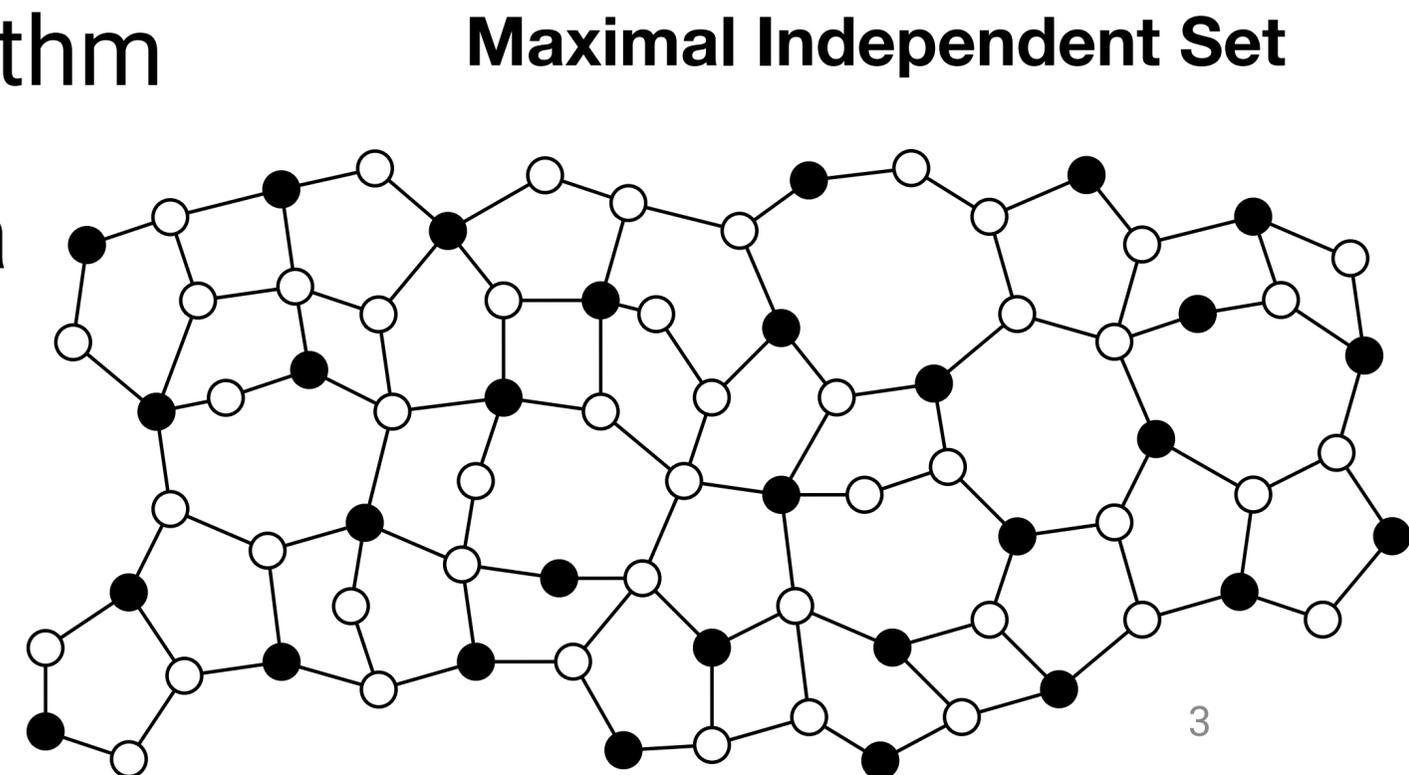
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- ▶ Easily solvable with a **greedy** algorithm
- ▶ The Maximum Independent Set is a different (much harder) problem



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- ▶ **Coloring trees**

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- **Randomized algorithms** for $(\Delta+1)$ -coloring and MIS: $O(\log n)$ time in general graphs!

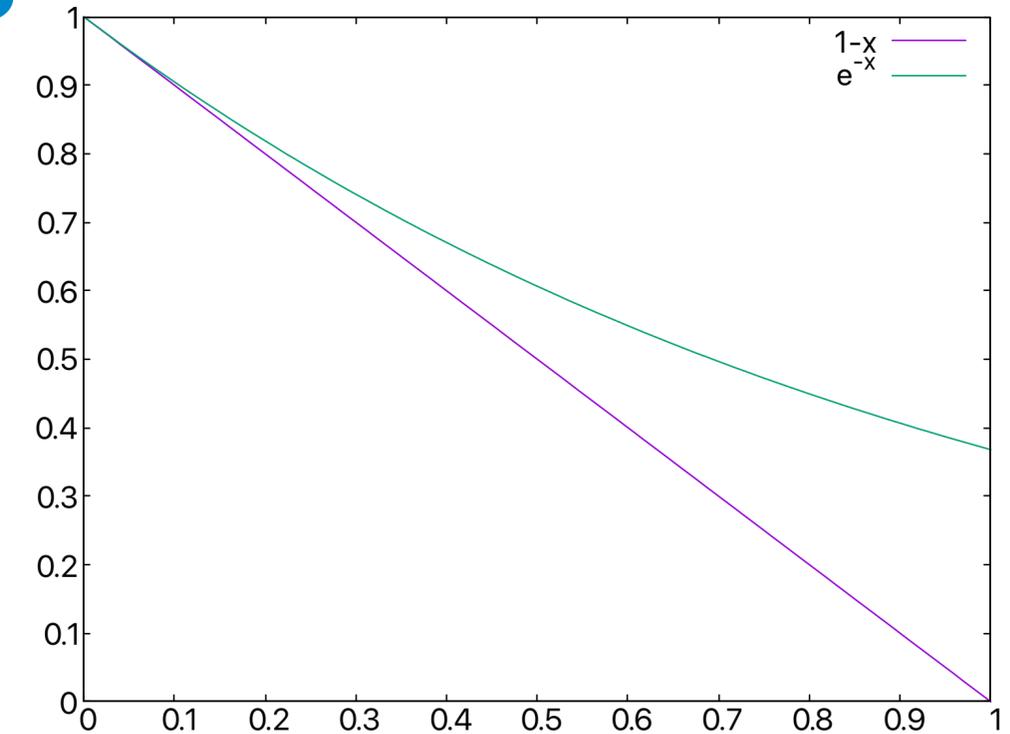
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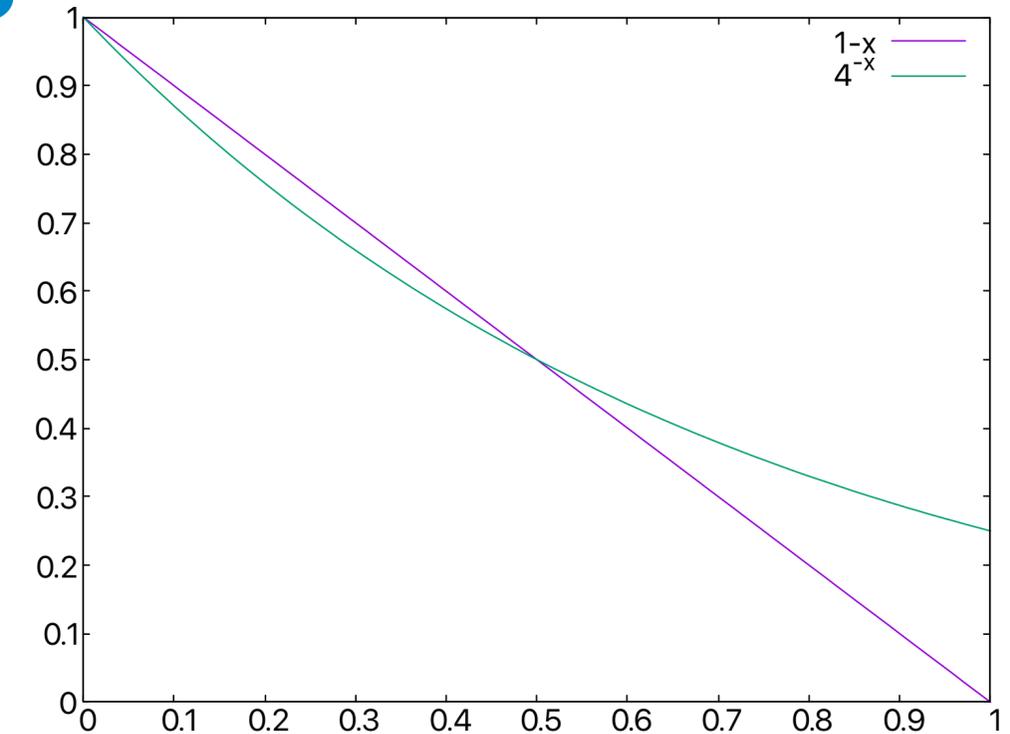
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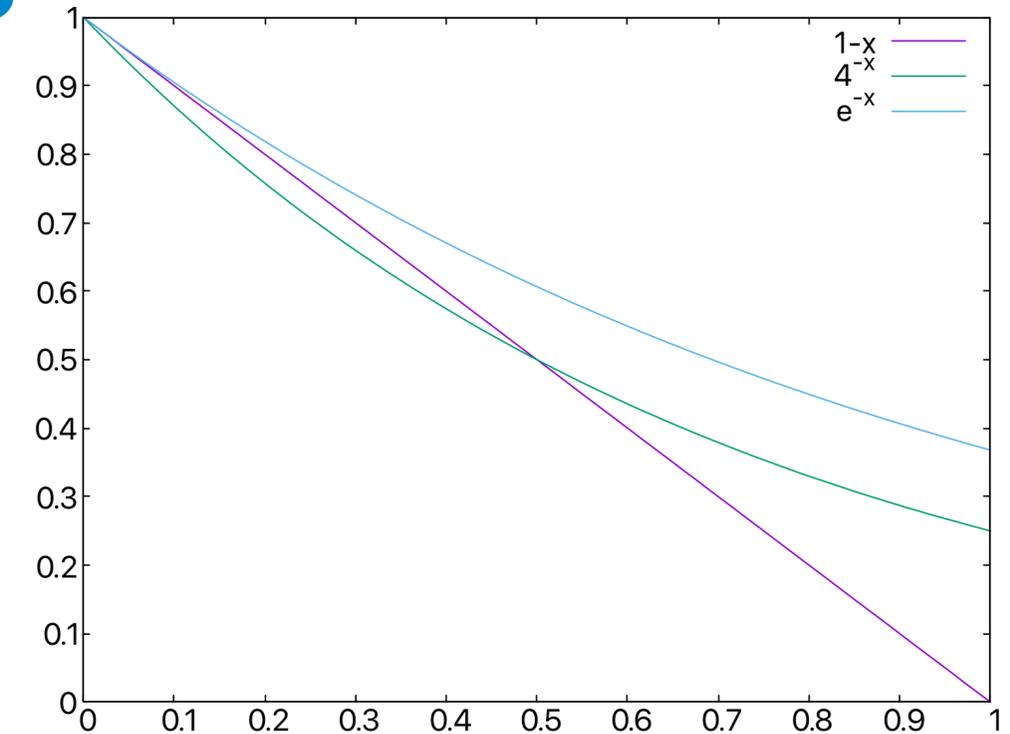
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 - $4^{-x} \leq 1 - x \leq e^{-x}$ for all $x \in [0, 1/2]$



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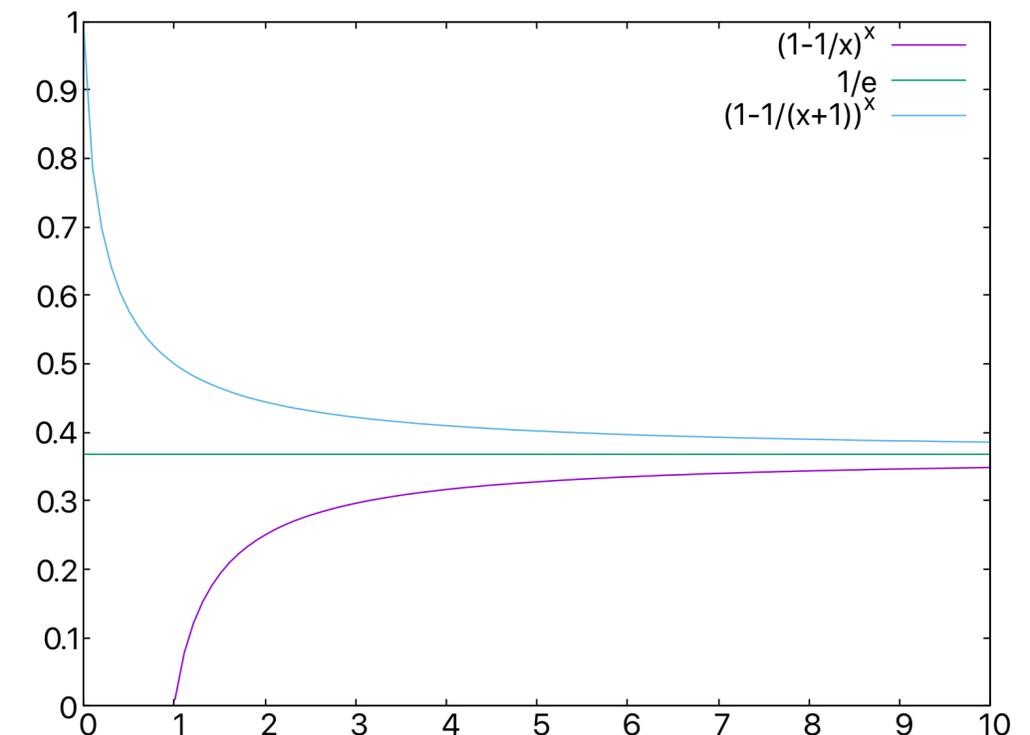
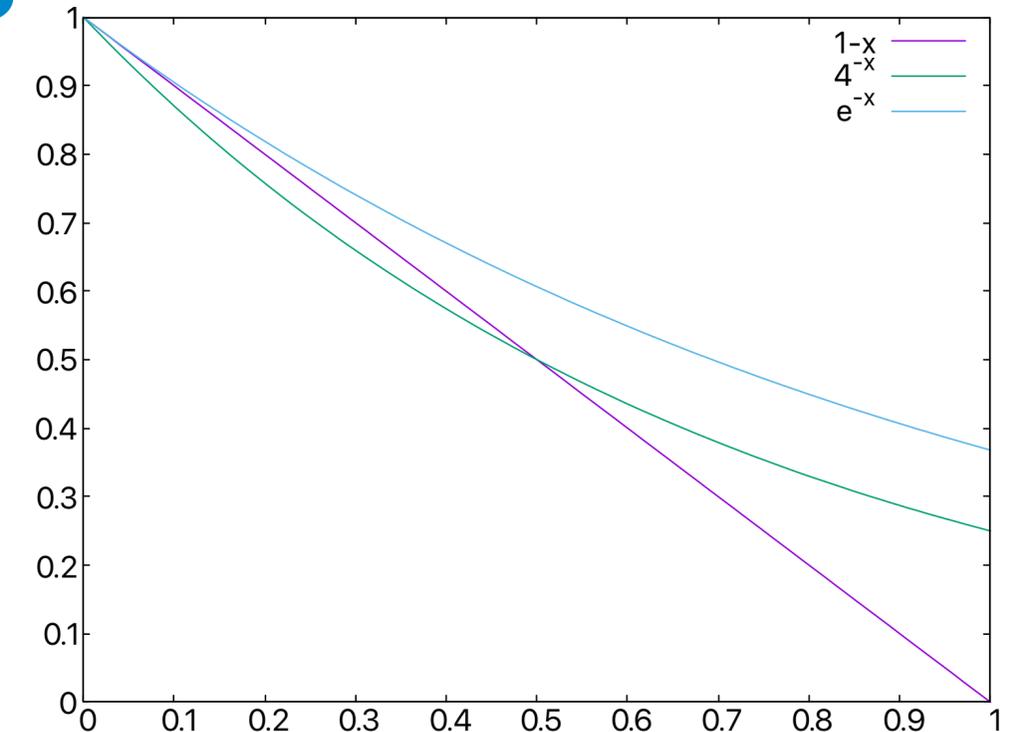
► $1 - x \geq 4^{-x}$ for all $x \in [0, 1/2]$

• $4^{-x} \leq 1 - x \leq e^{-x}$ for all $x \in [0, 1/2]$

► $\lim_{x \rightarrow \infty} (1 - 1/x)^x = 1/e$

• $(1 - 1/x)^x < 1/e$ for all $x \geq 1$

• $(1 - 1/(x+1))^x > 1/e$ for all $x > 0$



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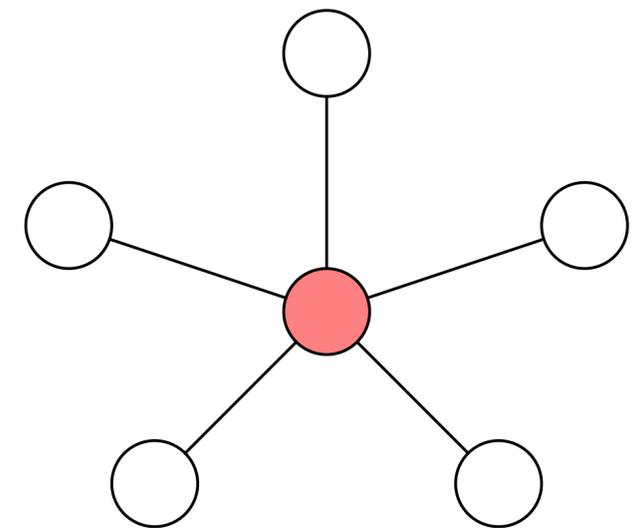
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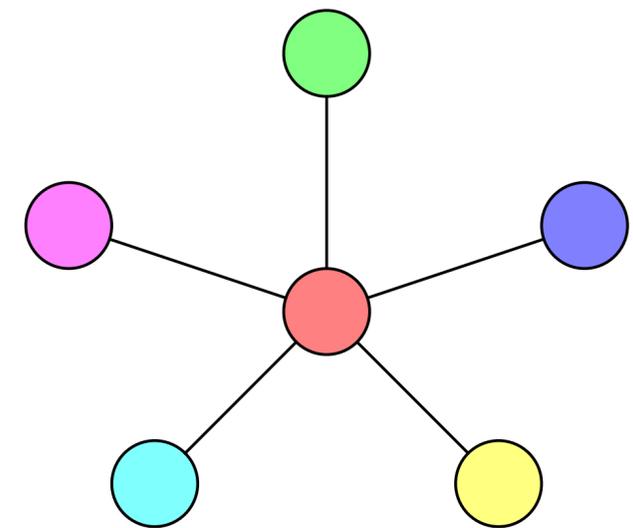
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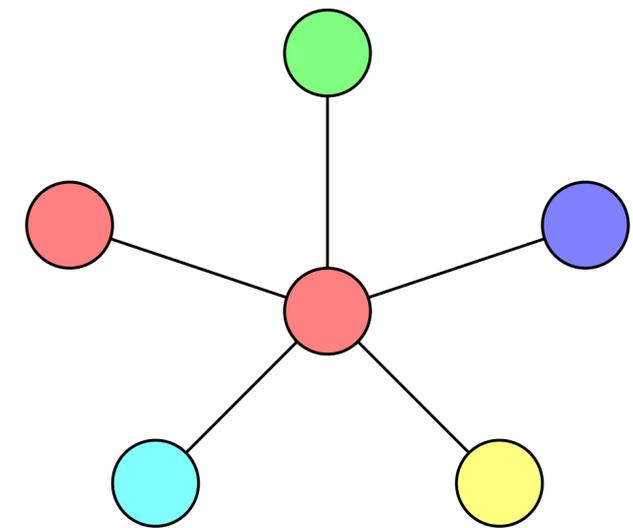
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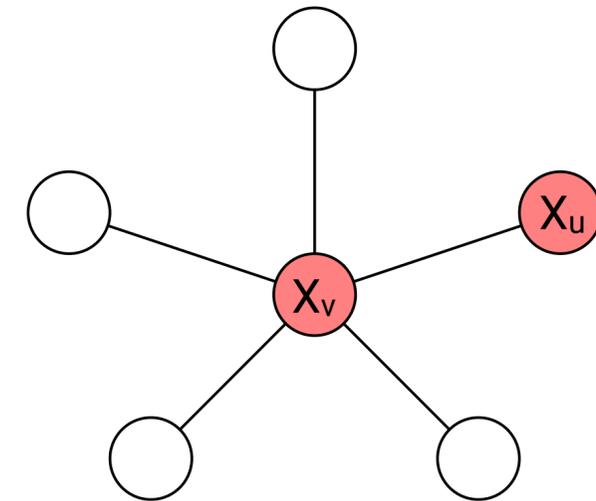
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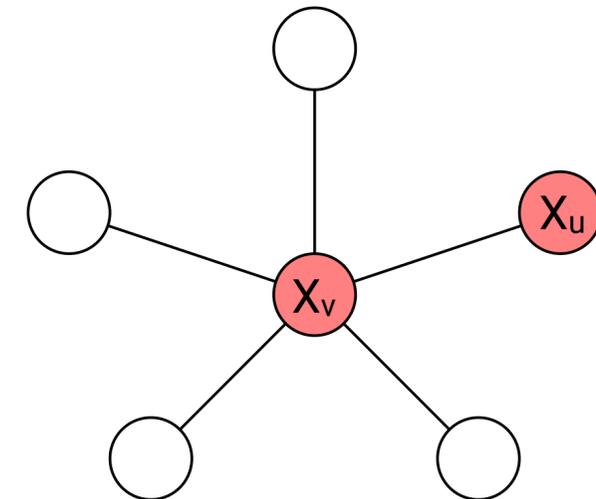
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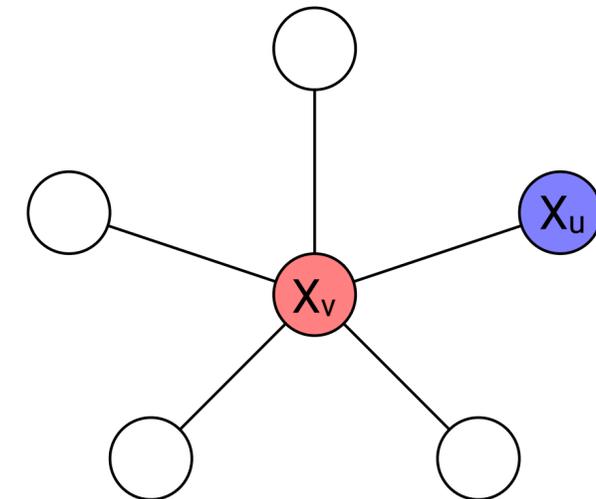


