

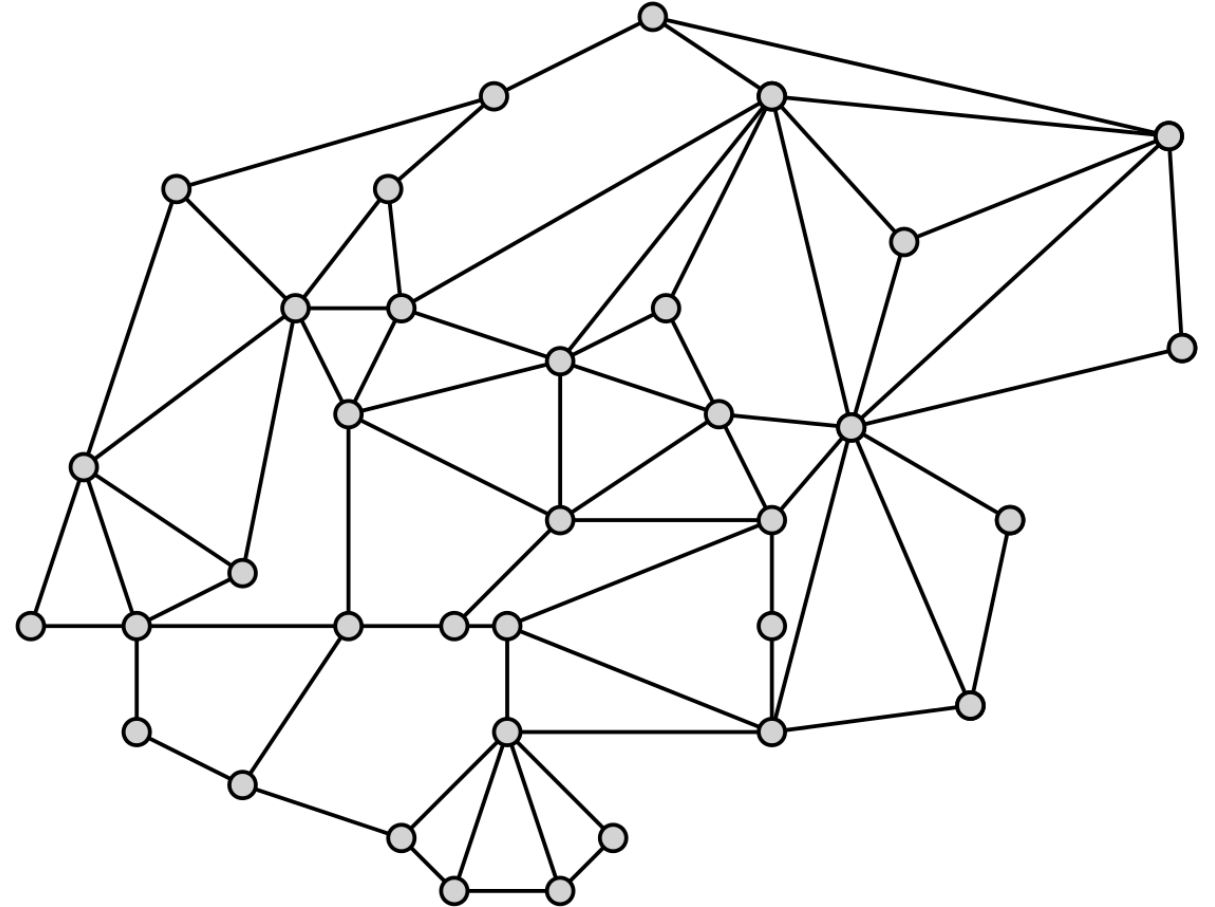
Sublogarithmic Distributed Algorithms for Lovász Local Lemma, and the Complexity Hierarchy

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LOCAL Model

Linial [FOCS'87]

- undirected graph $G = (V, E)$,
n nodes, maximum degree Δ
- synchronous message-passing rounds
- unbounded message size
- unbounded computation
- **Round Complexity:**
number of rounds to solve the problem



Chang, Pettie [FOCS'17]

Completeness of Lovász Local Lemma for Sublogarithmic Problems

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Any algorithm for an LCL problem on bounded-degree graphs with round complexity $o(\log n)$ can be **automatically sped up** to run in $O(T_{LLL}(n))$ rounds.