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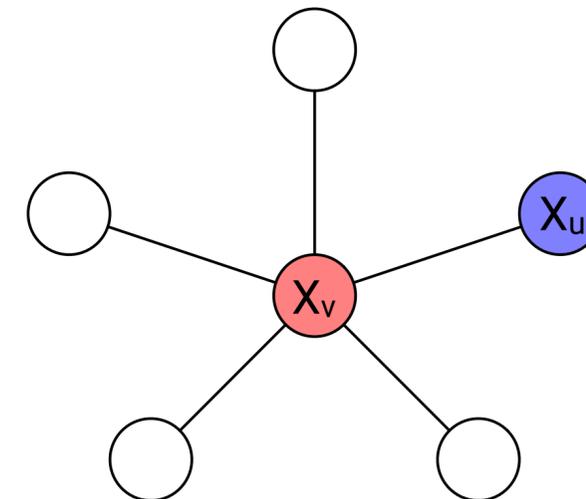
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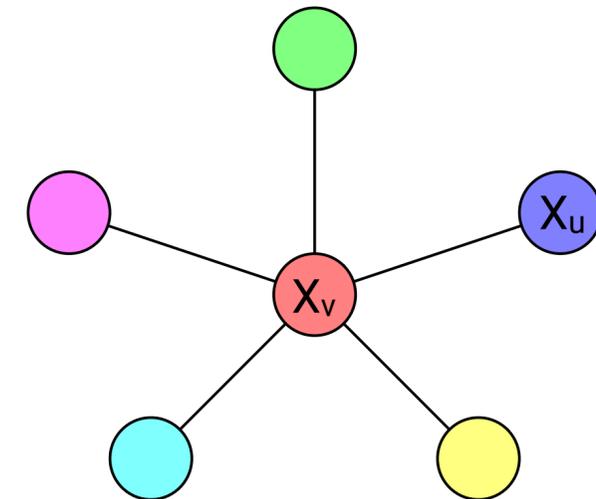
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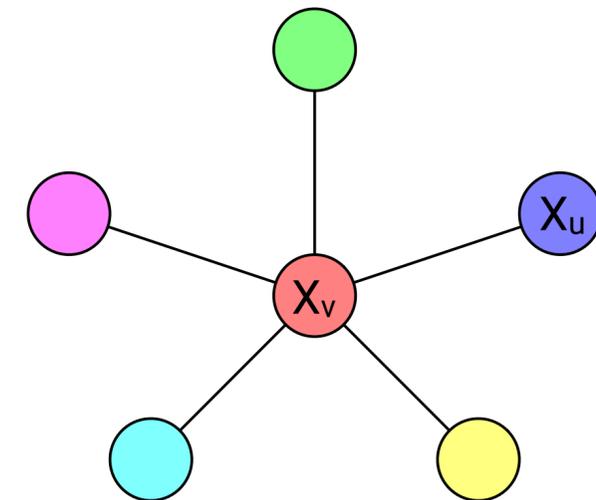
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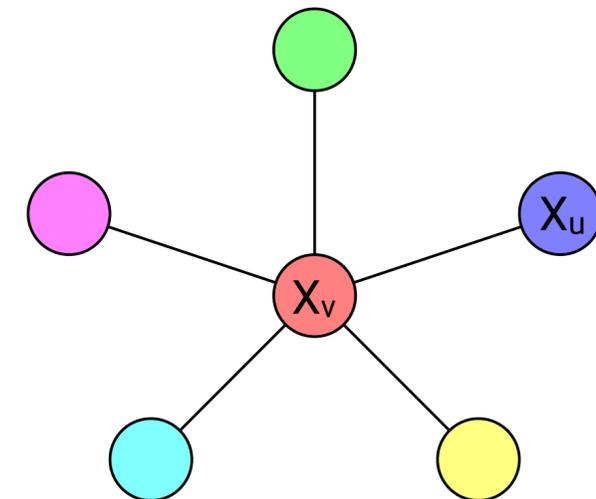
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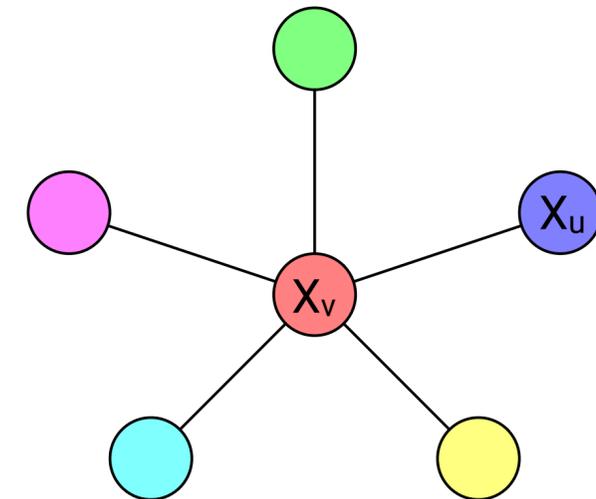
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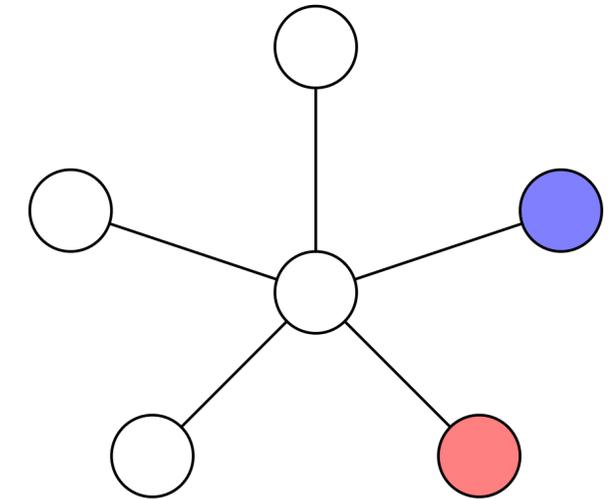
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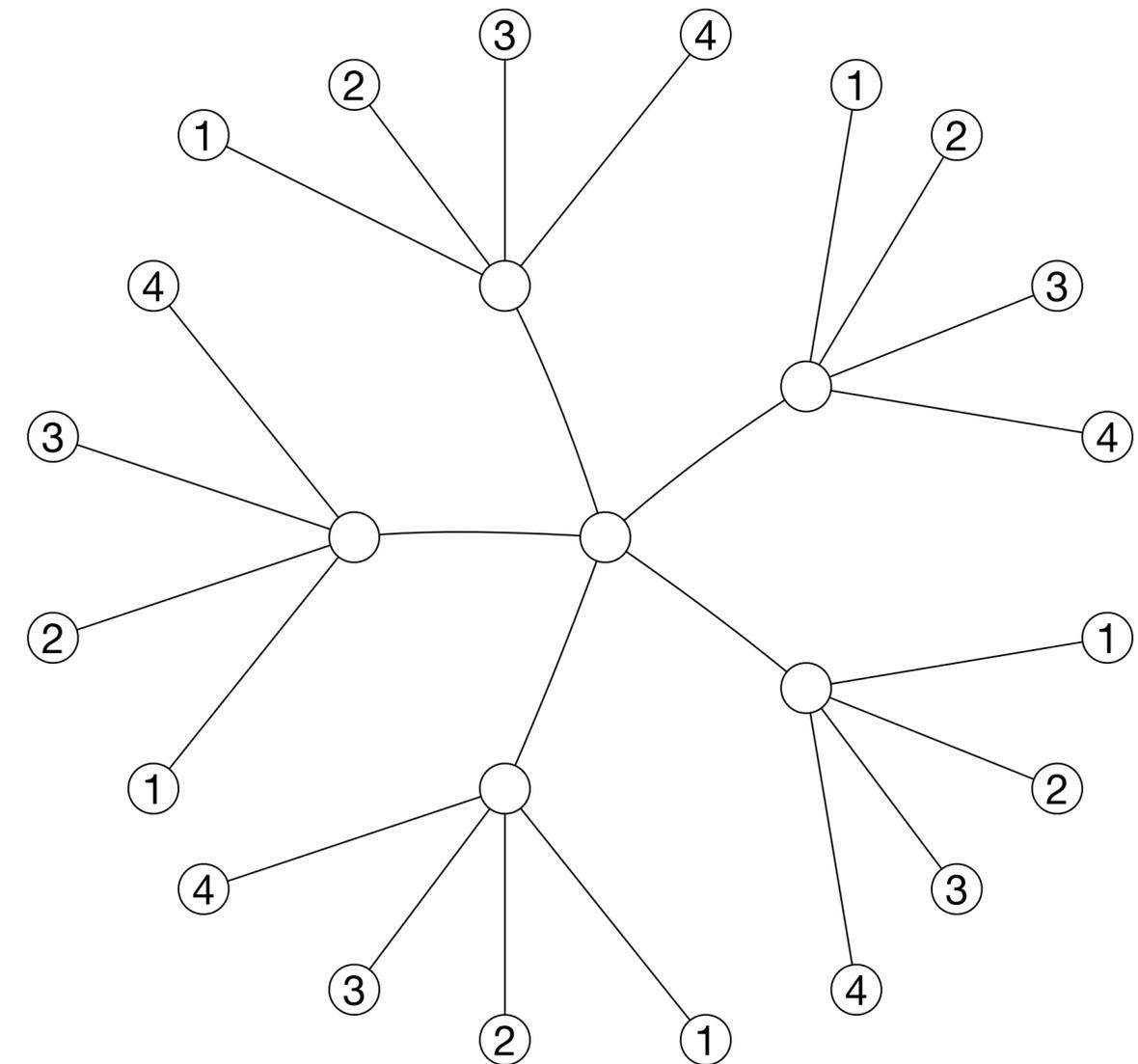
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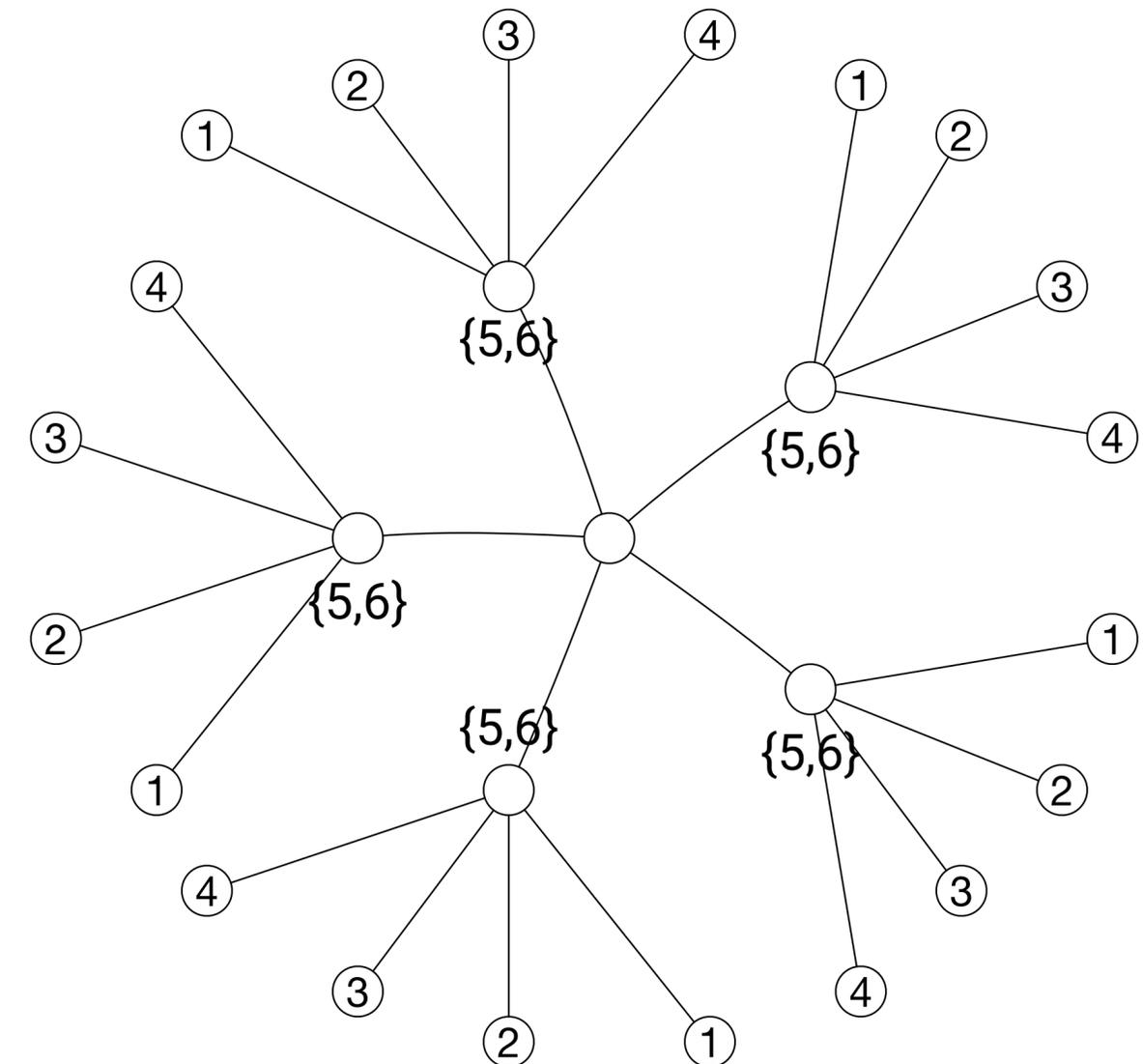
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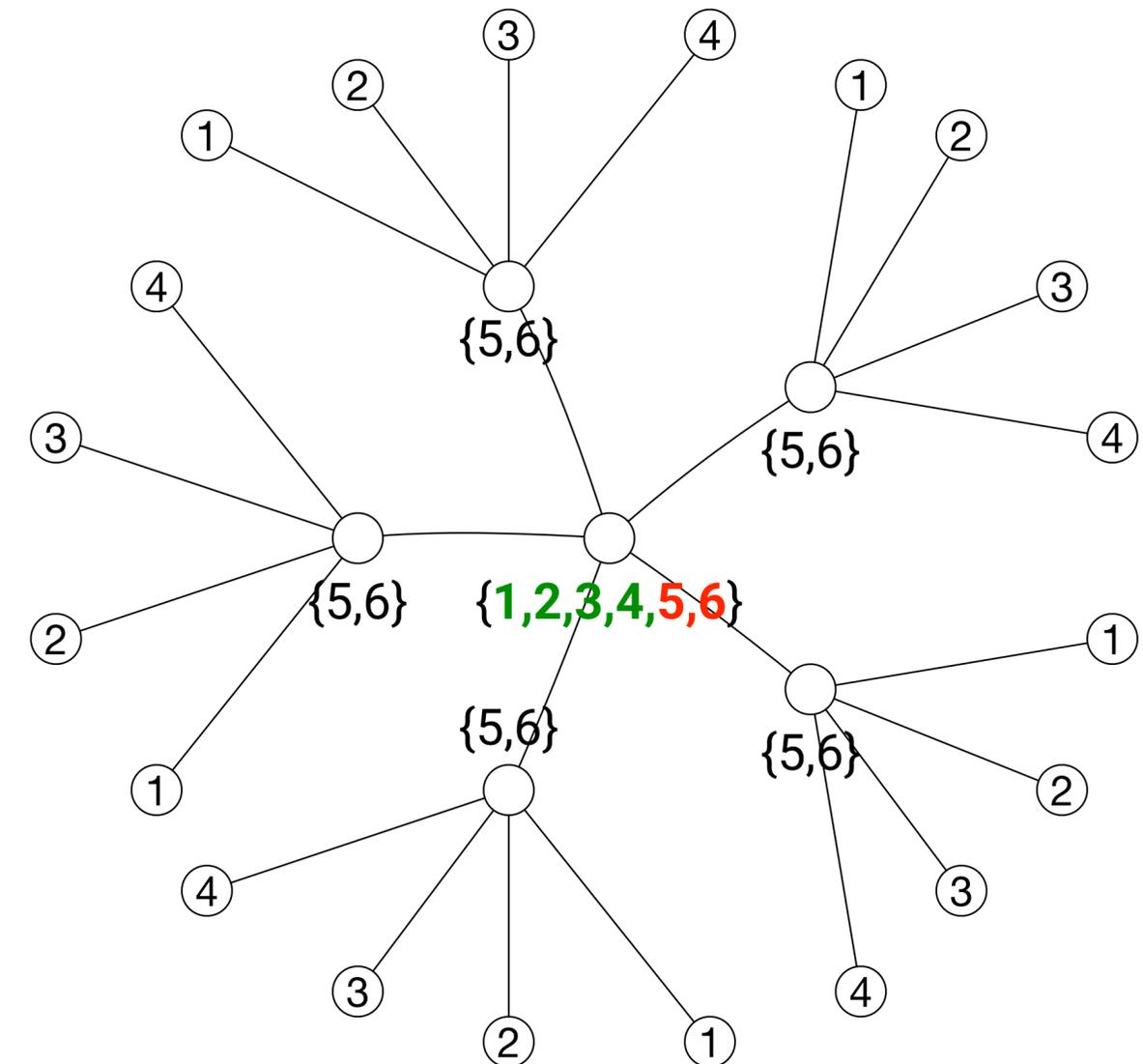
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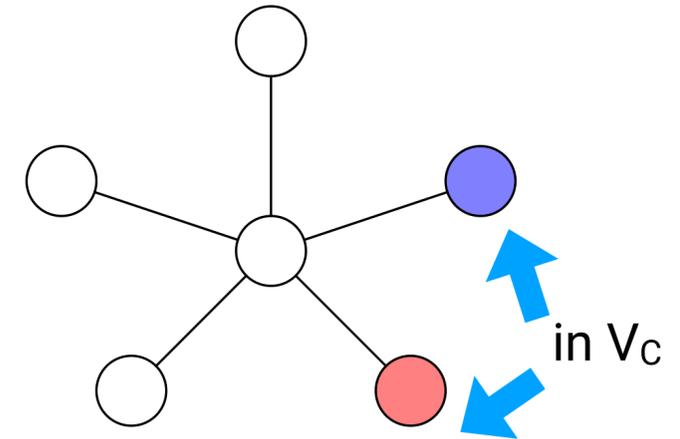
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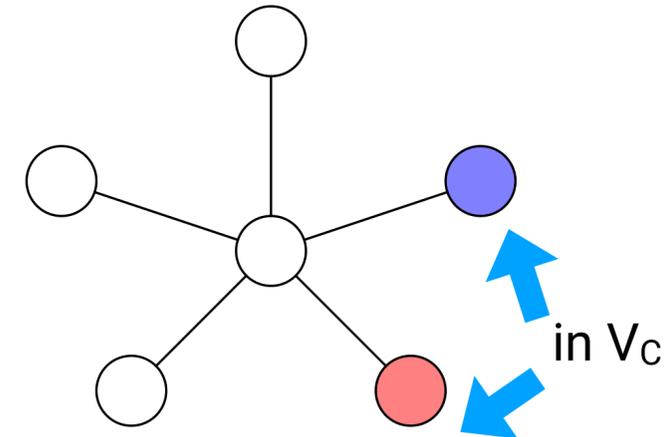
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- ▶ V_C : nodes $v \in V_C \subset V$ already have a color x_v such that $G[V_C]$ is properly colored



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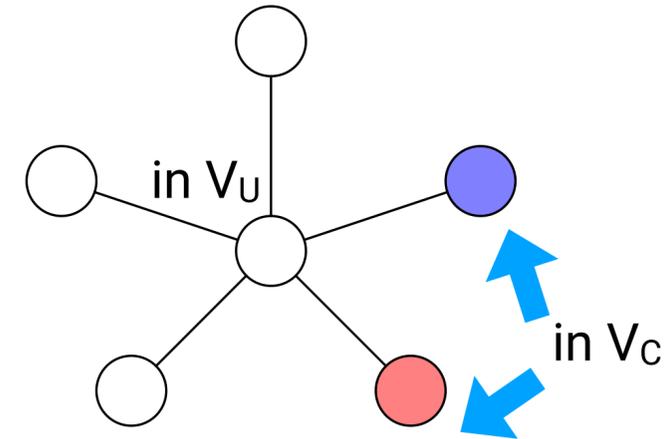
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Subgraph induced by nodes in V_c



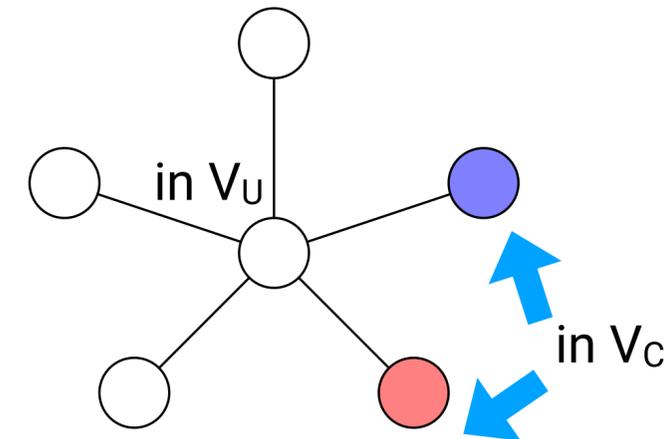
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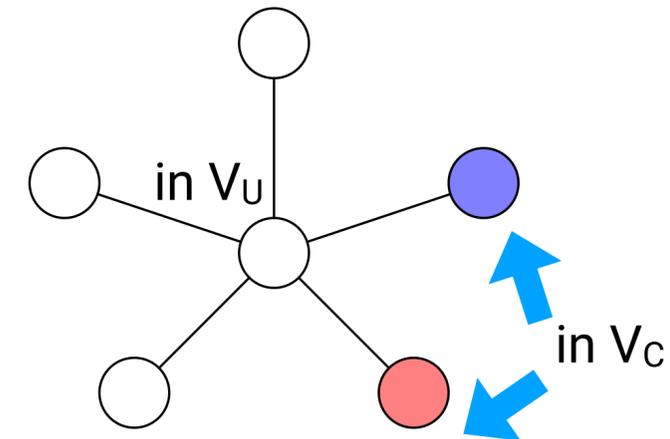
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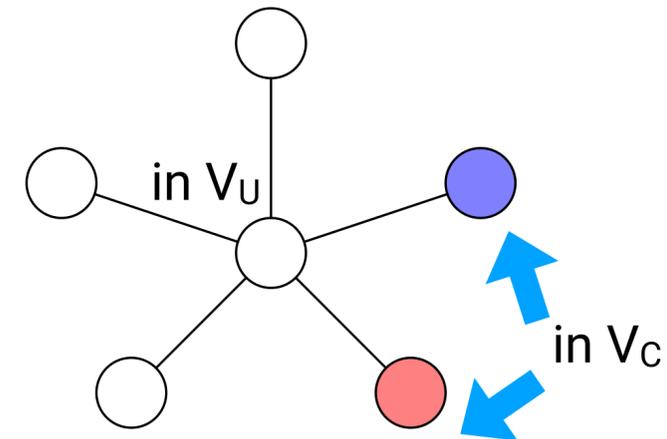
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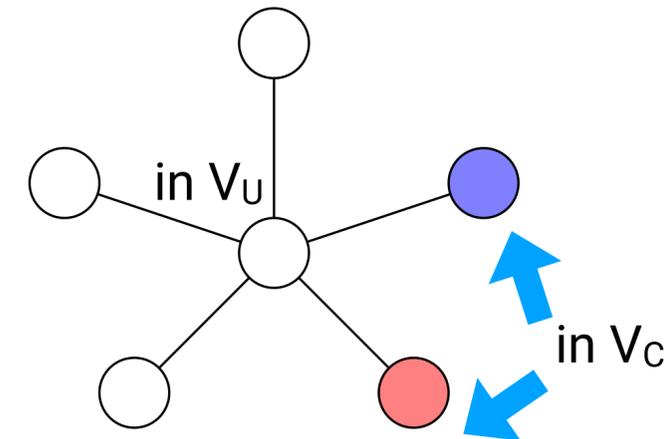
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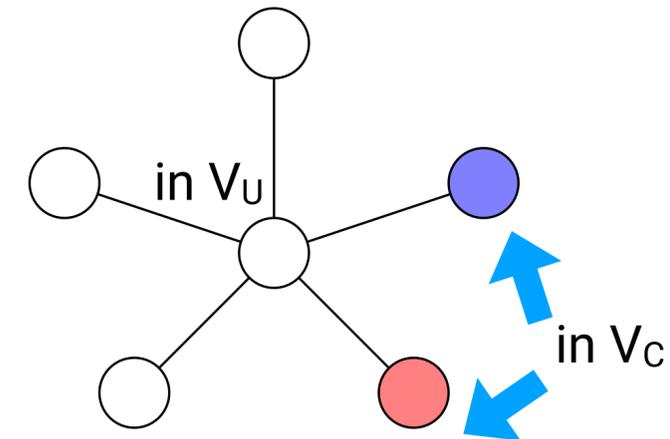
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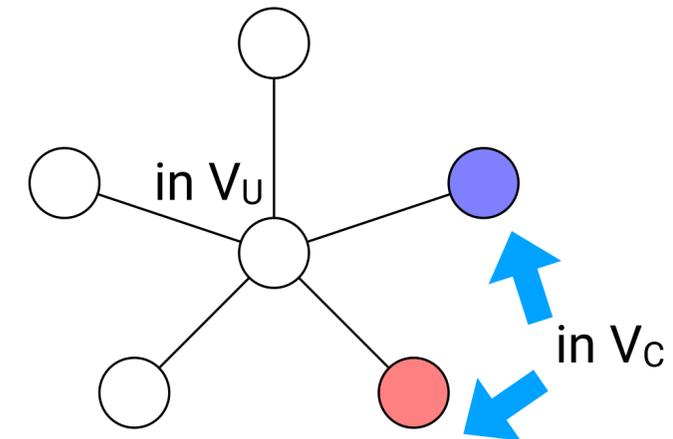
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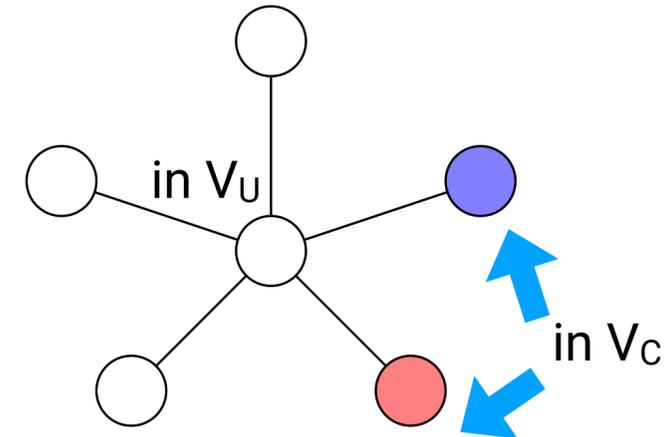
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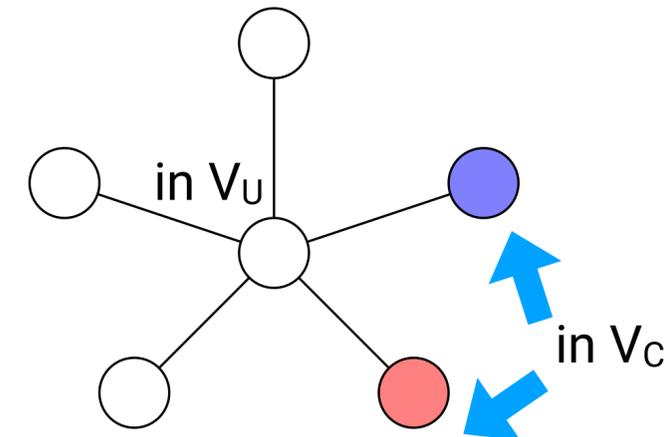
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Extending an Existing Coloring

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Extending an Existing Coloring

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Extending an Existing Coloring

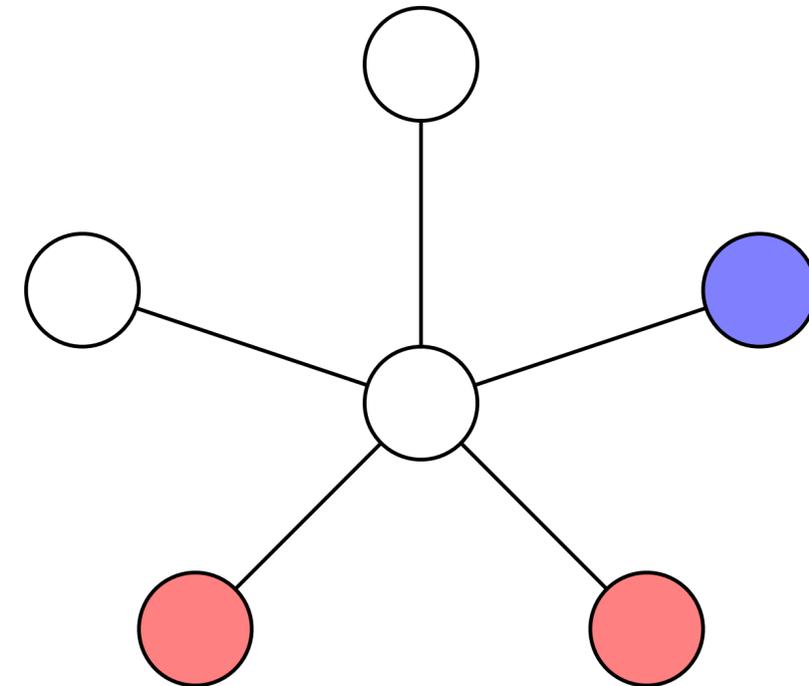
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← Let $f(w_x) = 4^{-w_x(v)}$. This is **average**($f(w_x)$)

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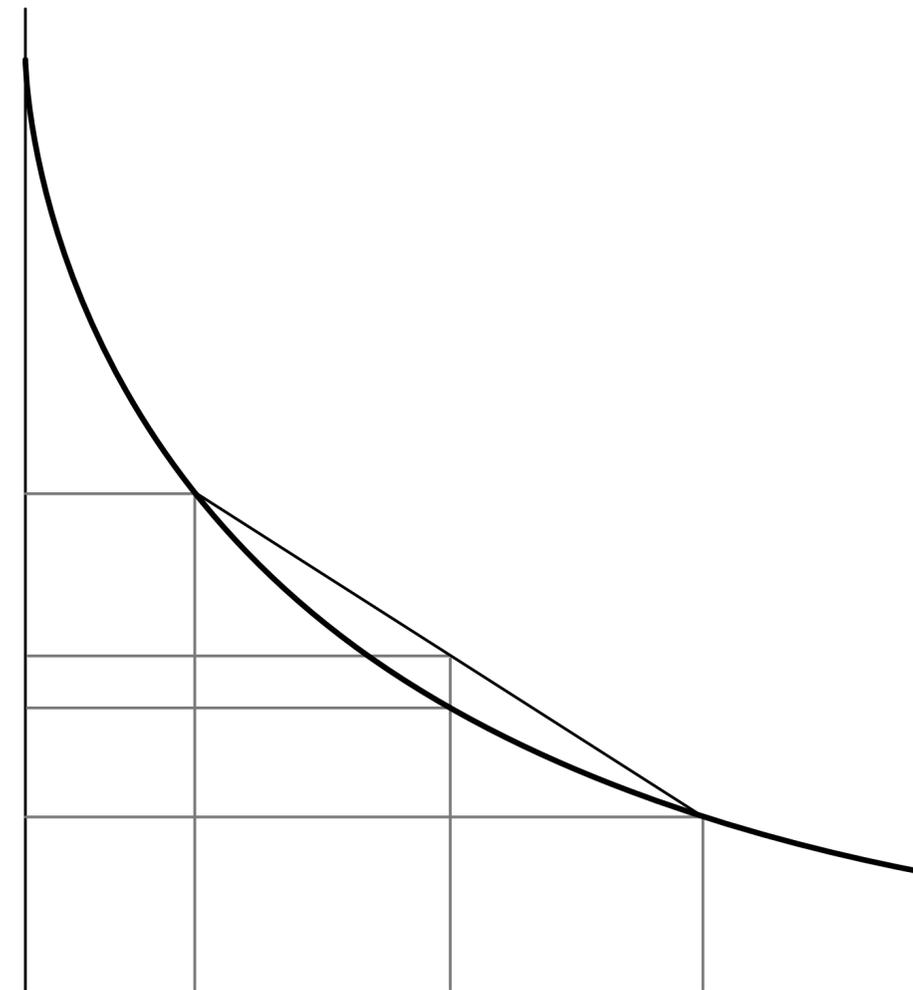
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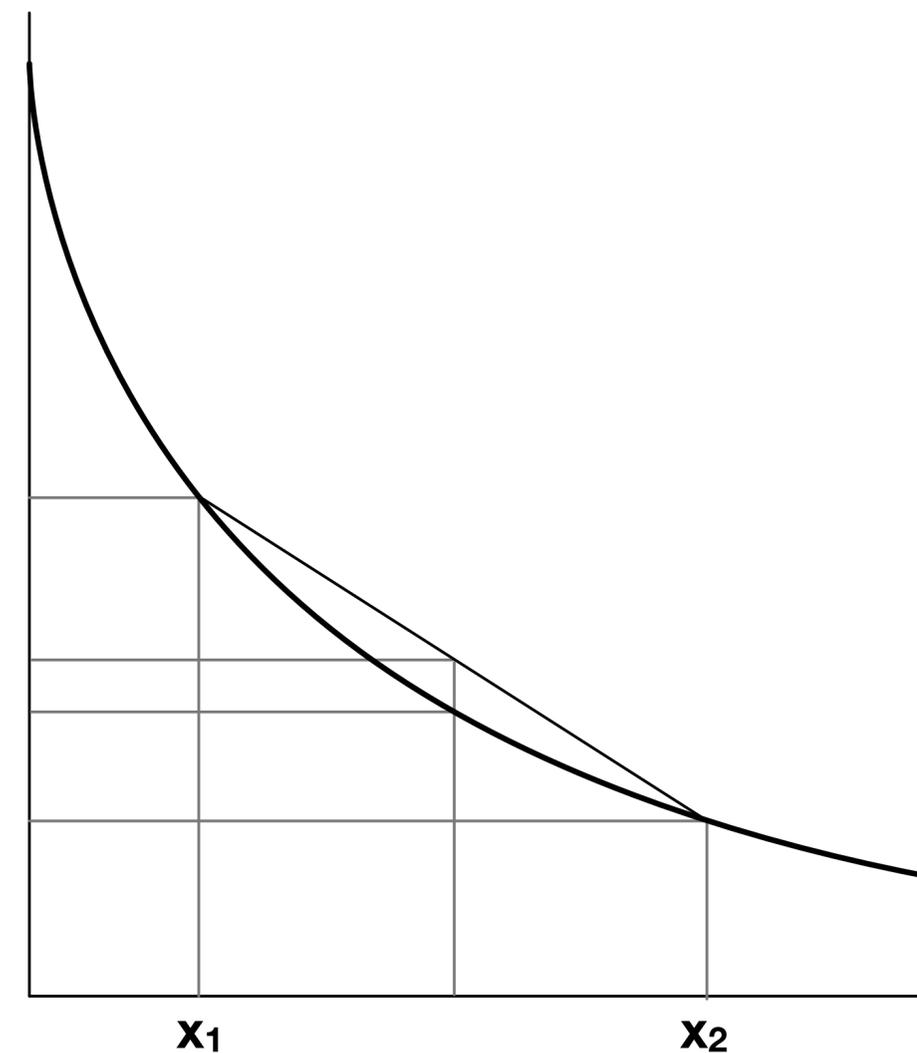
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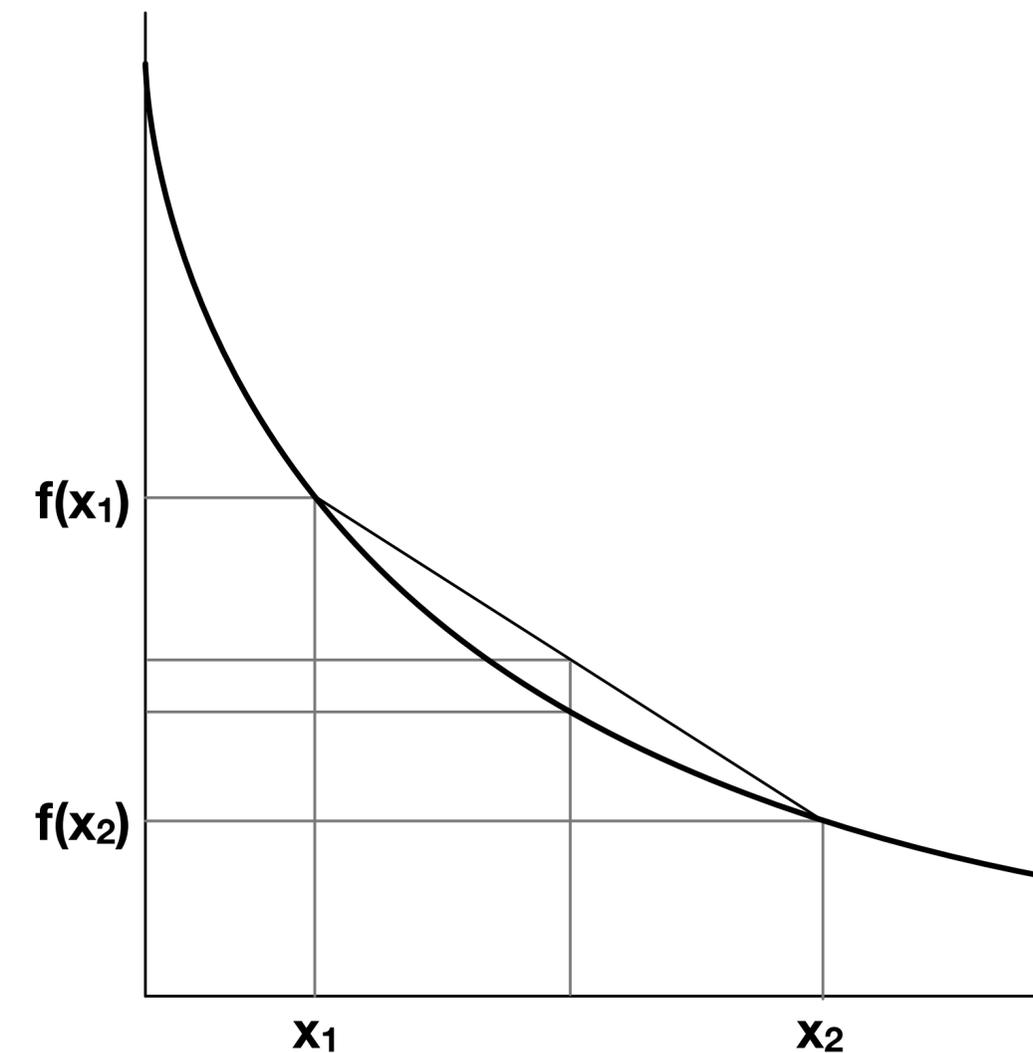
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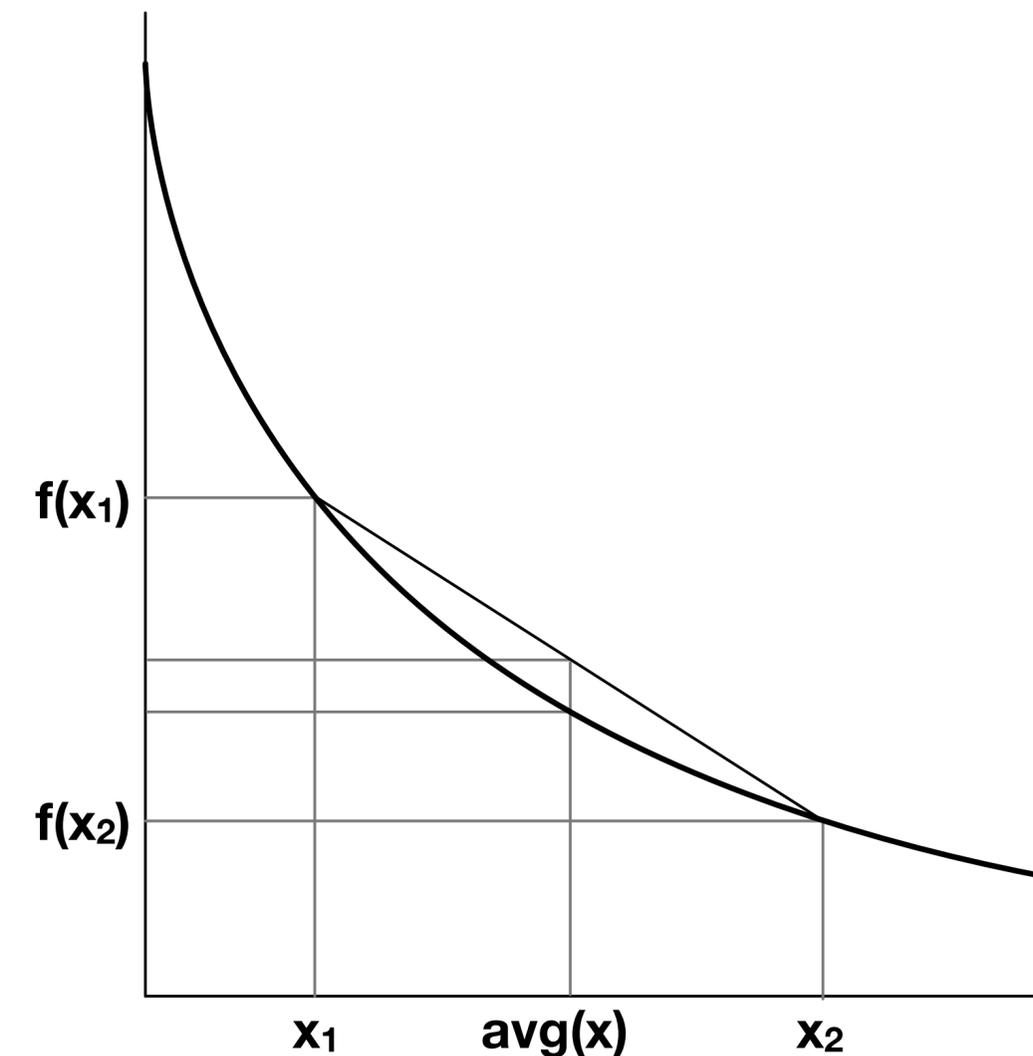
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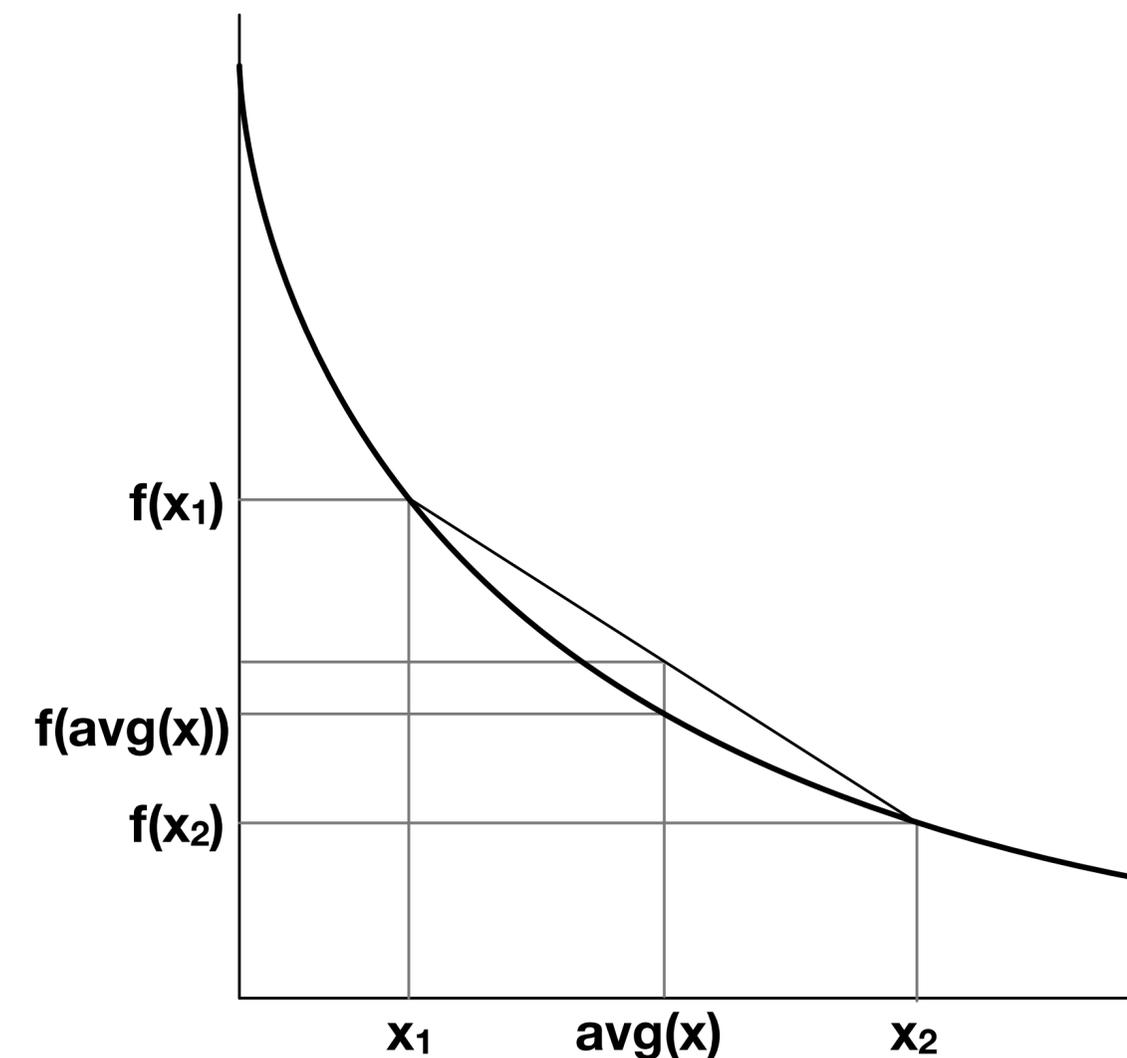
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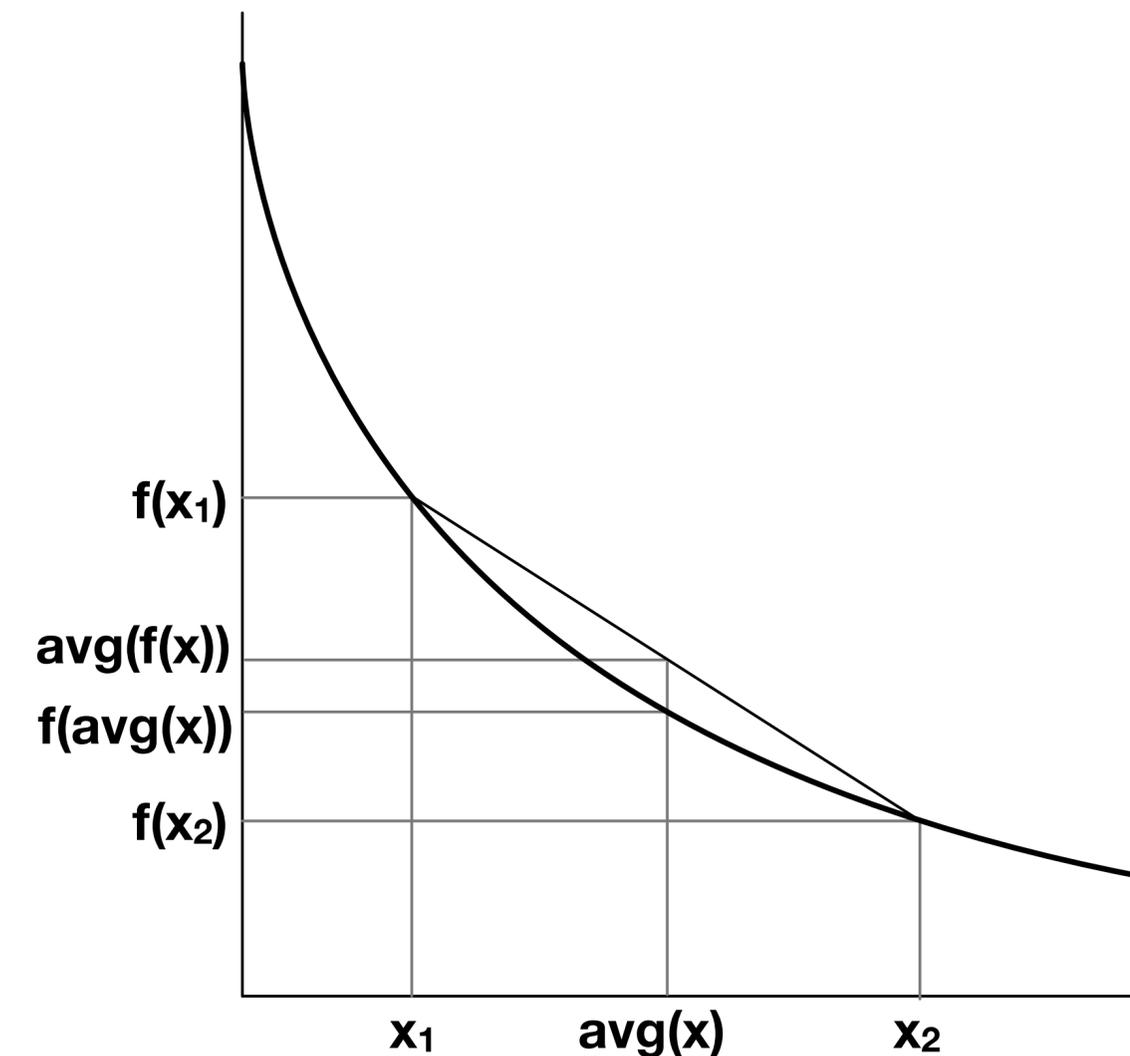
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Union Bound: $P(A \cup B) = P(A) + P(B) - P(A \cap B) \leq P(A) + P(B)$