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$T_{LLL}(n)$ = round complexity of Lovász Local Lemma
on n events/nodes
under $pd^c < 1$ for some constant c ,
for $d = O(1)$

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Naor, Stockmeyer '95

Locally Checkable Labeling (LCL)

solution checkable in $O(1)$ rounds

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Lovász Local Lemma (LLL)

Erdős, Lovász '75

Lovász Local Lemma (LLL) *Erdős, Lovász '75*

independent variables \mathcal{V} (w.l.o.g. fair coins)

Not too likely bad events:

n bad events \mathcal{X} with $vbl(A) \subseteq \mathcal{V}$ for all $A \in \mathcal{X}$

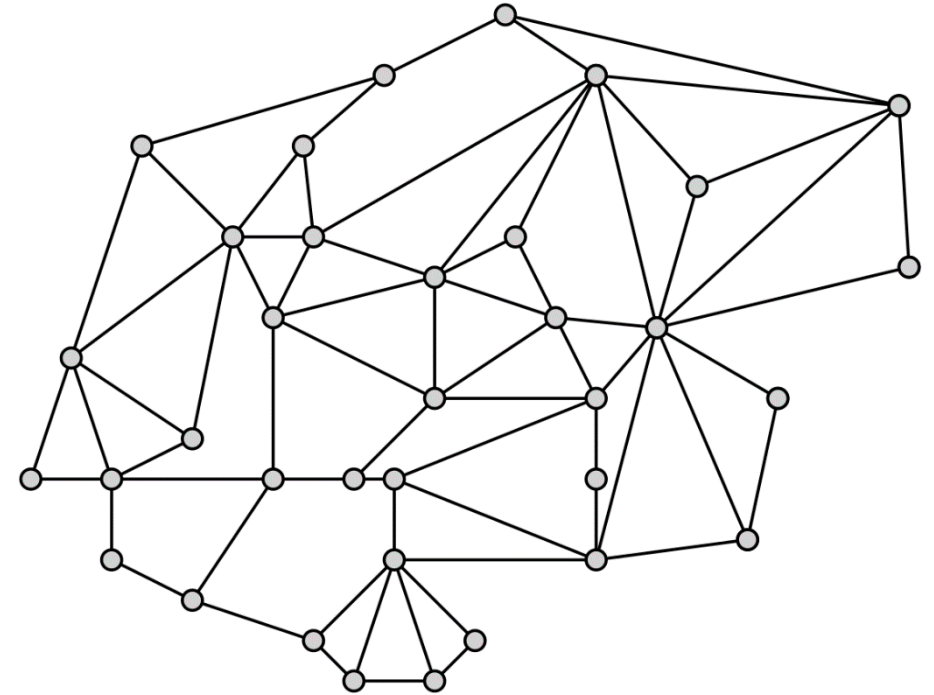
$\Pr[A] \leq p$ for all $A \in \mathcal{X}$

Not too many dependencies:

dependency graph $G = (\mathcal{X}, E)$

$E = \{(A, B) : vbl(A) \cap vbl(B) \neq \emptyset\}$

maximum degree d



If local union bound (with some slack) is satisfied, then all bad events can be avoided!

If $epd \leq 1$, then $\Pr[\bigcap_{A \in \mathcal{X}} \bar{A}] > 0$.