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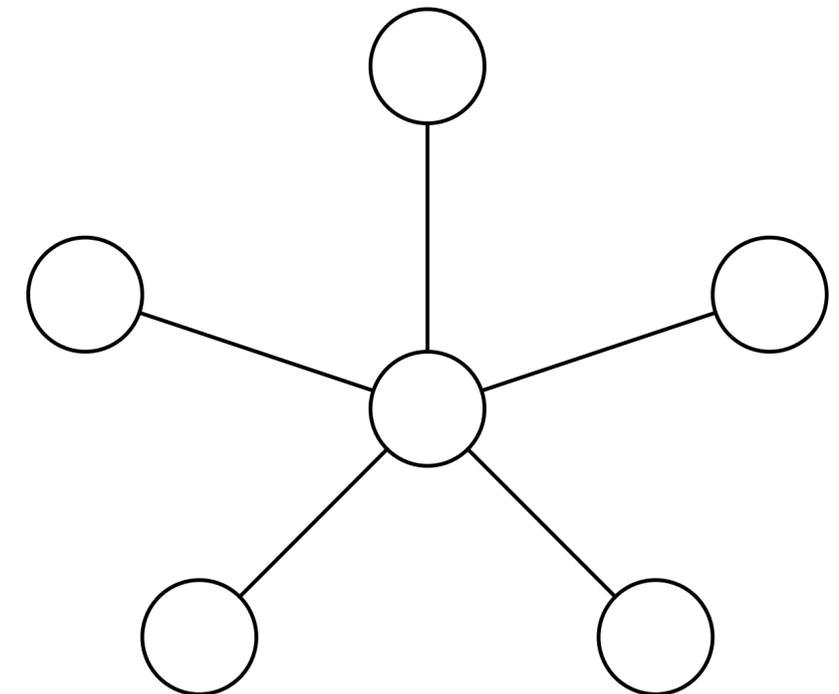
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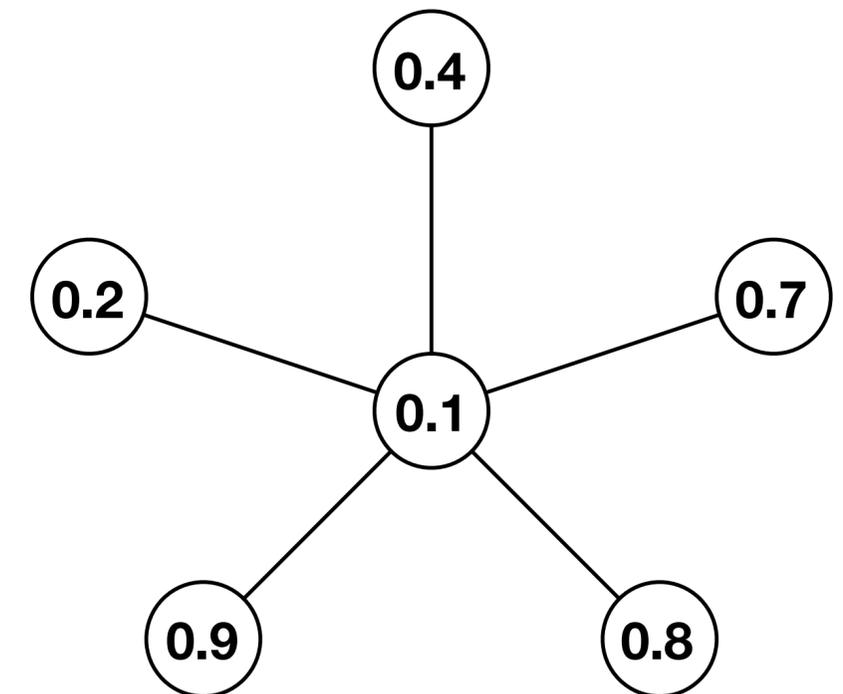
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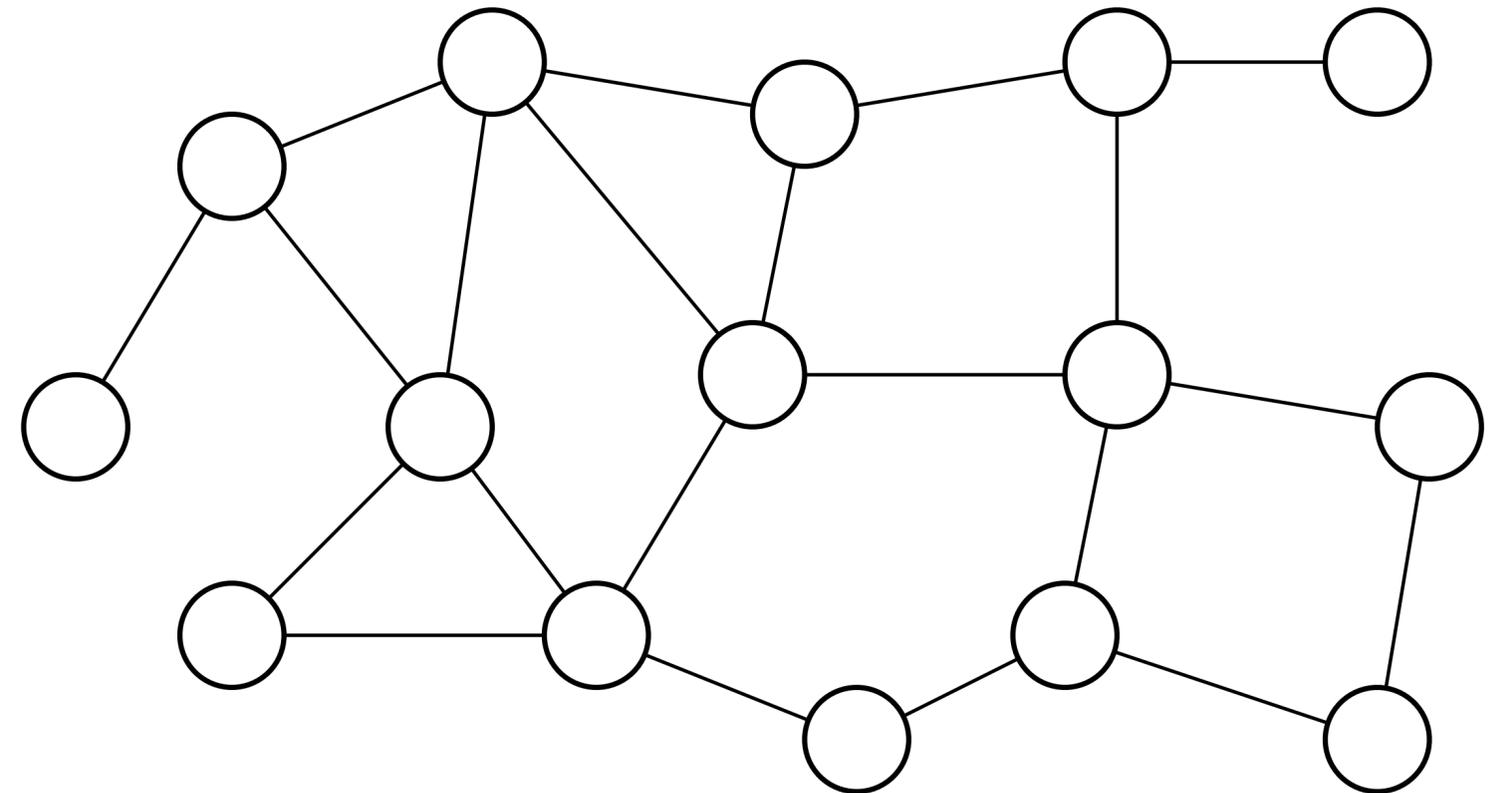
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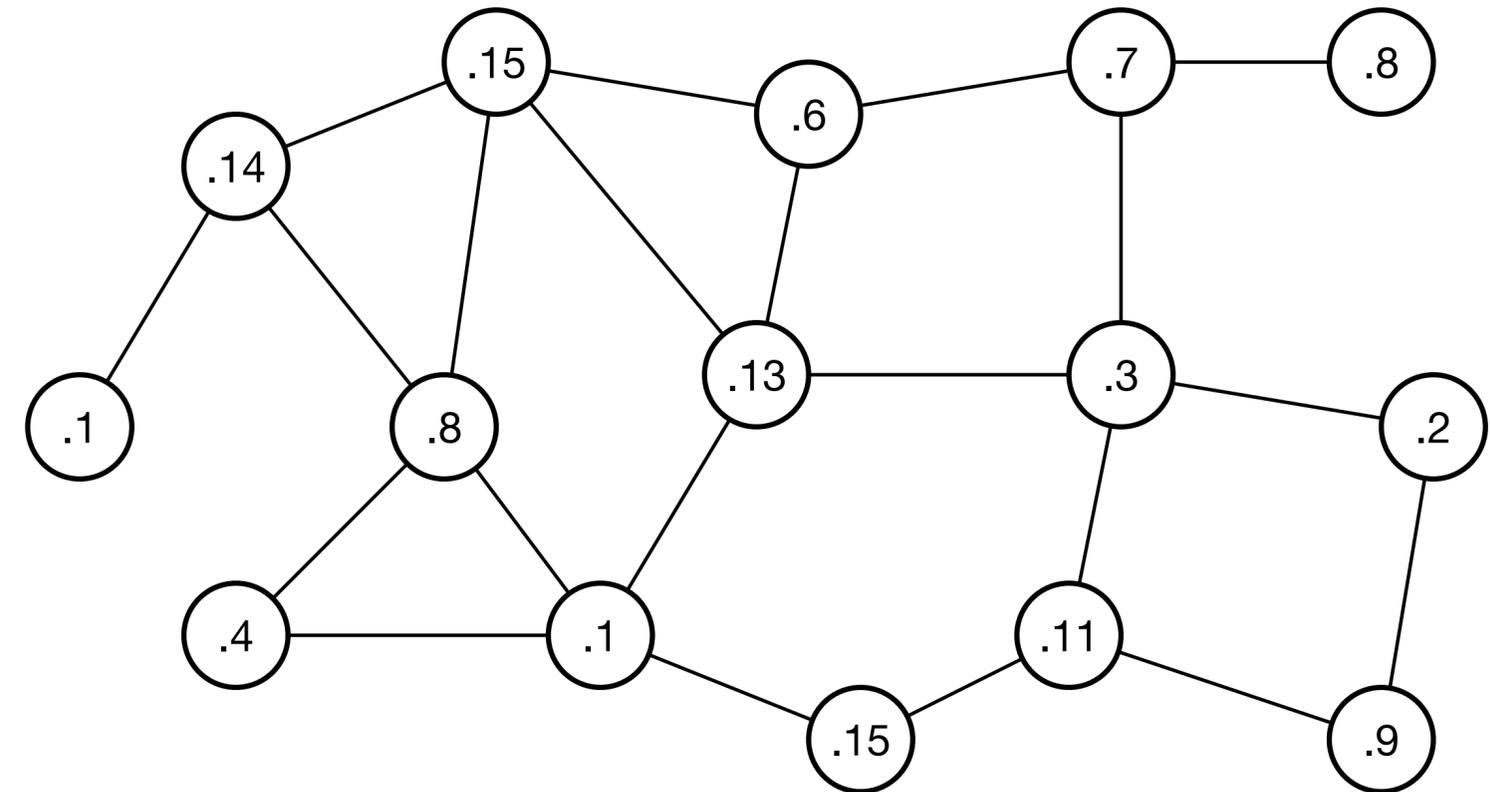
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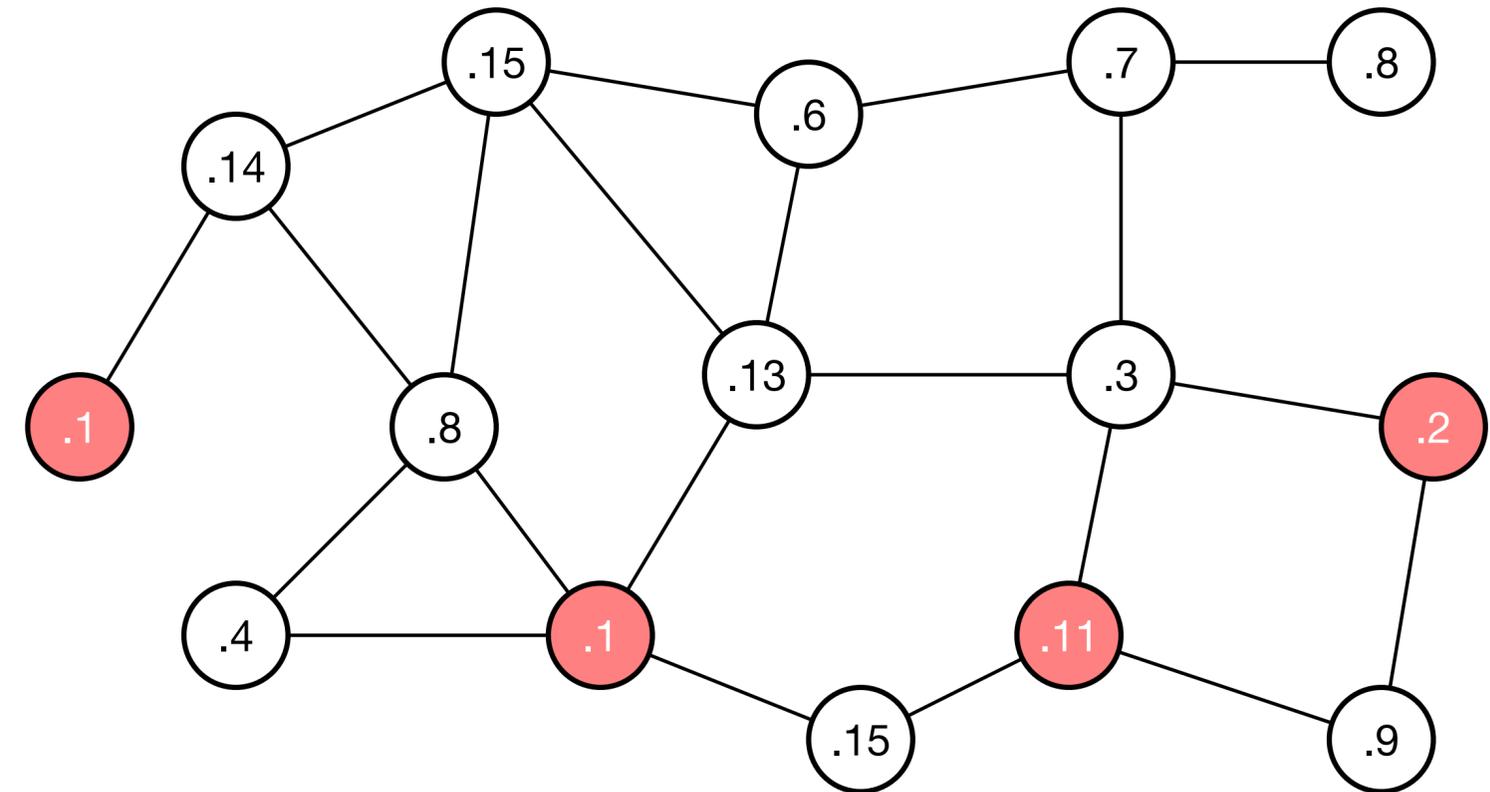
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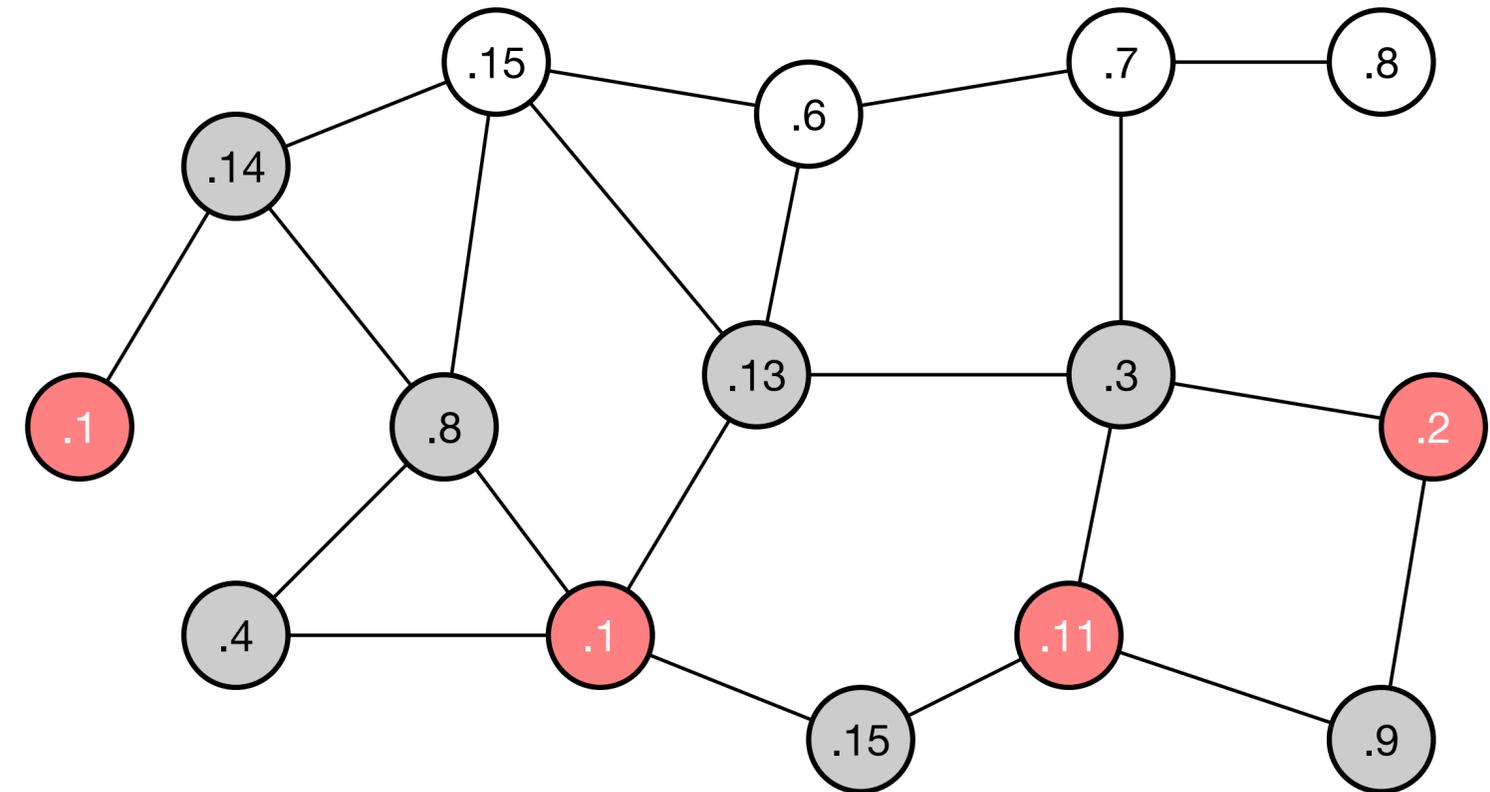
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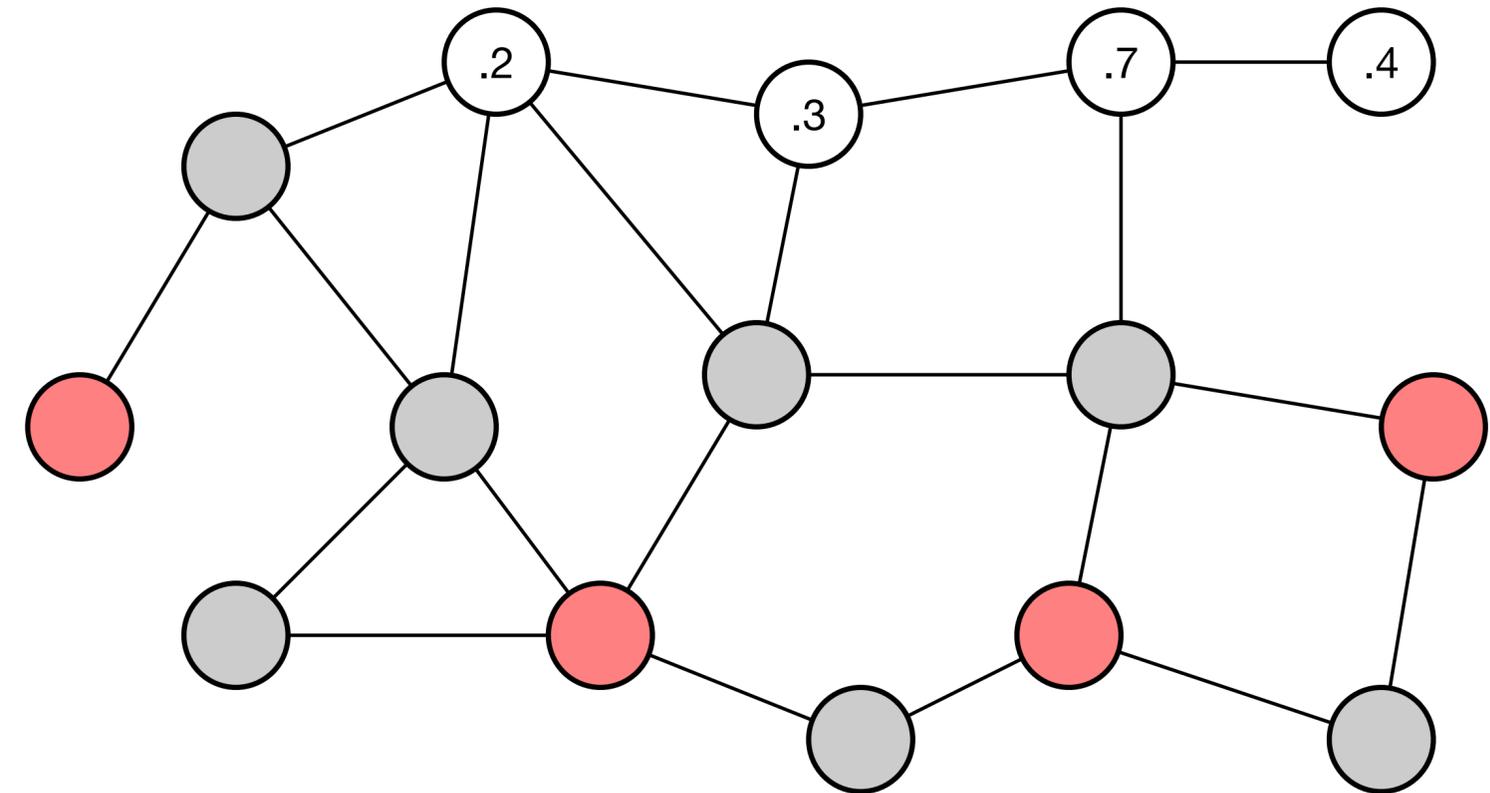
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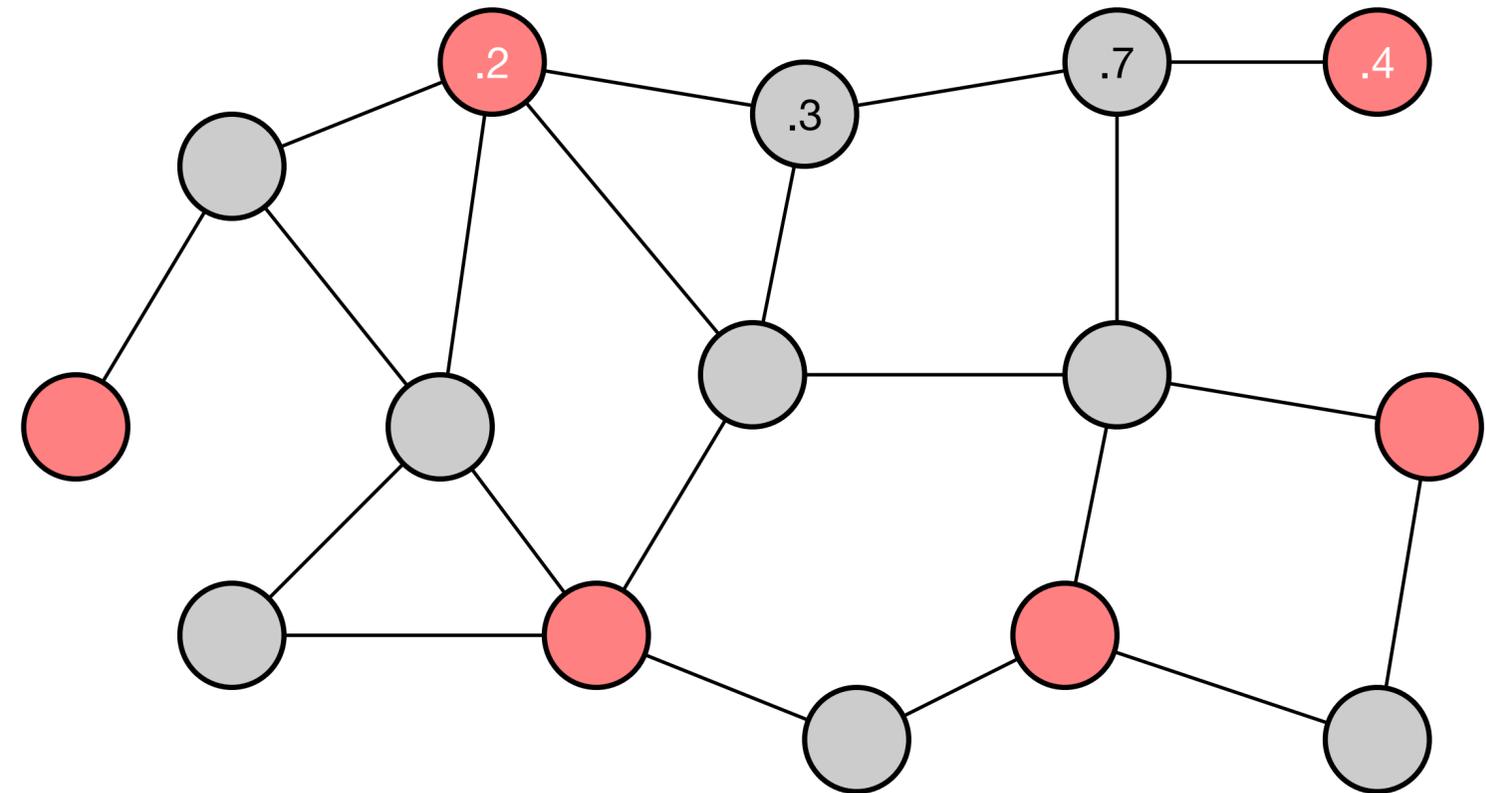
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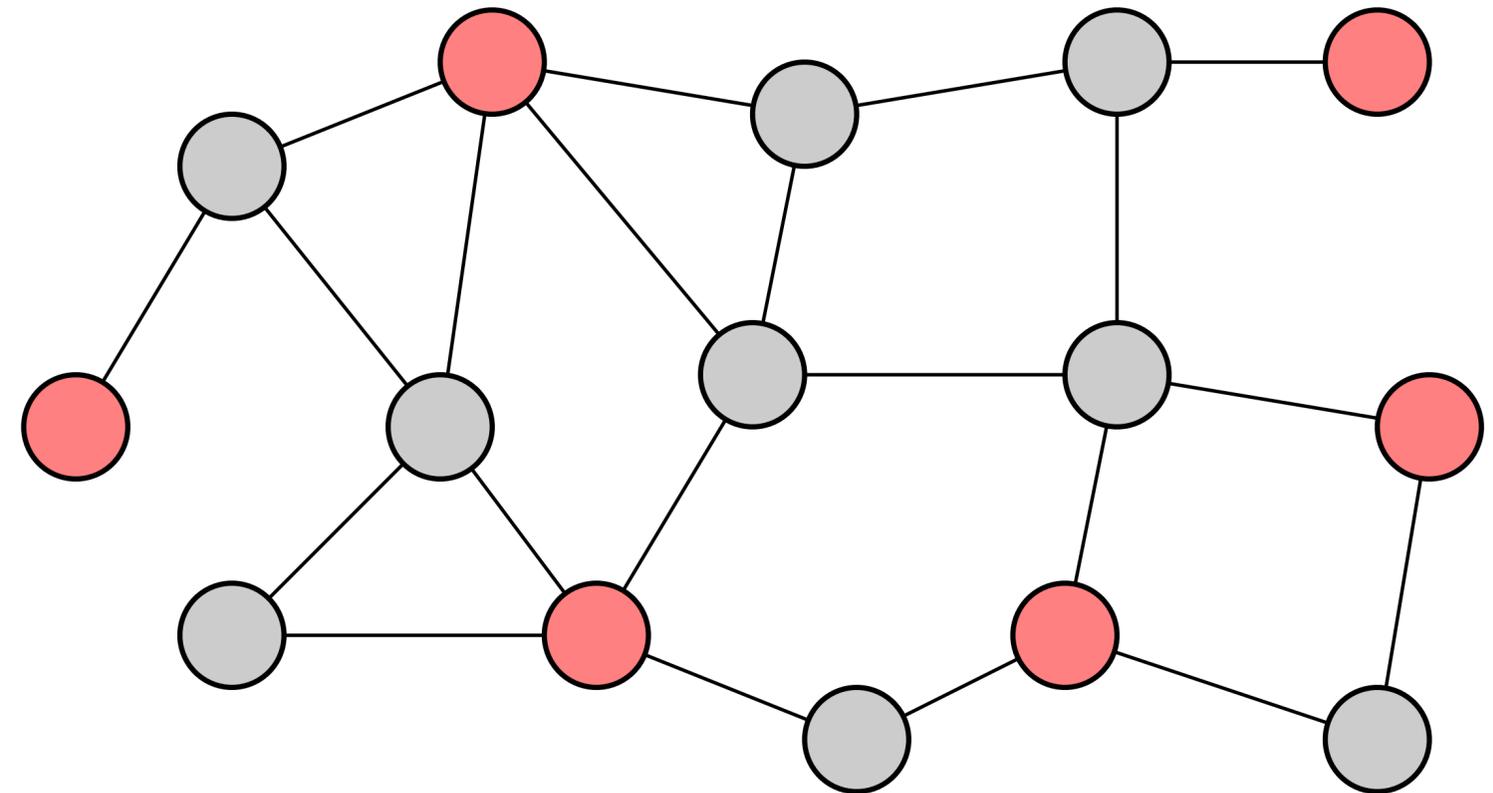
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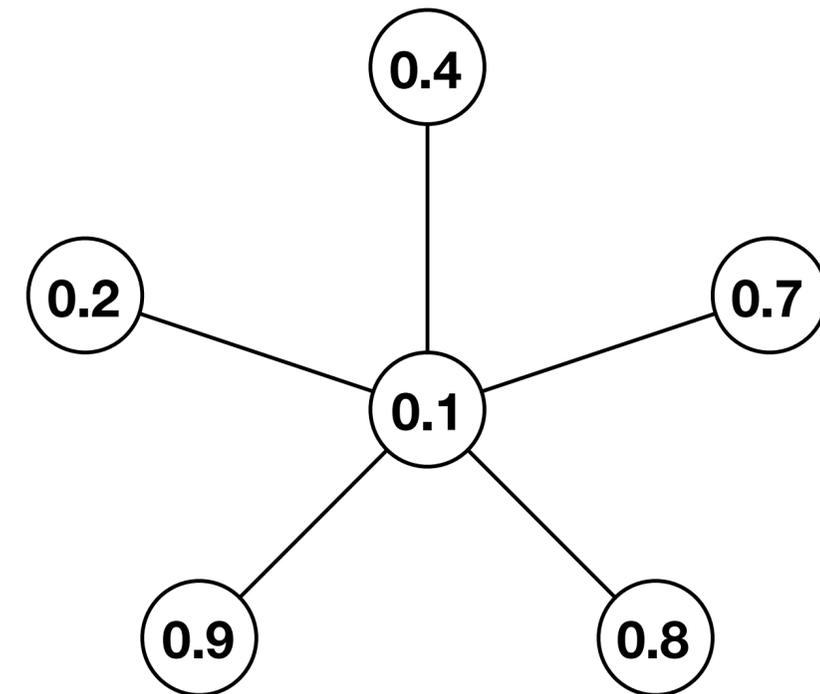


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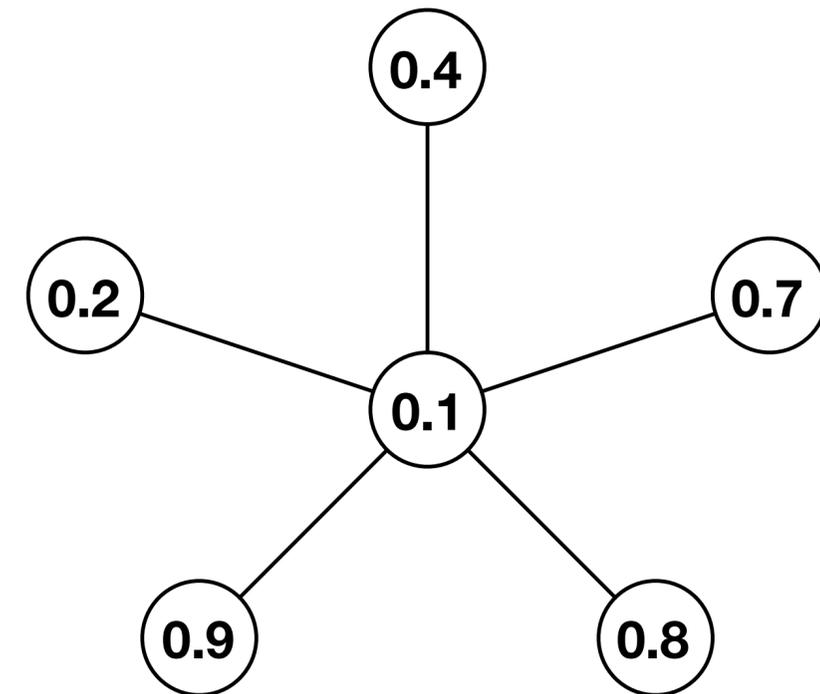
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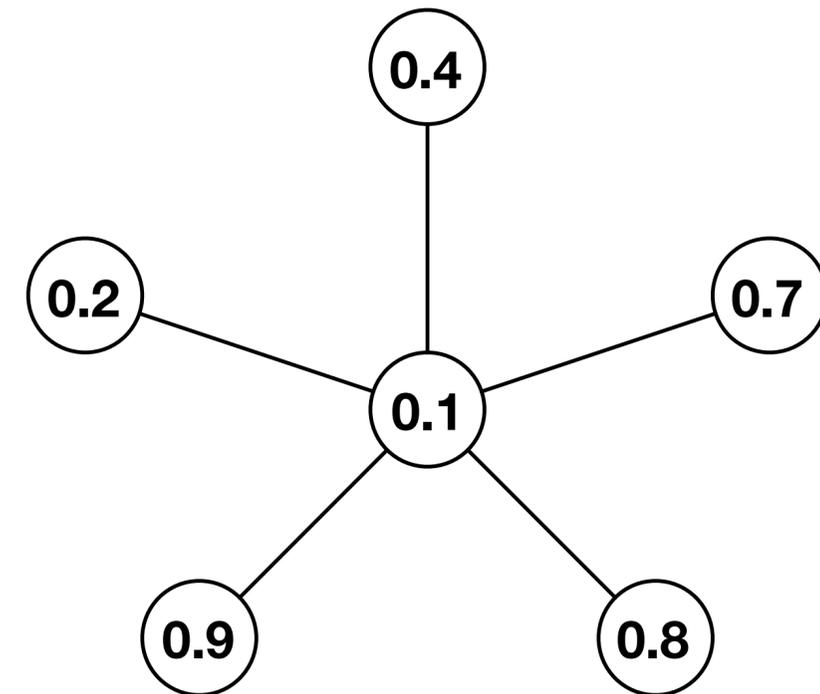
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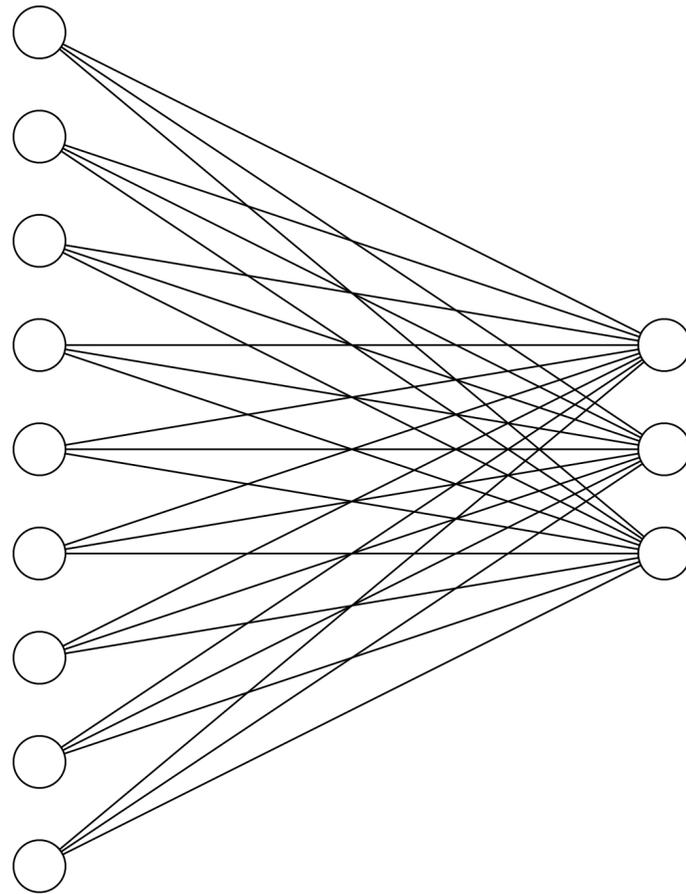
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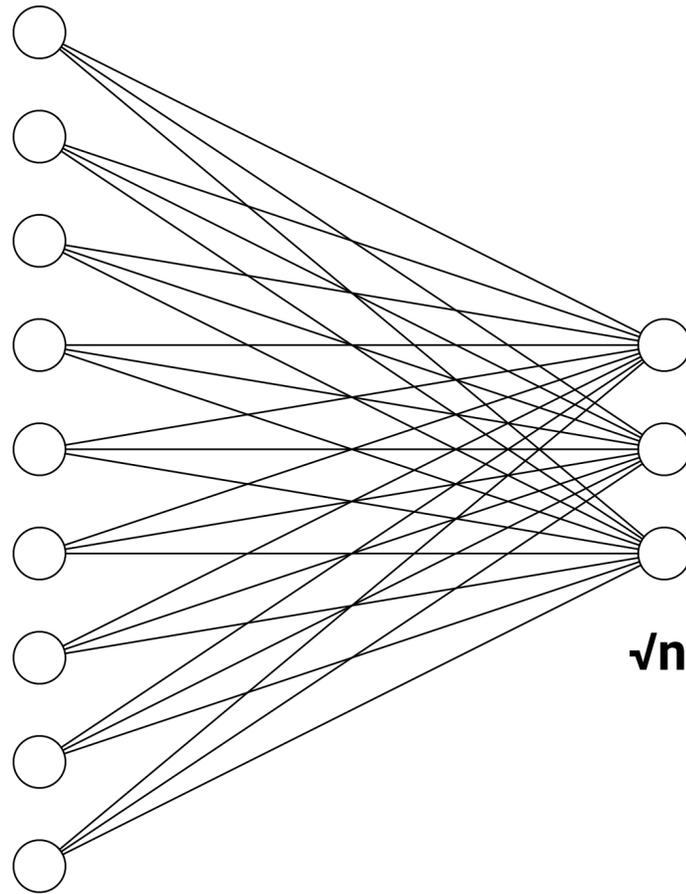
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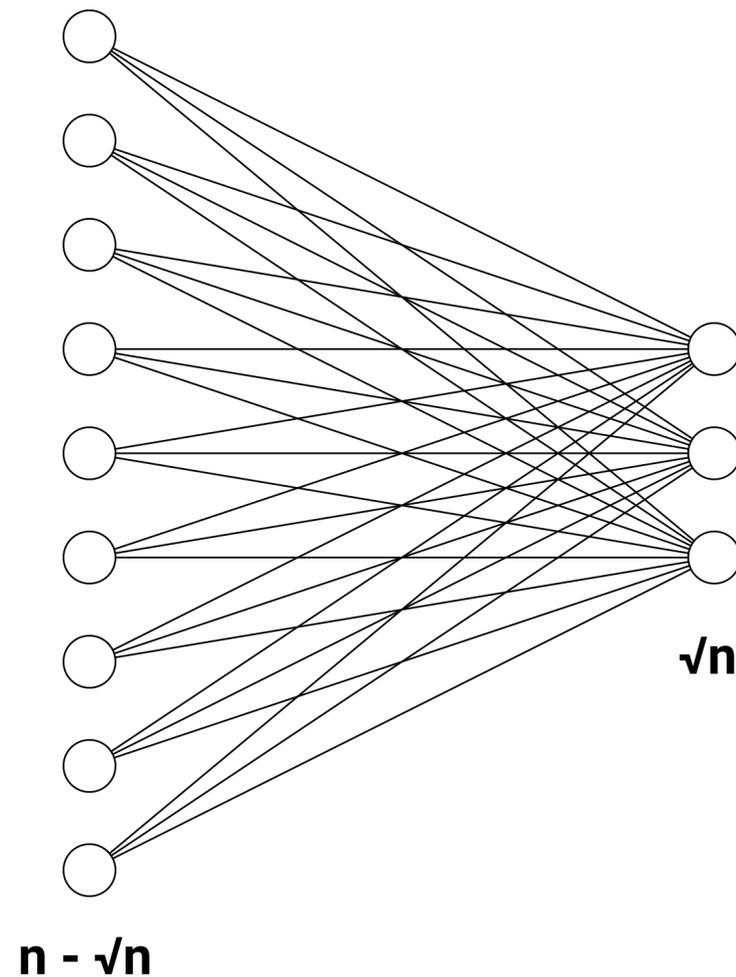
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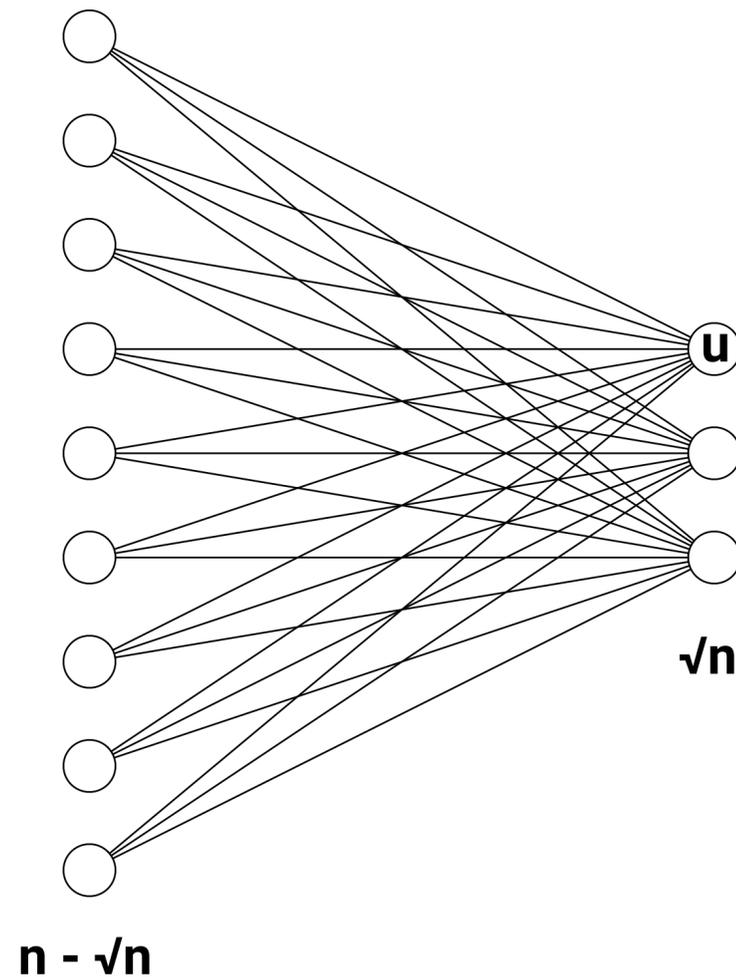
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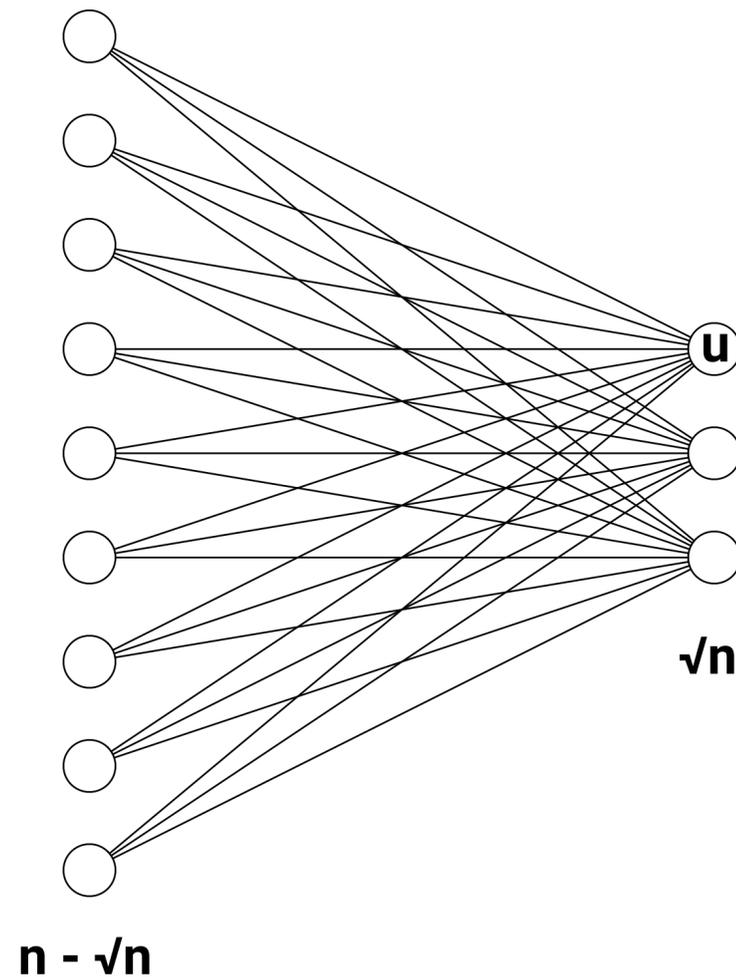
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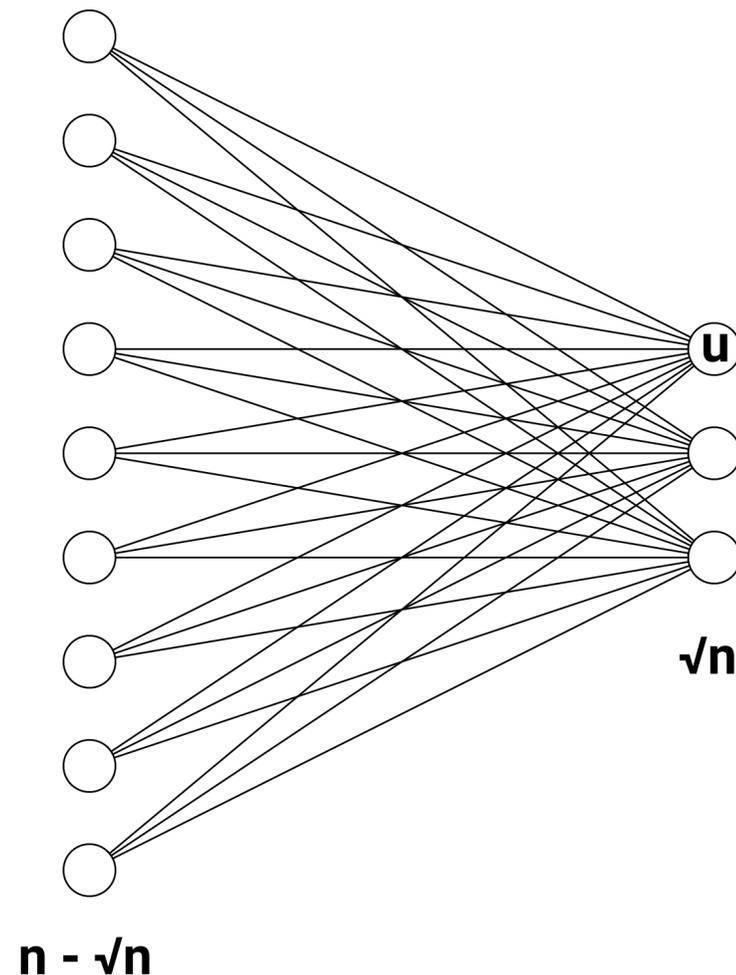
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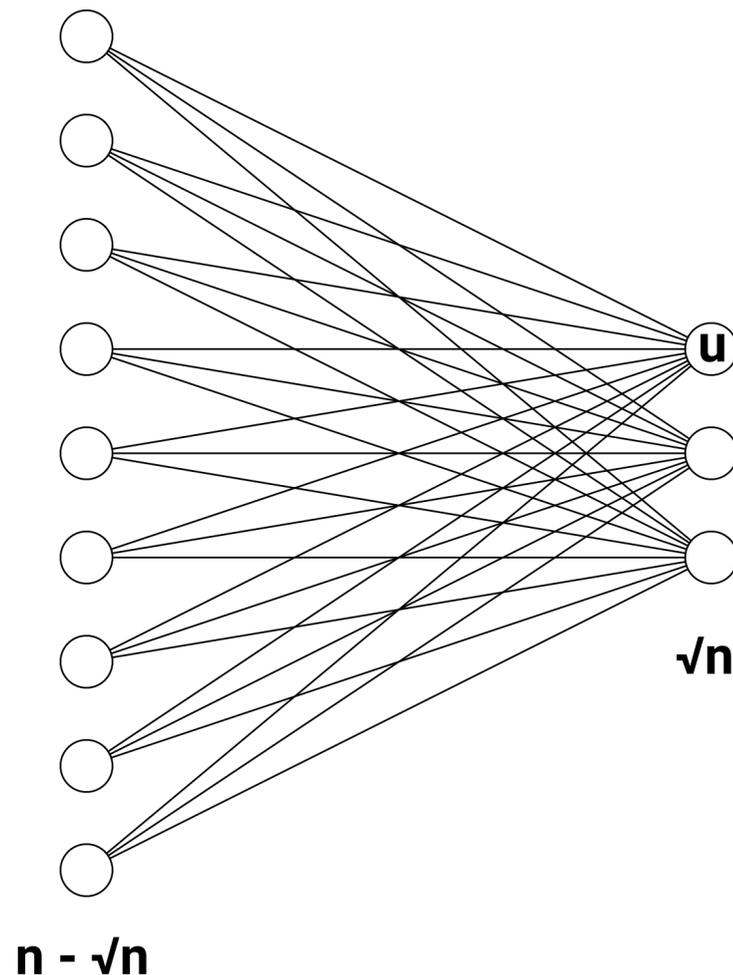


$$P(\mathbf{u} \text{ joins MIS}) \approx 1 / n$$

$$P(\mathbf{\text{some node of the right side joins MIS}}) \approx 1 / \sqrt{n}$$

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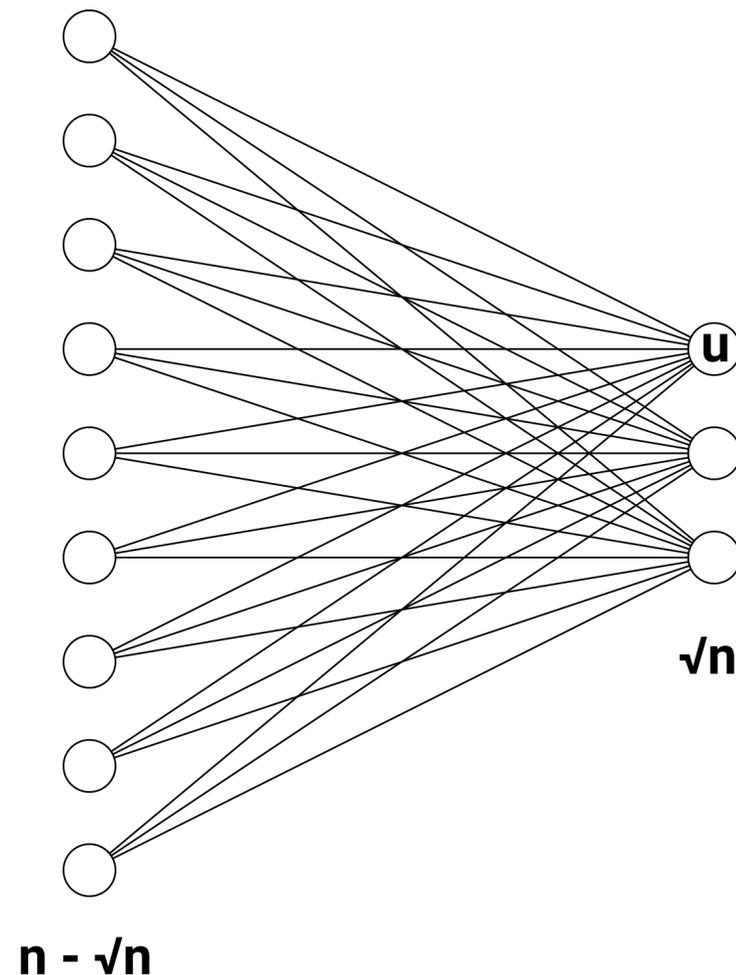
$$P(u \text{ joins MIS}) \approx 1/n$$

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with high probability, **no node** of the **right** side joins the MIS

Randomized MIS: Luby's Algorithm

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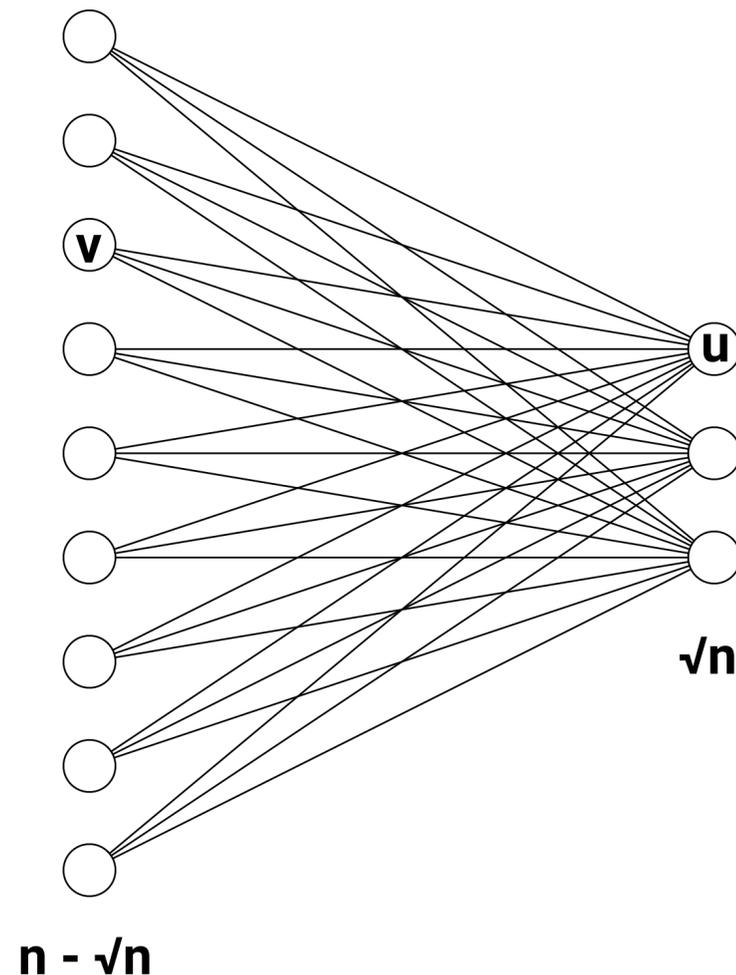
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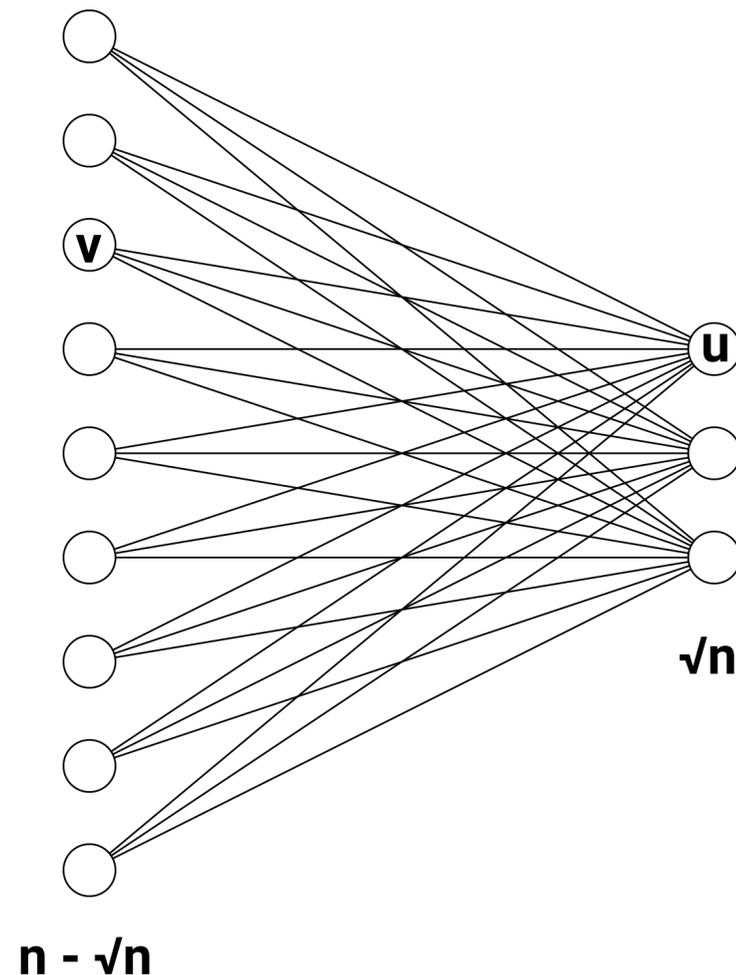
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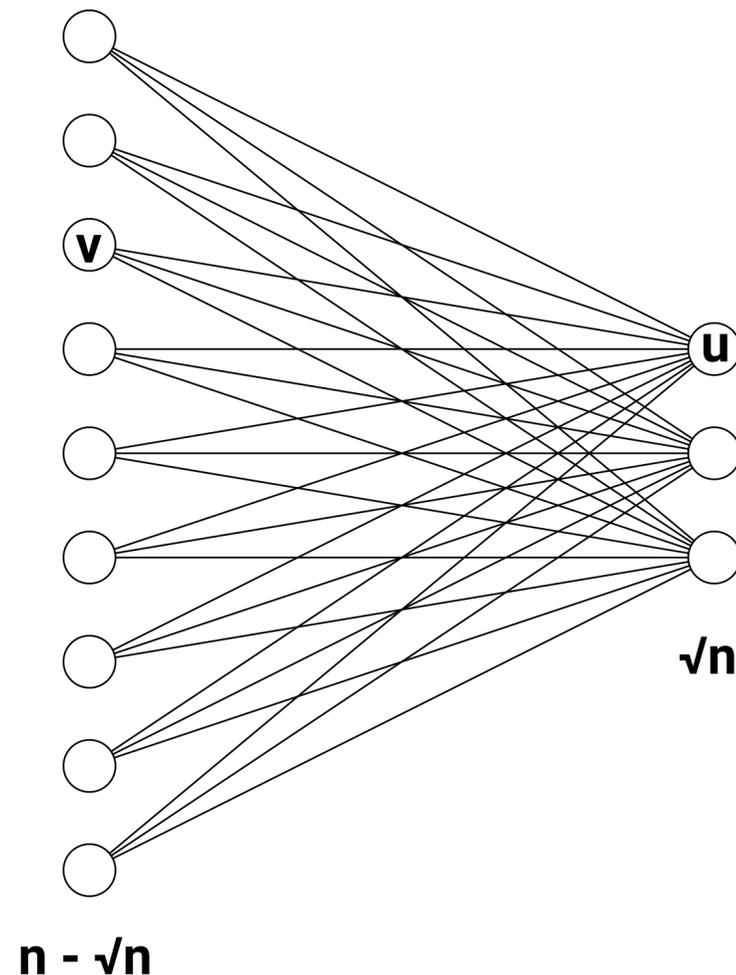
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only a $1/\sqrt{n}$ fraction of nodes join the MIS

Randomized MIS: Luby's Algorithm

Randomized MIS: Luby's Algorithm

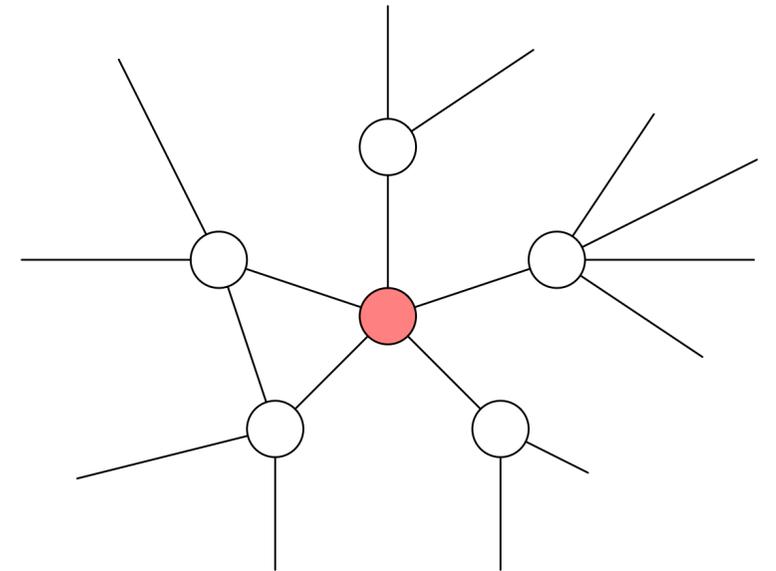
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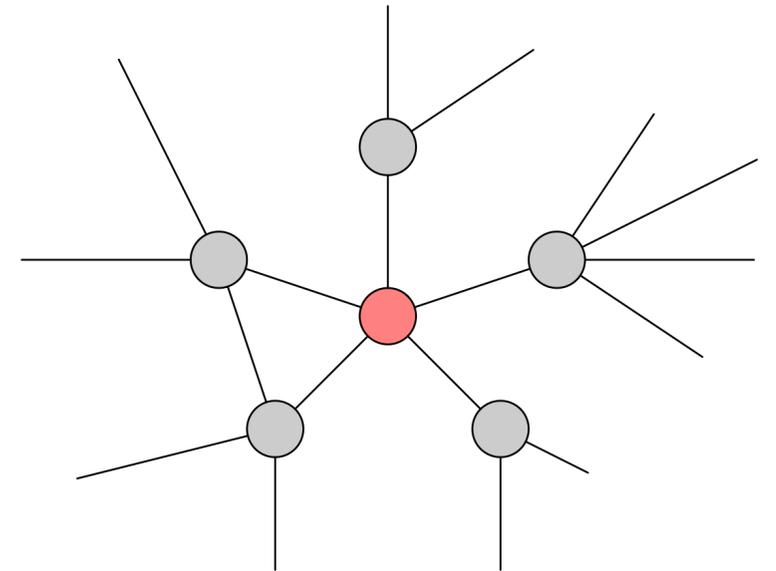
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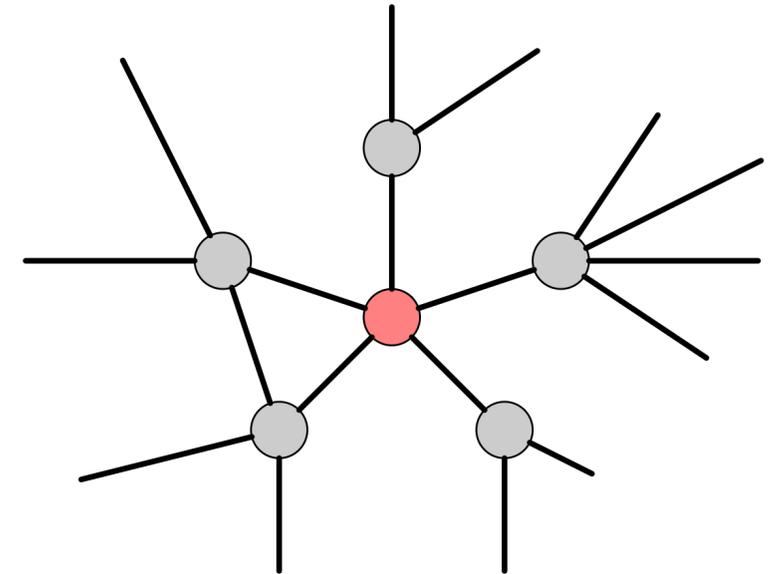
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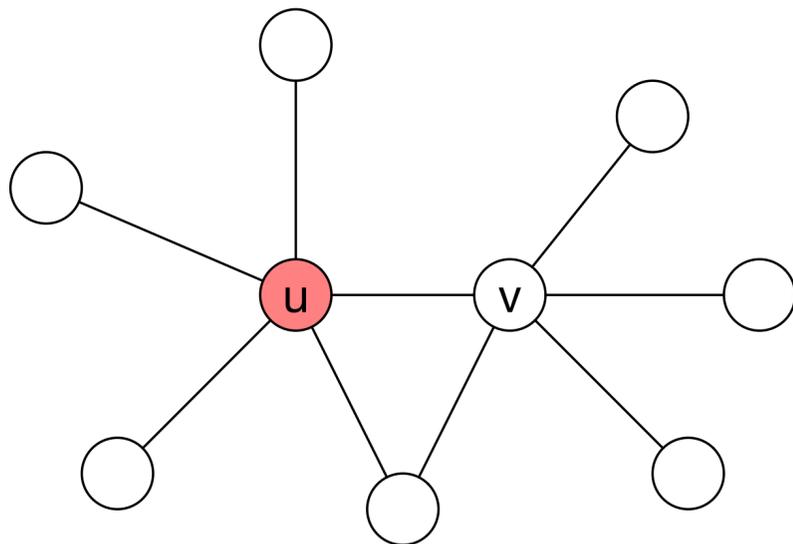
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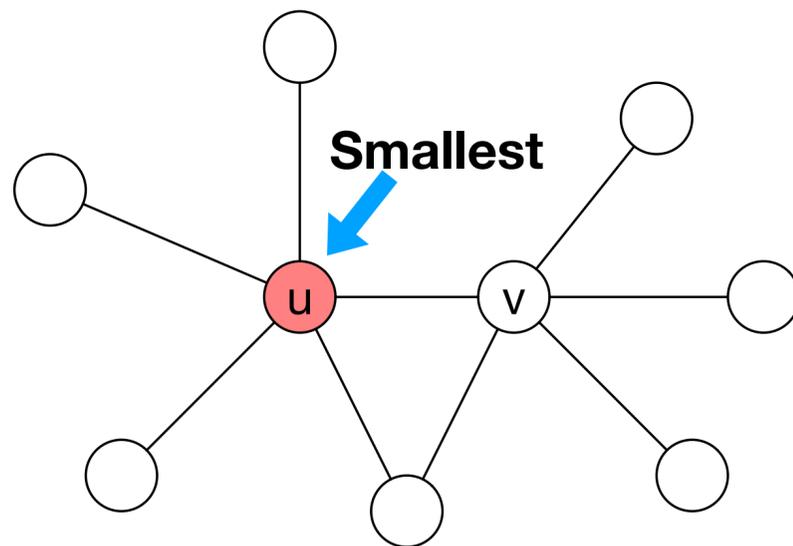
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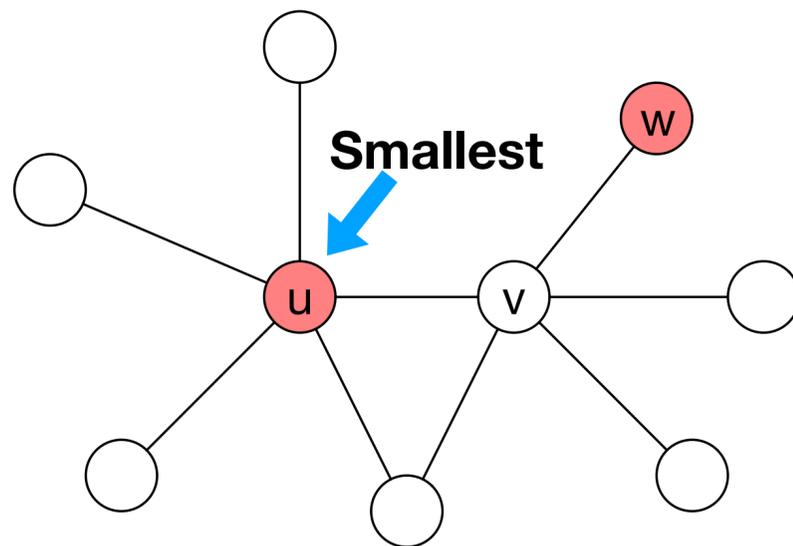
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$\mathcal{E}_{u,v}$ is true

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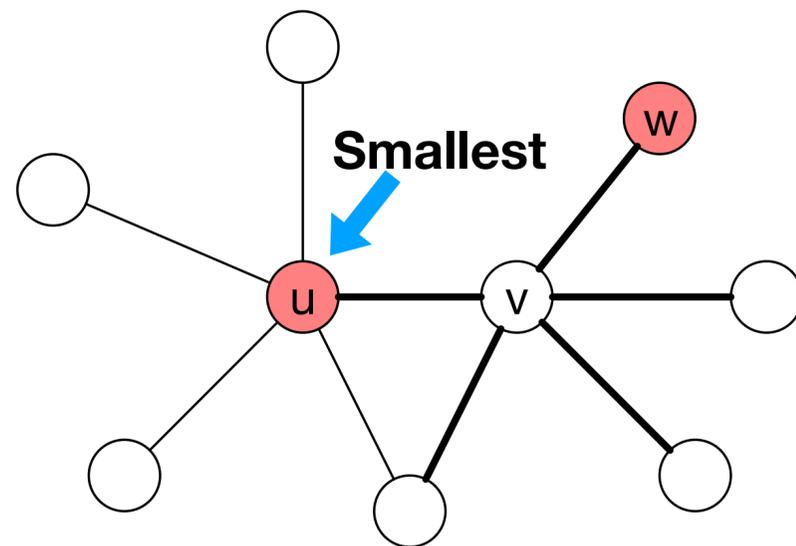


$\mathcal{E}_{u,v}$ is **true**

$\mathcal{E}_{w,v}$ is **false**

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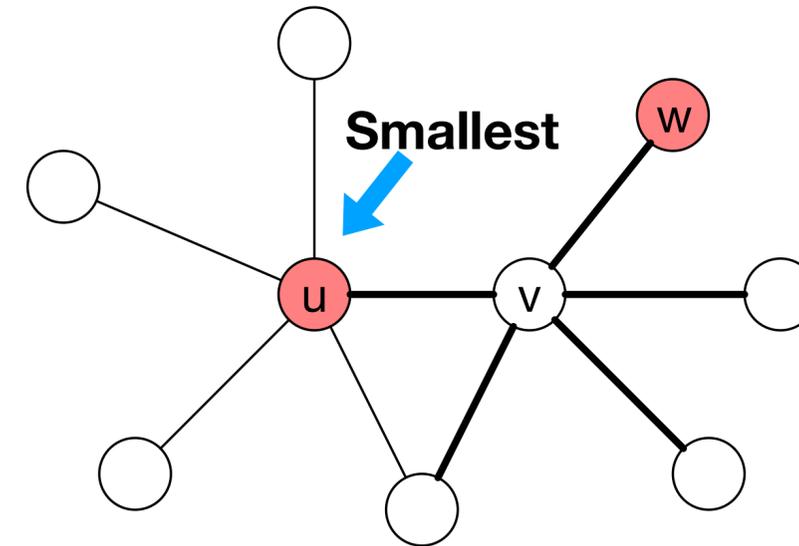
$\mathcal{E}_{u,v}$ is **true** \Rightarrow all edges of **v** get removed "**because of u**"

$\mathcal{E}_{w,v}$ is **false**

Randomized MIS: Luby's Algorithm

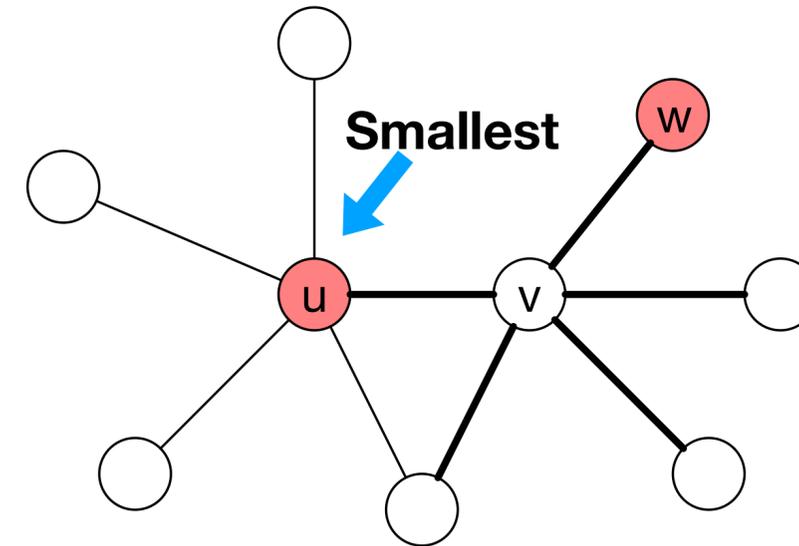
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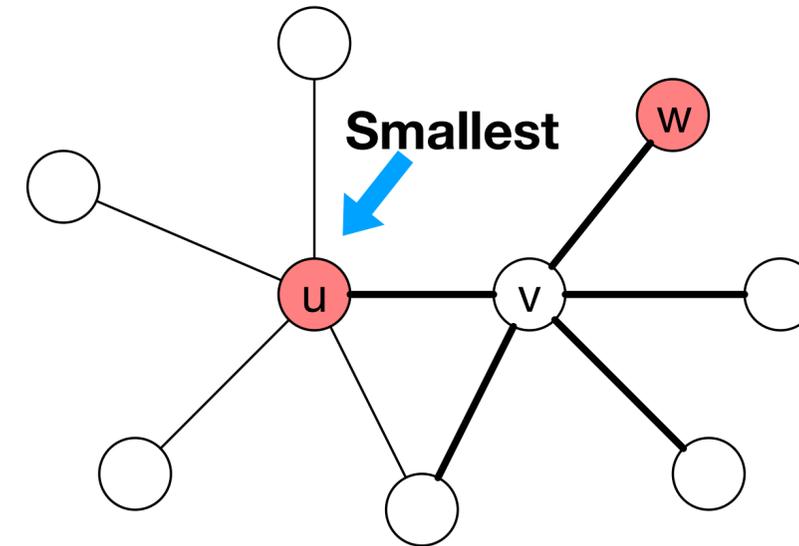
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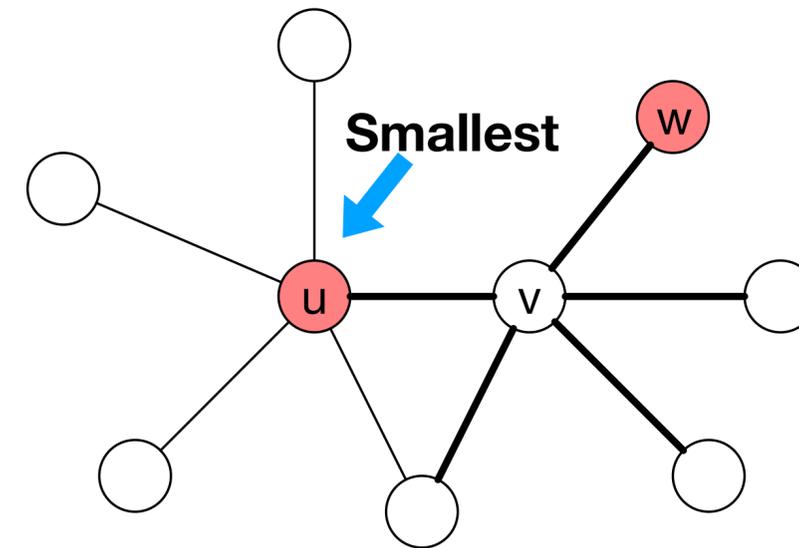
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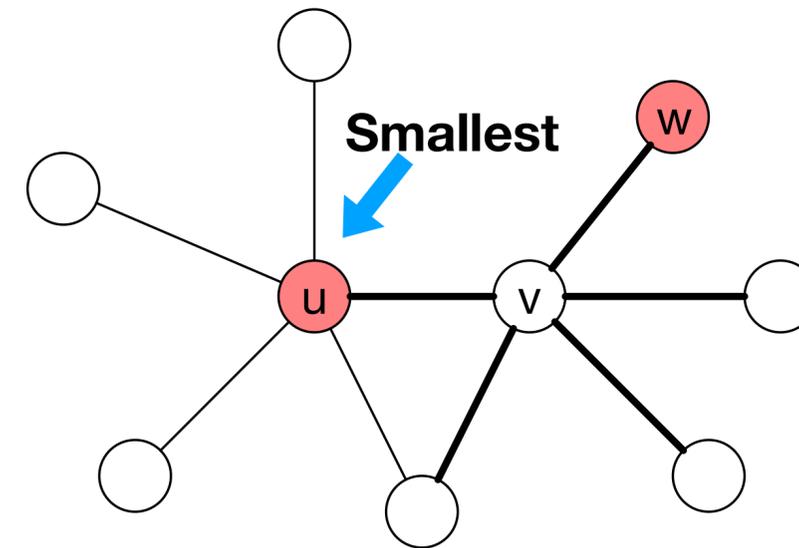
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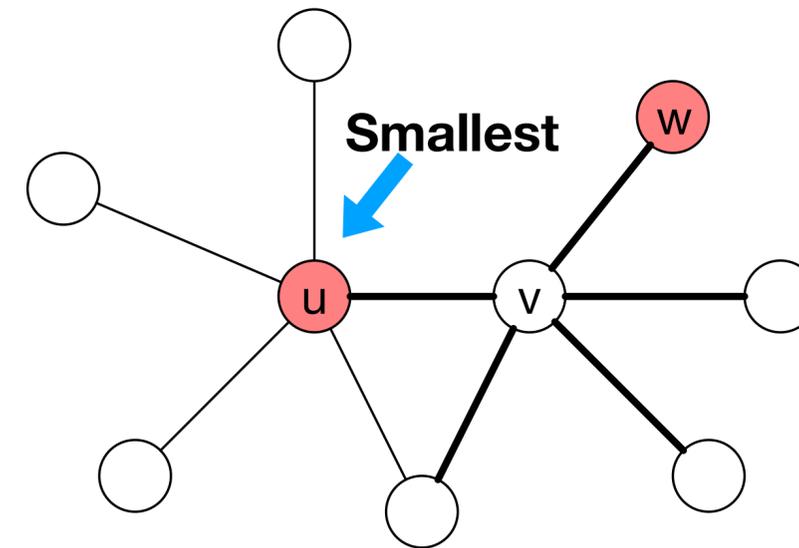
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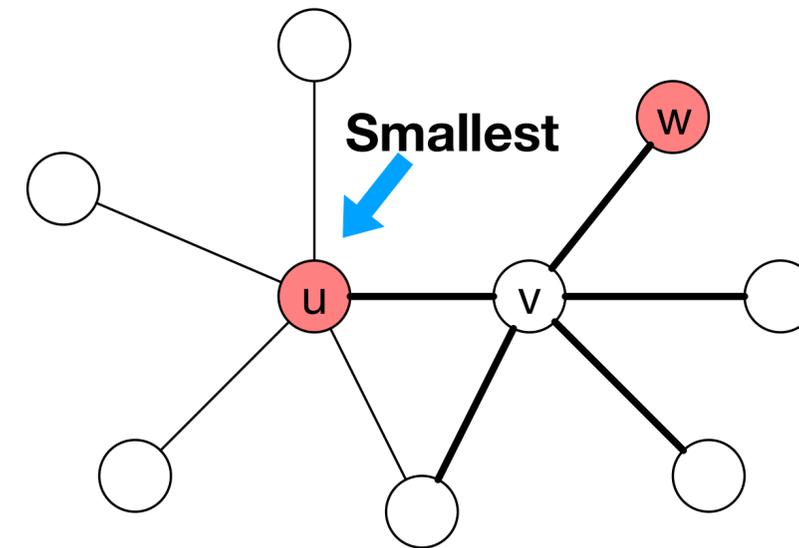
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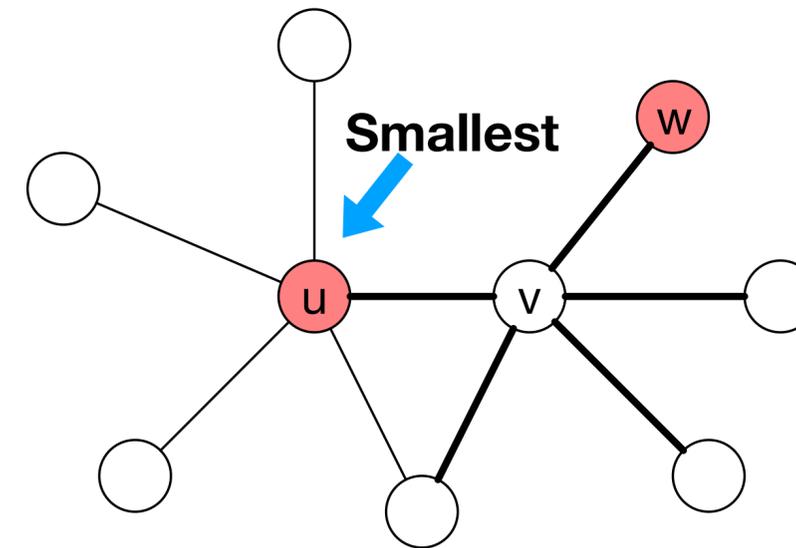
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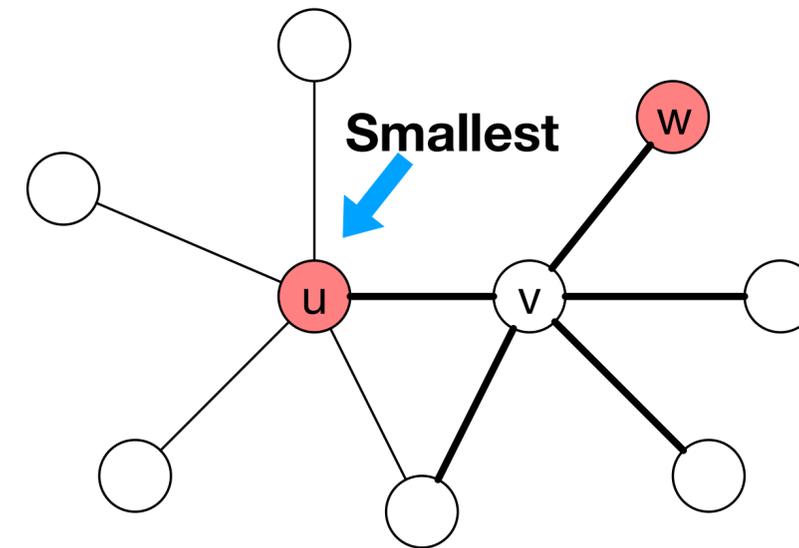
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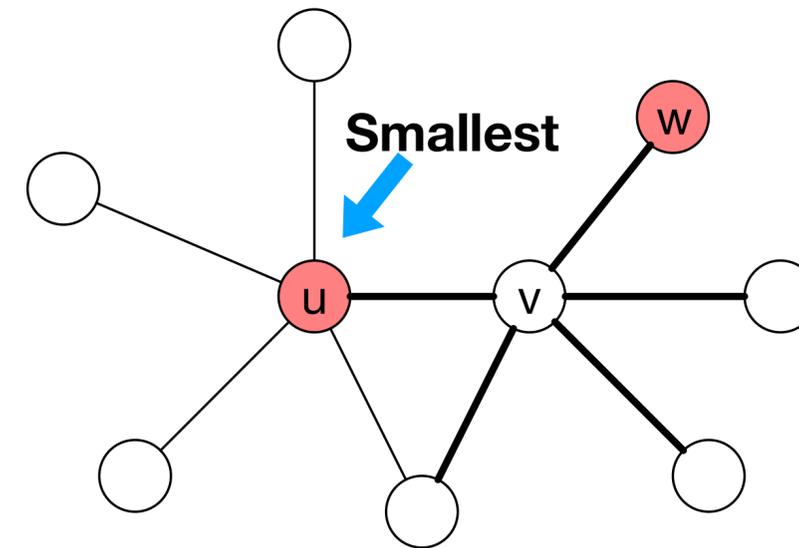
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- $\mathcal{E}_{w,v}$ is **false**
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- For each node **v**, at most one incident edges satisfies $\mathcal{E}_{u,v}$
- so every edge incident to **v** is **counted once, for v**

Randomized MIS: Luby's Algorithm

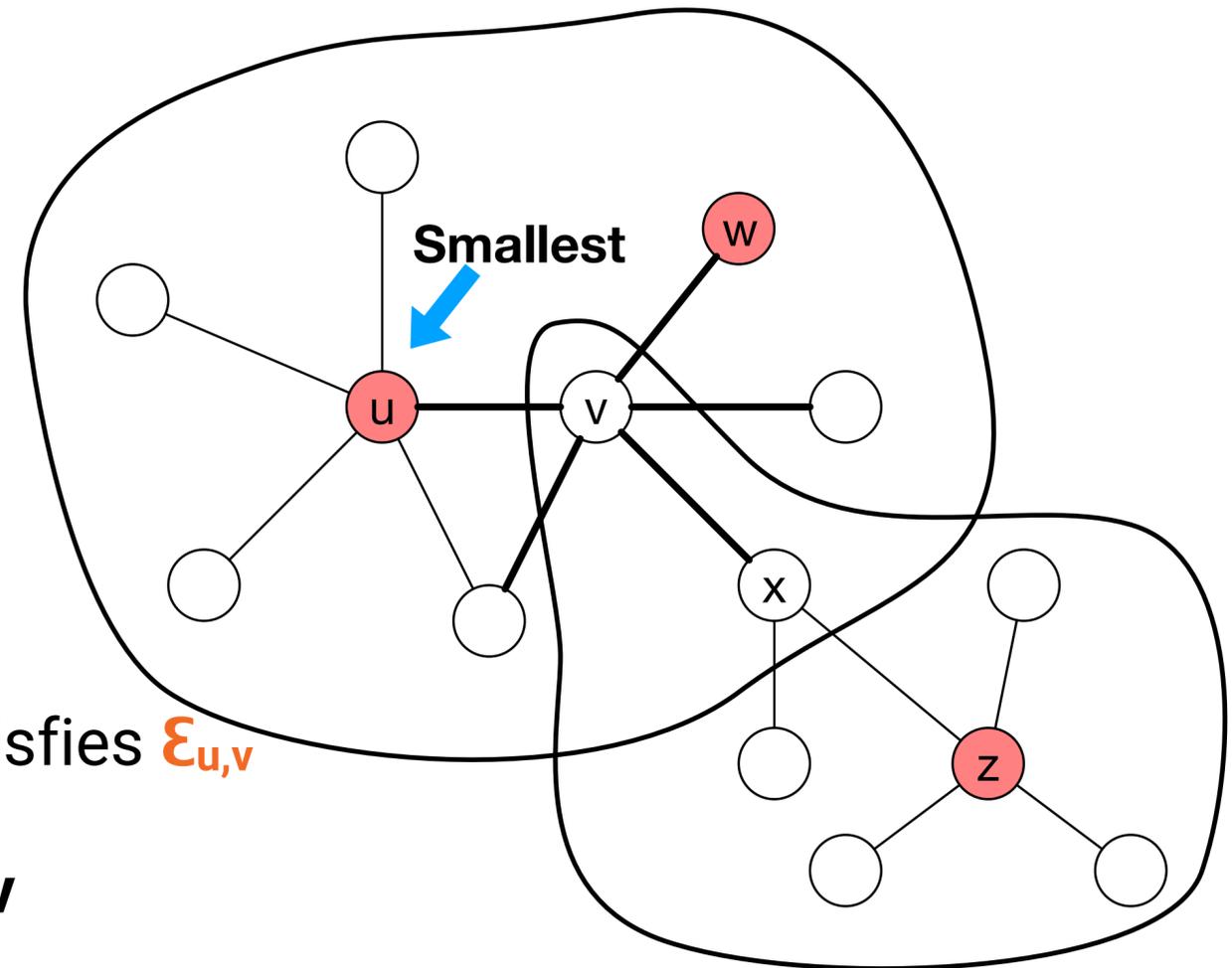
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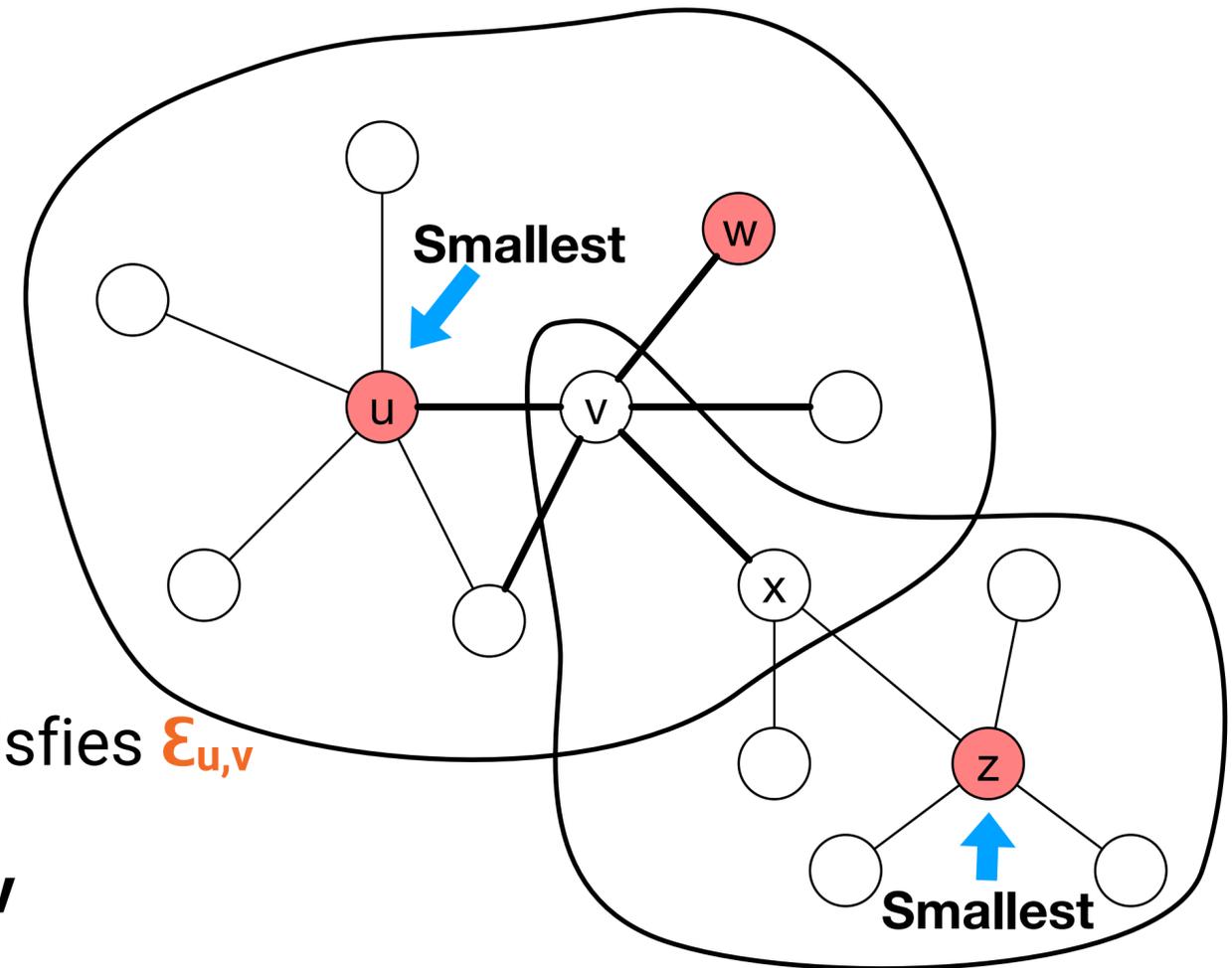
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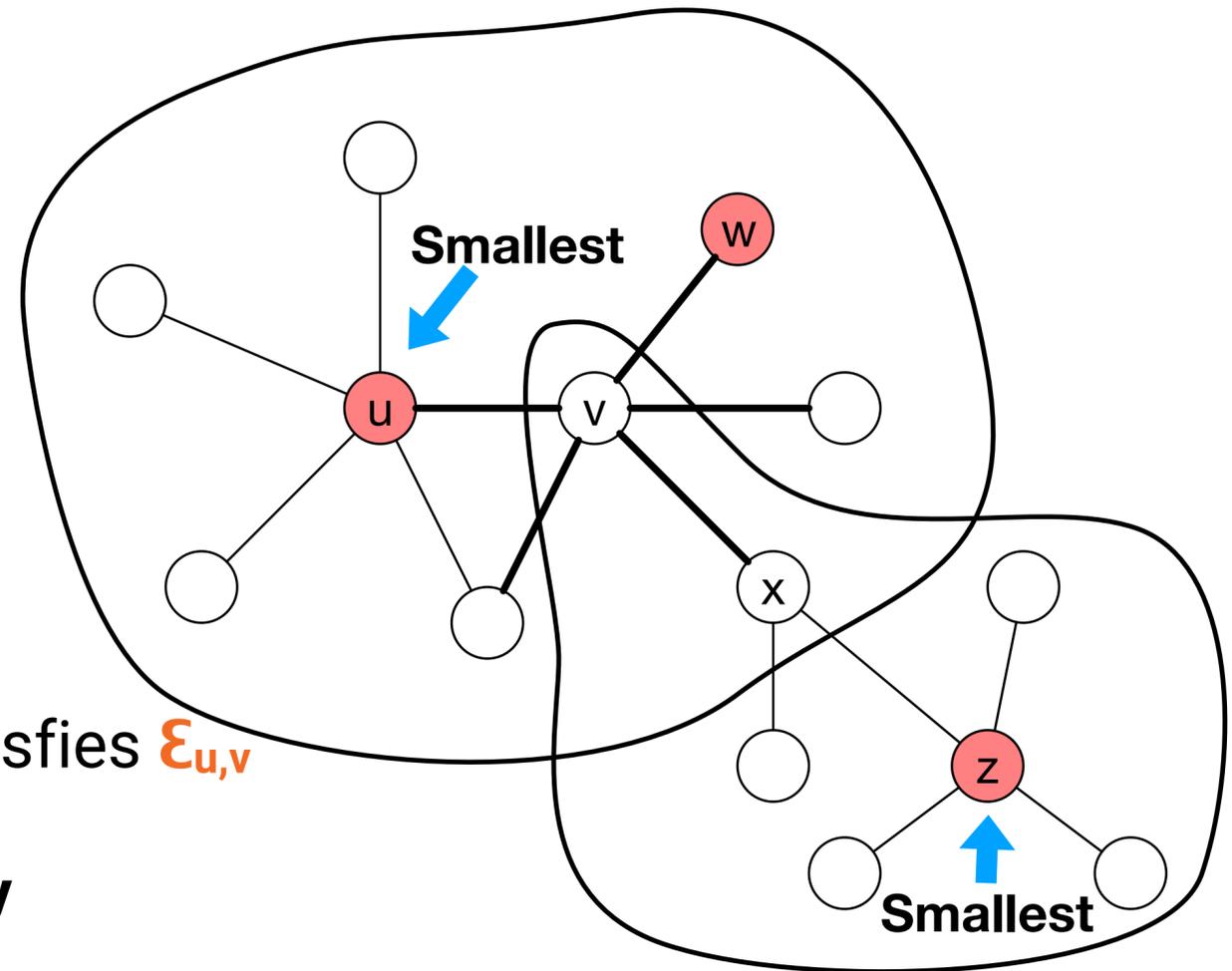
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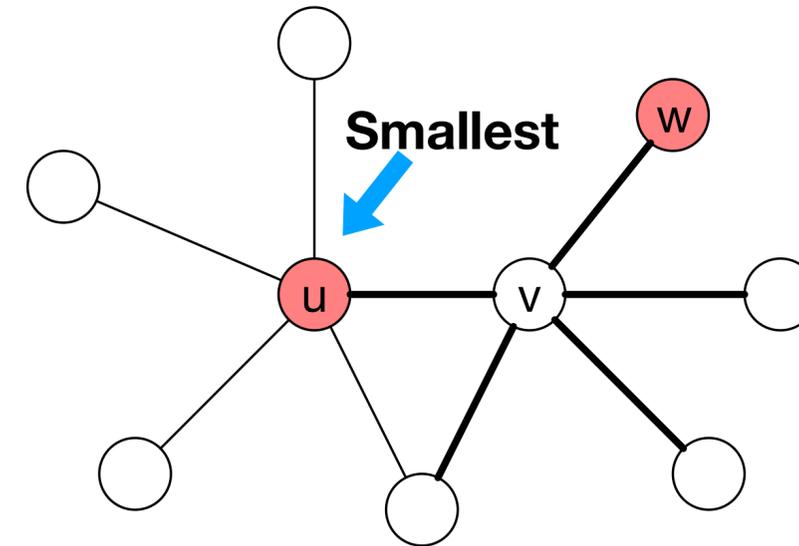
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- every edge can be **counted once**, for each endpoint



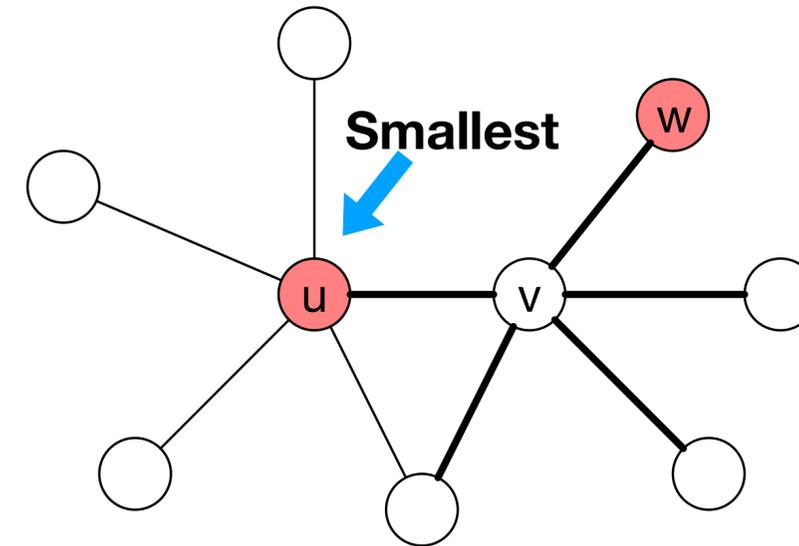
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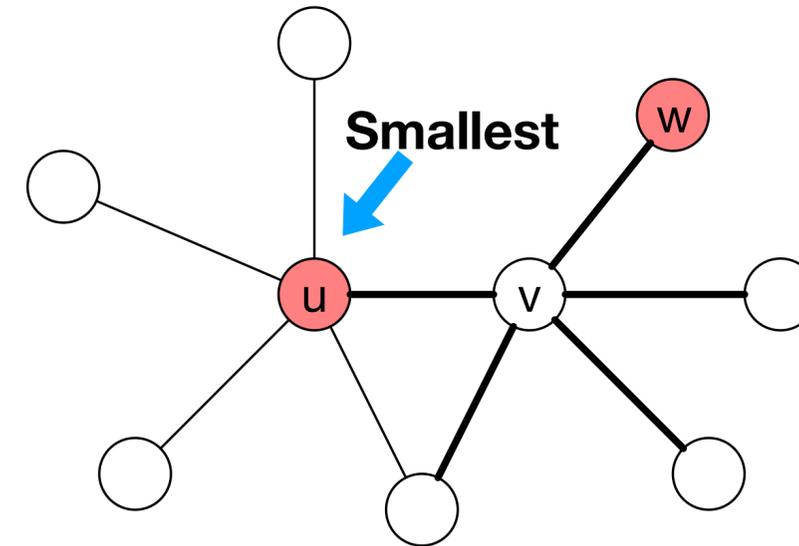
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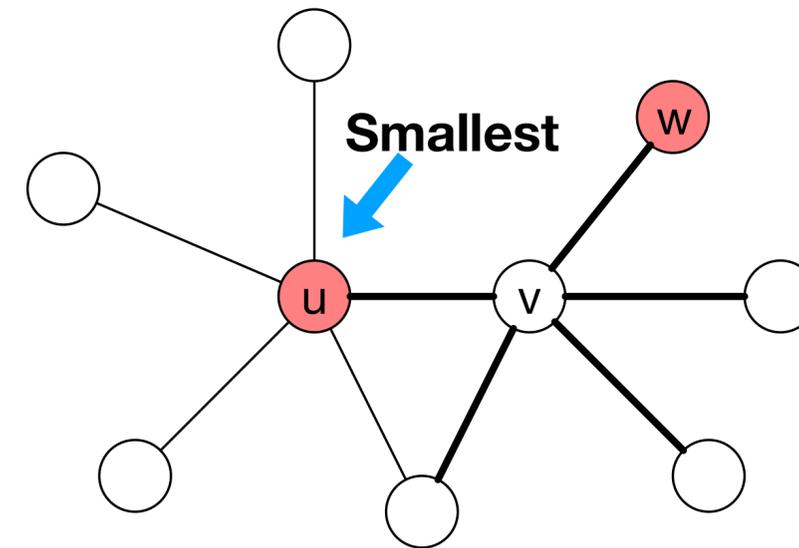
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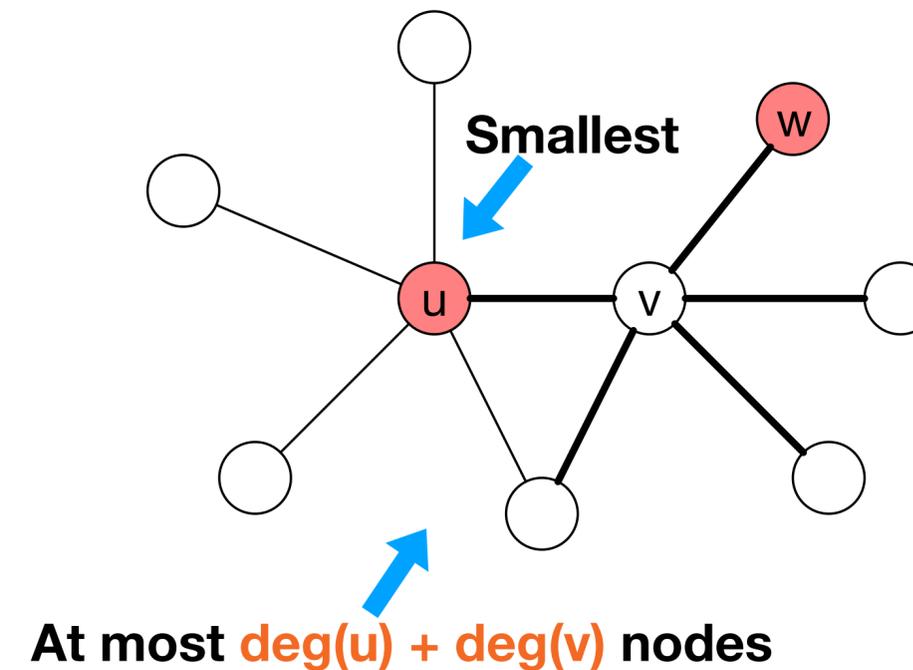
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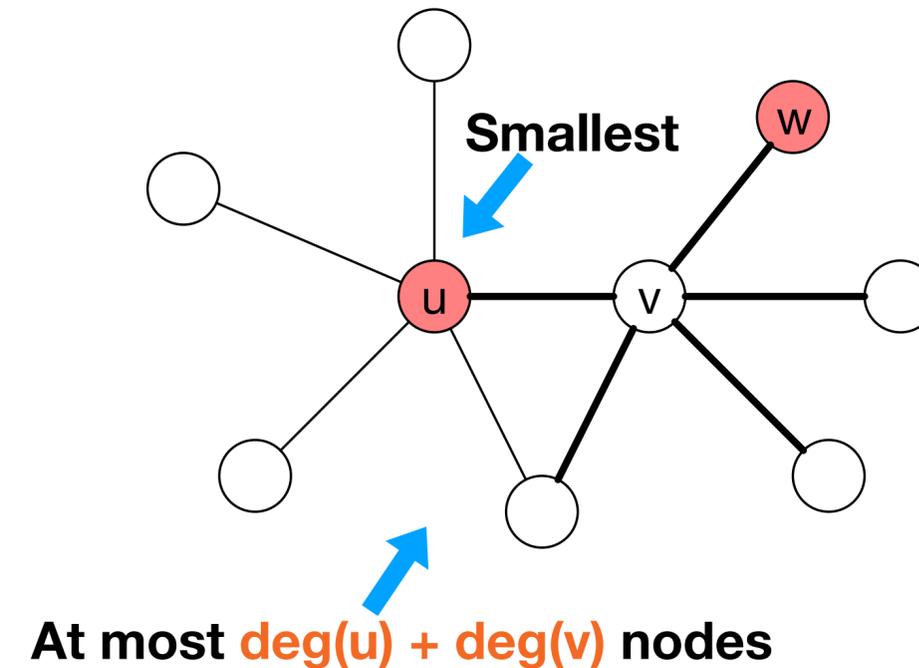
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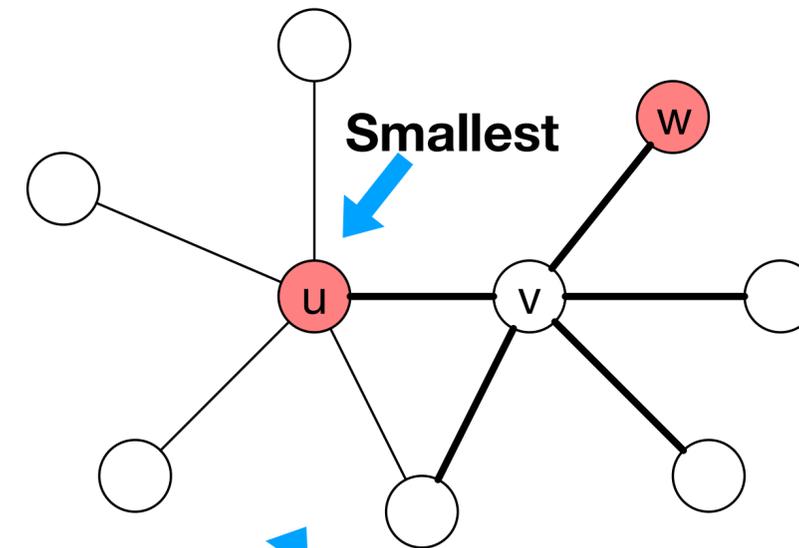
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• $\mathbb{E}[X_{u,v}] = \text{deg}(v) \cdot \mathbf{P}(\mathcal{E}_{u,v}) \geq \text{deg}(v) \cdot 1 / (\text{deg}(u) + \text{deg}(v)) = \text{deg}(v) / (\text{deg}(u) + \text{deg}(v))$



At most $\text{deg}(u) + \text{deg}(v)$ nodes

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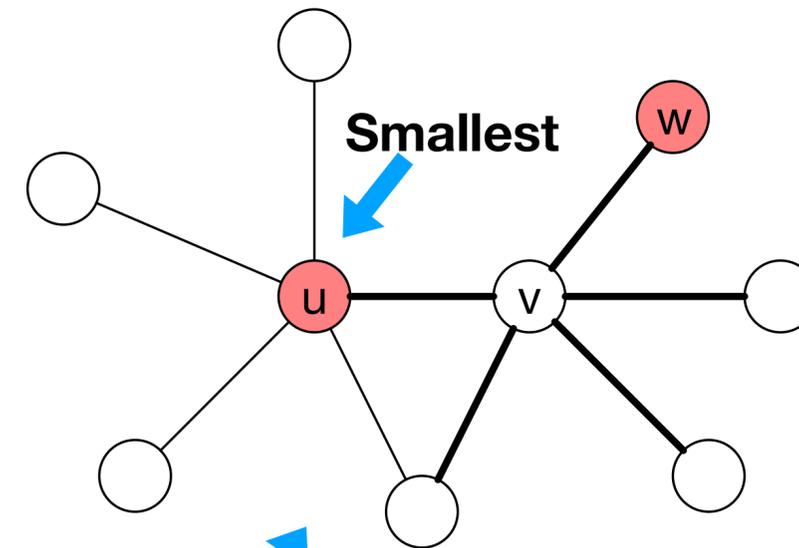
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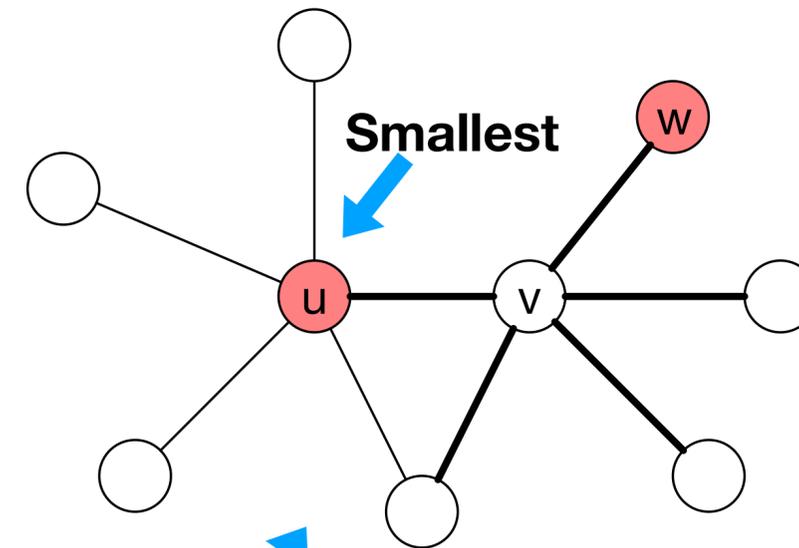
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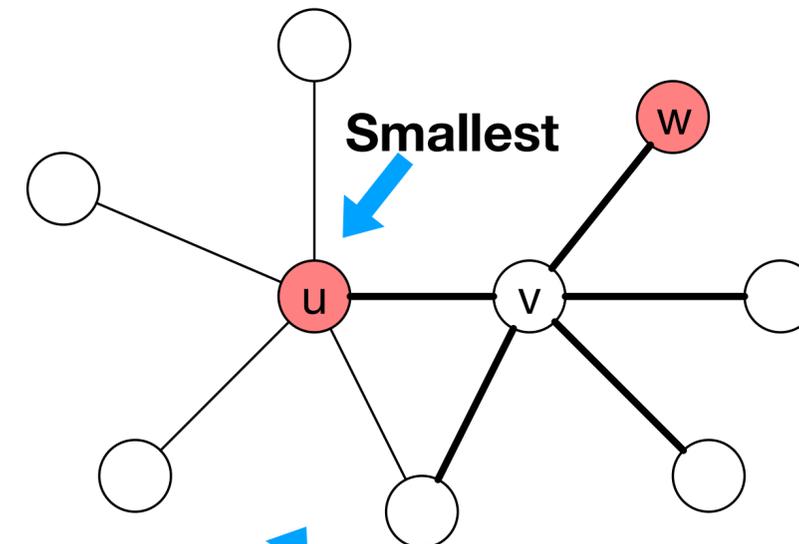
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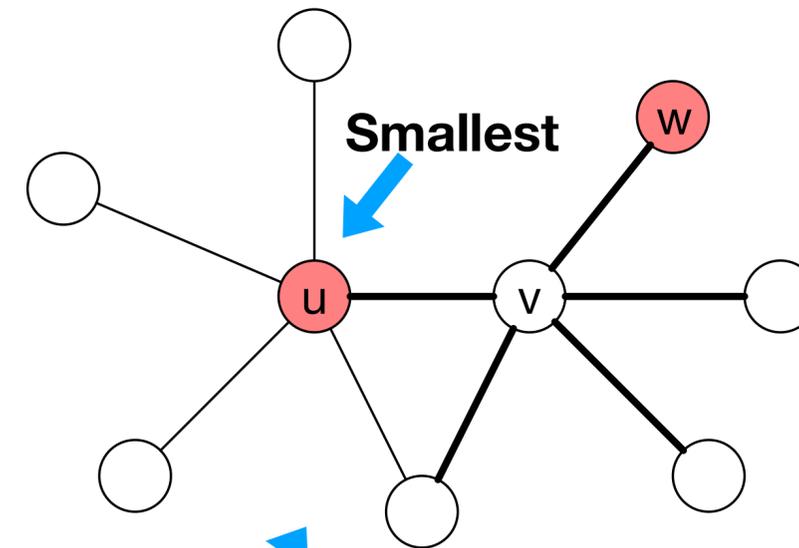
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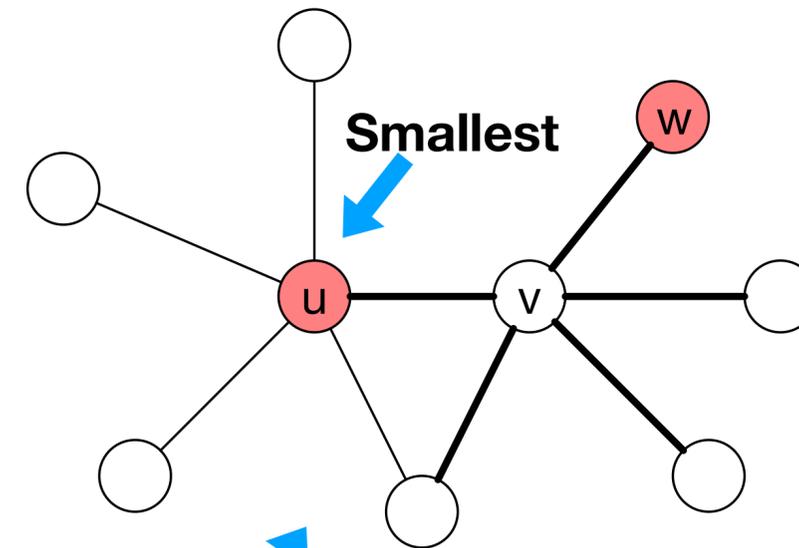
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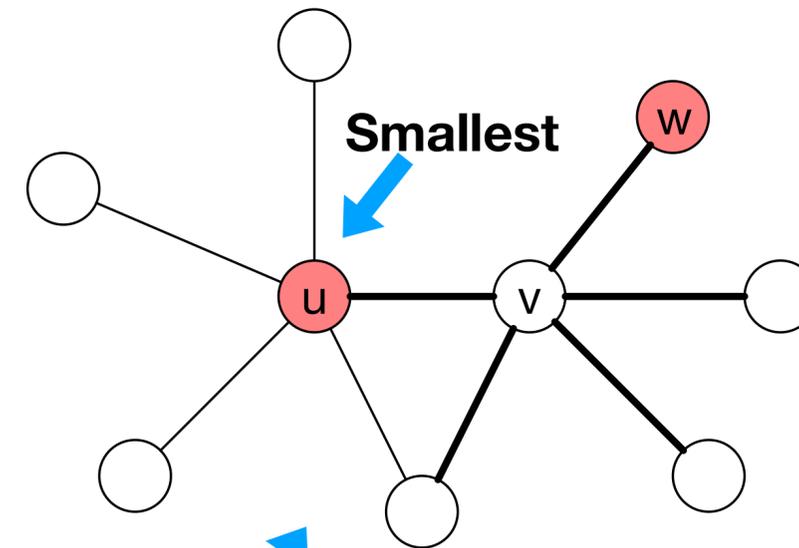
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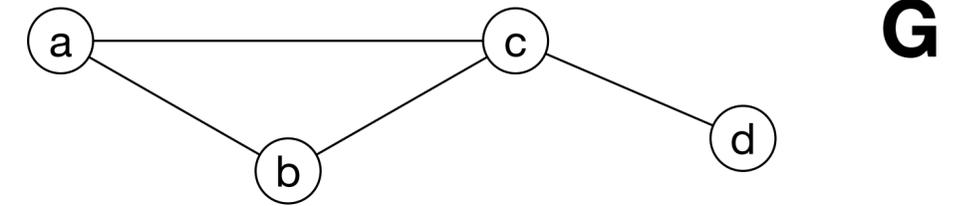
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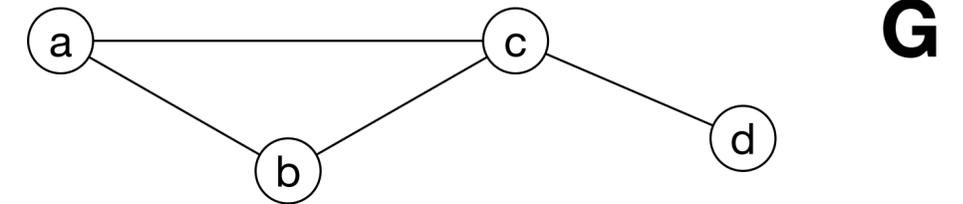
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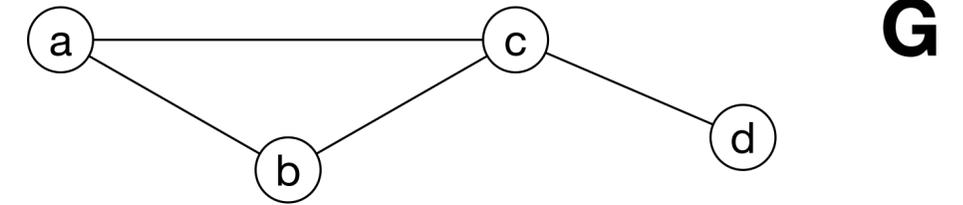
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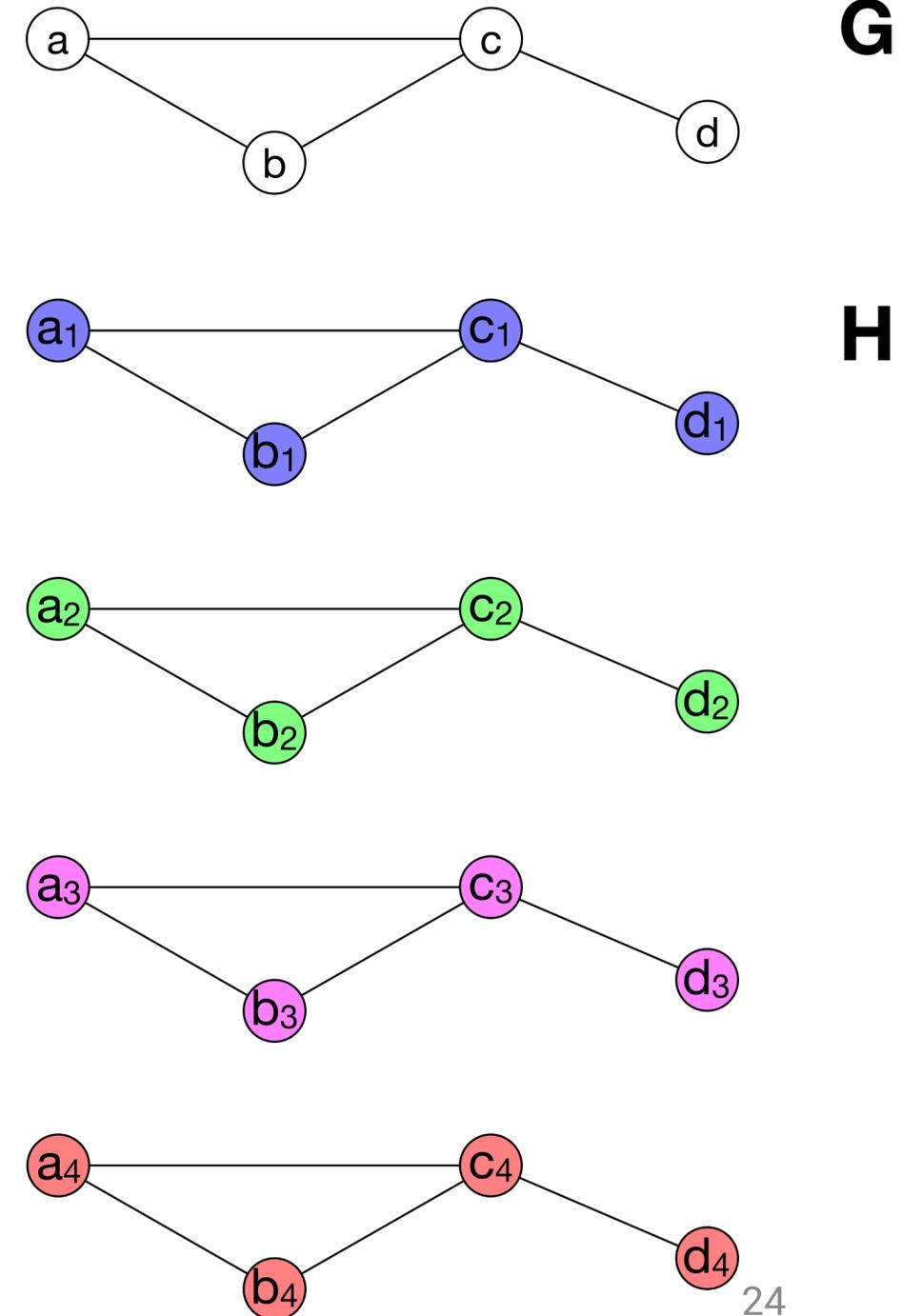
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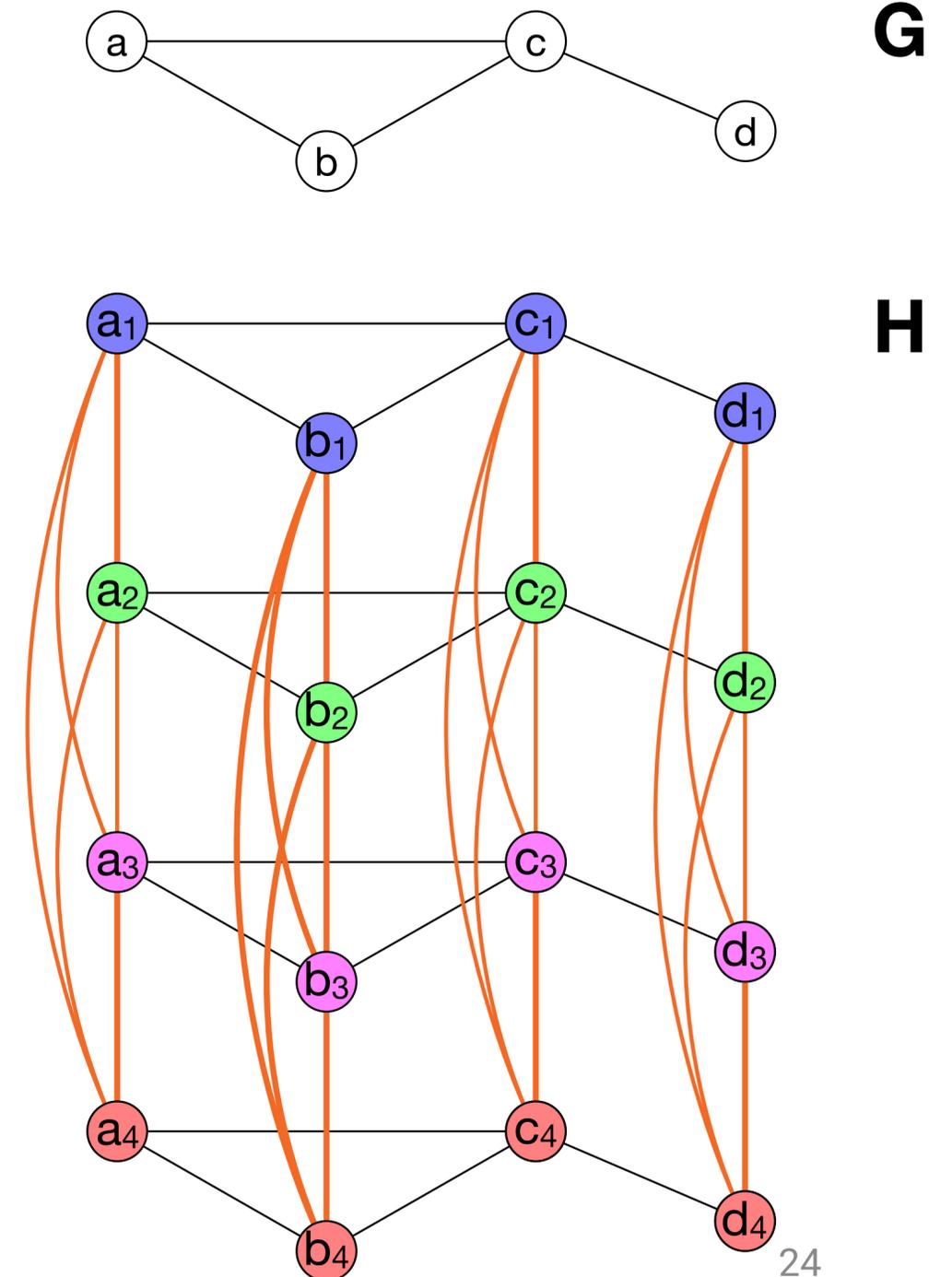
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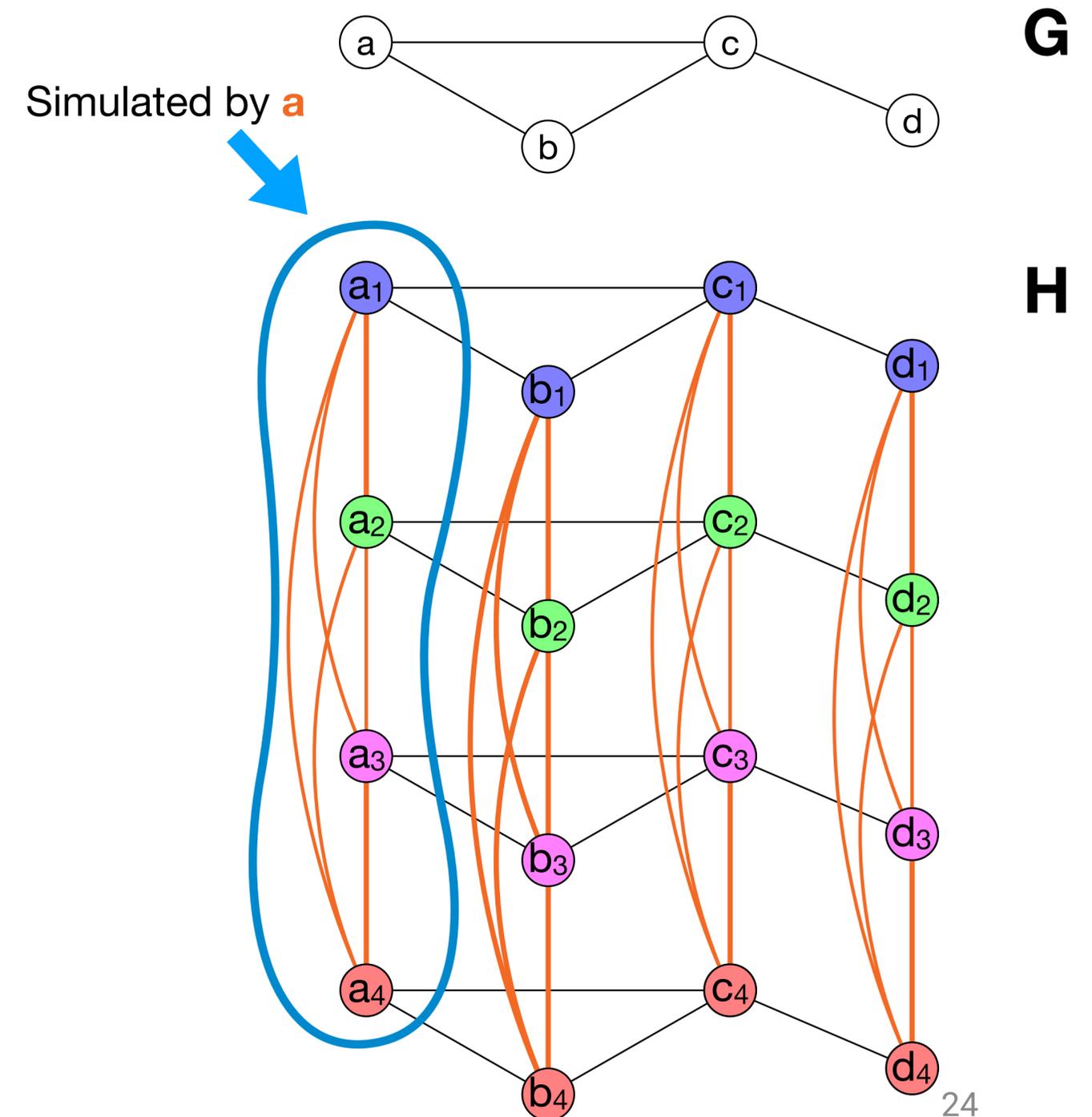
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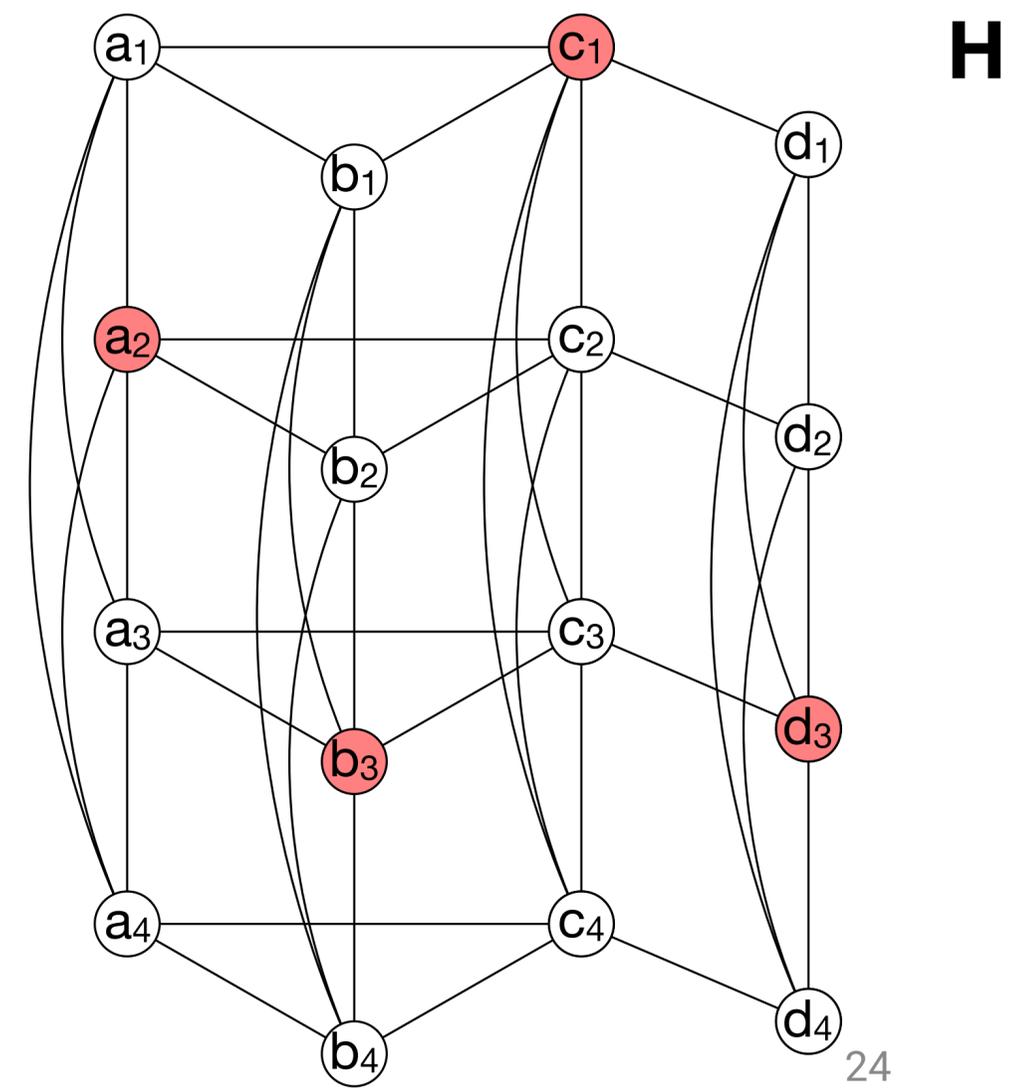
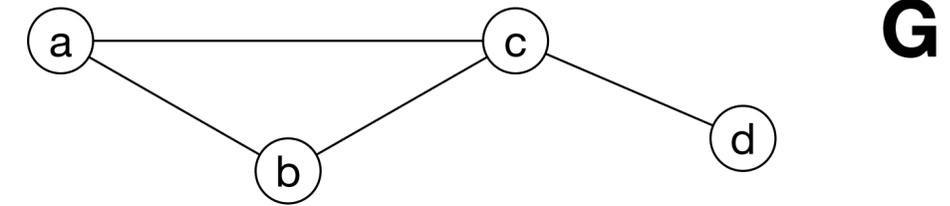
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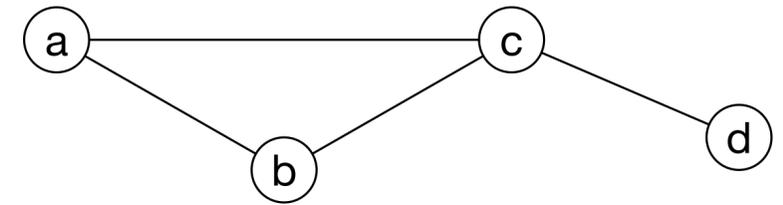


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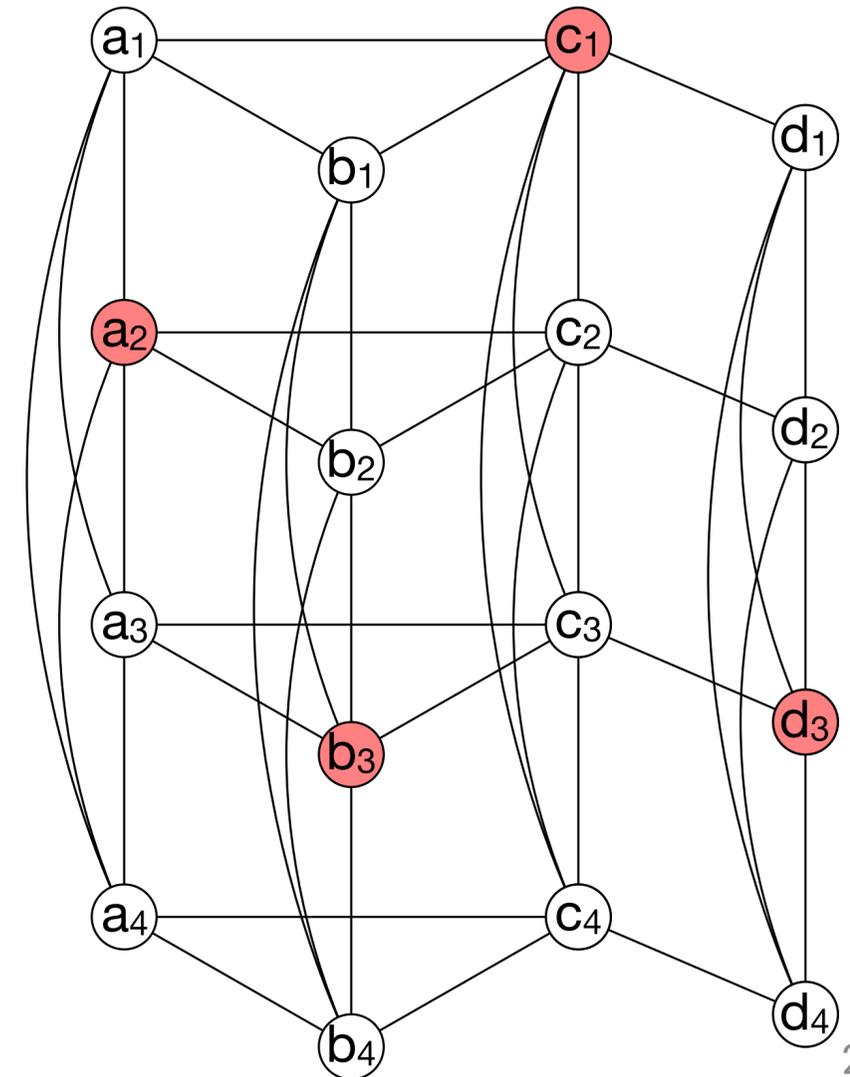
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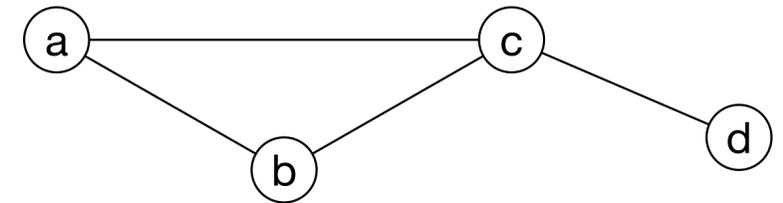
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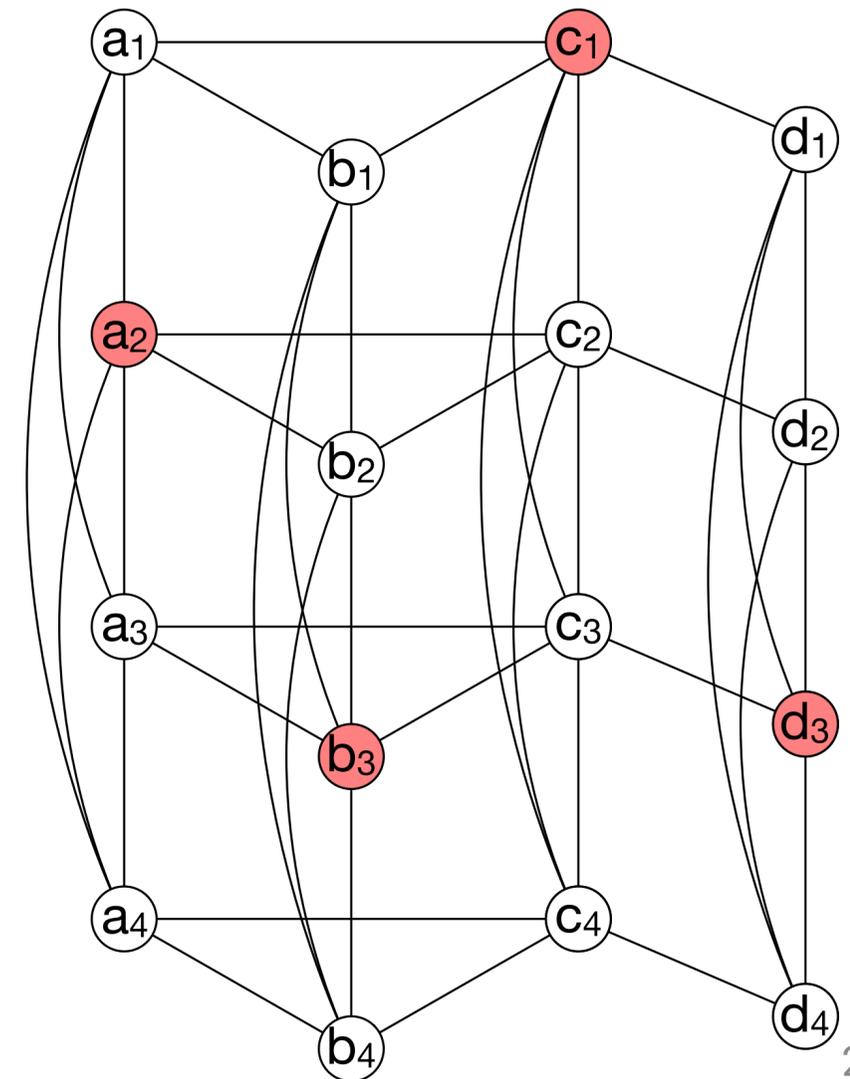
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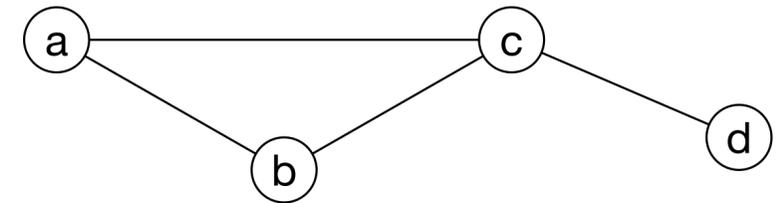
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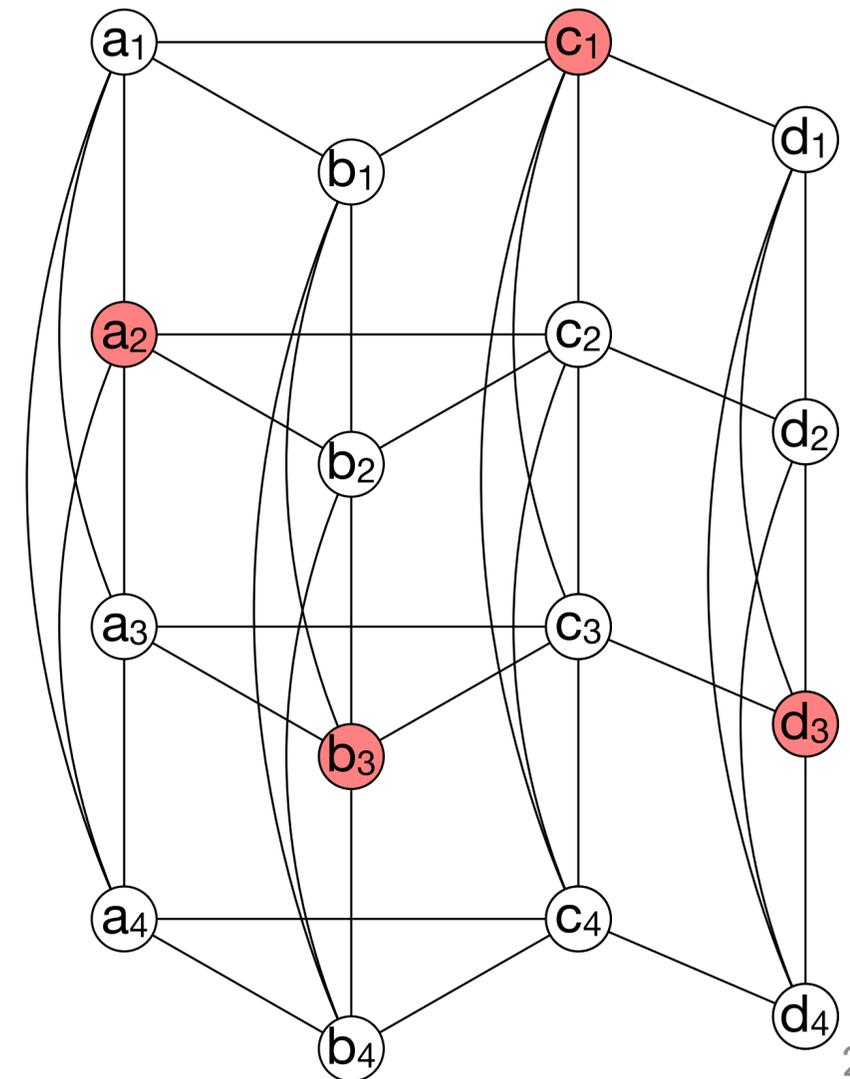
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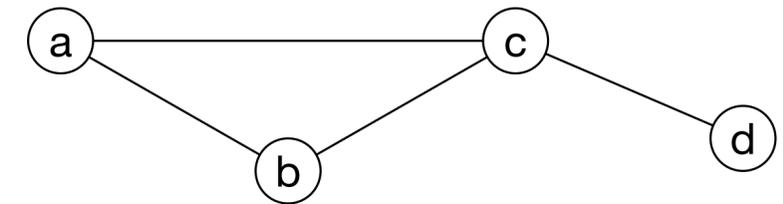
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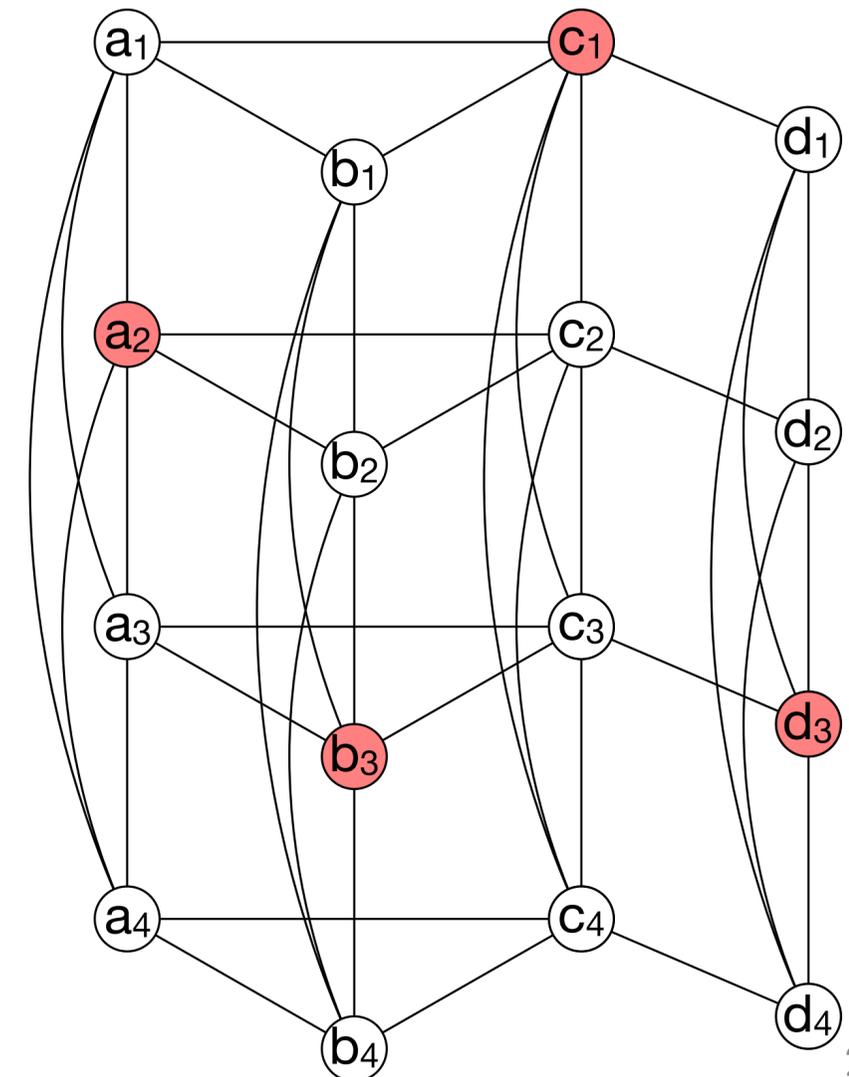
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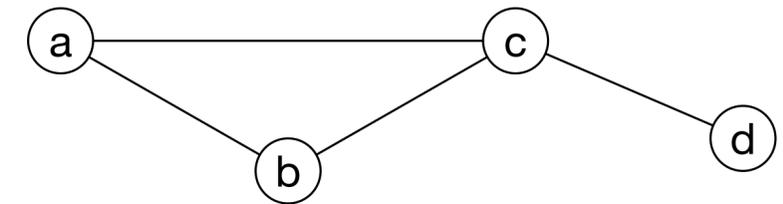
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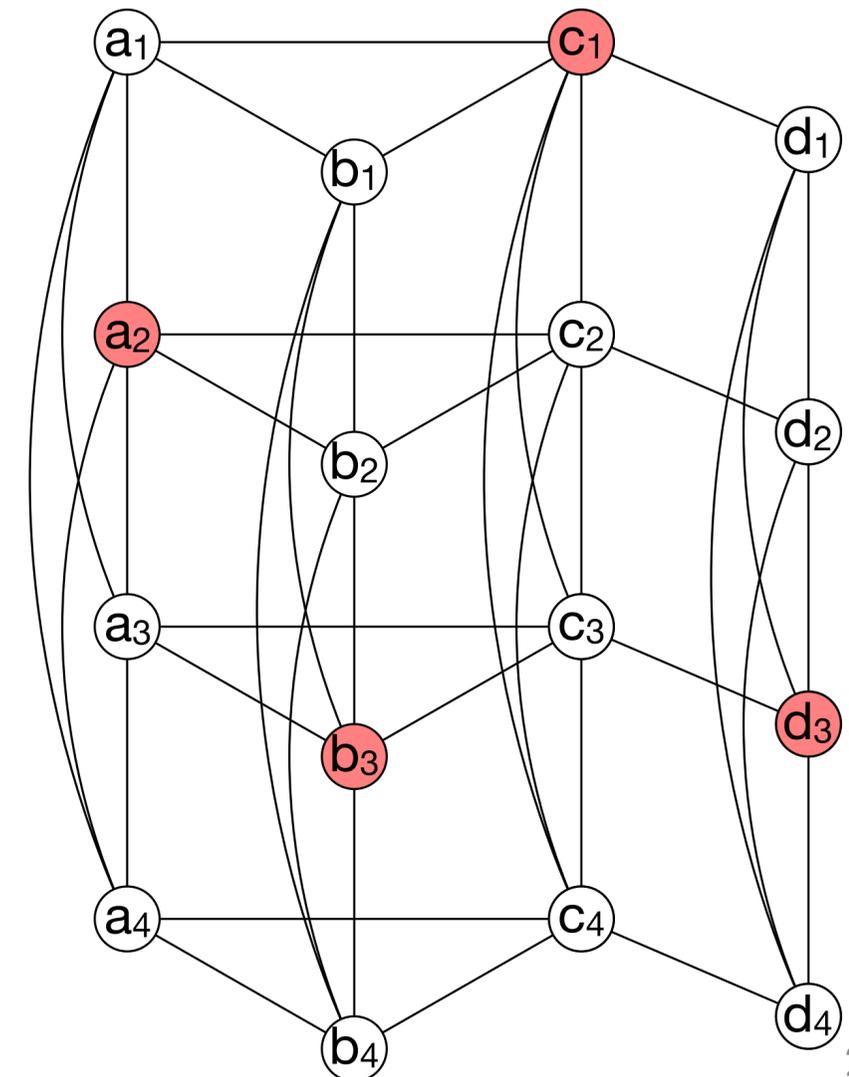
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Summary

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- ▶ Best deterministic algorithms:
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- ▶ Best randomized algorithms:
 - MIS: $O(\log \Delta + \log^c \log n)$ [Ghaffari '16]
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- ▶ Best lower bounds:
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 - $\Omega(\log n / \log \log n)$ [Balliu, Brandt, Kuhn, Olivetti '21] <- result from 3 weeks ago!