Lovász Local Lemma (LLL) *Erdős, Lovász '75*

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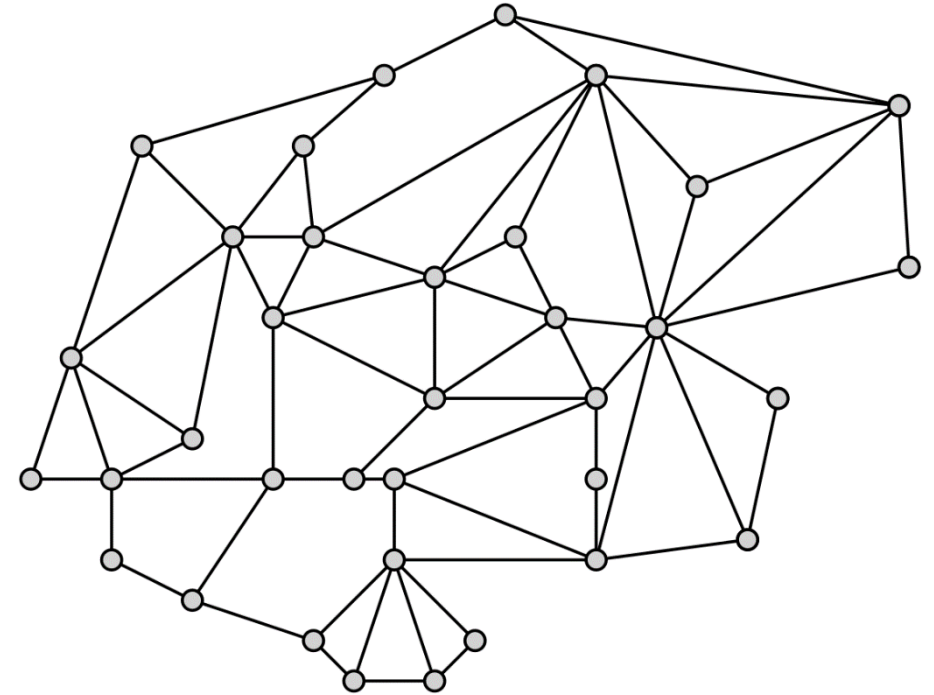
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standard criterion

LOCAL Complexity of the Lovász Local Lemma

$O(\log^2 n)$
for $epd \leq 1$ (STANDARD)

Moser, Tardos [JACM'10]

$O(\log n)$
for $epd^2 < 1$ (POLYNOMIAL)

Chung, Pettie, Su [PODC'14]

$O(\log n / \log \log n)$
for $p2^d < 1$ (EXPONENTIAL)

Chung, Pettie, Su [PODC'14]

CONJECTURE:
 $T_{LLL}(n) = O(\log \log n)$

Chang, Pettie [FOCS'16]

$\Omega(\log \log n)$
for $p2^d < 1$ (EXPONENTIAL)

Brandt et al. [STOC'16]

$T_{LLL}(n)$ = round complexity of Lovász Local Lemma
on n events/nodes
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Our Results

$$T_{LLL}(n) = 2^{O(\sqrt{\log \log n})}$$

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Automatic
Speedup



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Gap in Distributed Complexity Hierarchy
for LCLs on bounded-degree graphs

$$o(\log n) \rightarrow 2^{O(\sqrt{\log \log n})}$$

$$o(\log \log n) \rightarrow O(\log^* n)$$

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Other Applications

$2^{O(\sqrt{\log \log n})}$ - round Graph Coloring Algorithms

- f -defective $O\left(\frac{\Delta}{f}\right)$ -coloring
- β -frugal $120 \Delta^{1+\frac{1}{\beta}}$ -coloring
- List-vertex-coloring

Previously best known $O(\log n)$ by *Chung, Pettie, Su* [PODC'14]

Algorithm for Lovász Local Lemma

BASE ALGORITHM

$O(d^2) + \lambda \cdot \log^{\frac{1}{\lambda}} n \cdot 2^{O(\sqrt{\log \log n})}$ rounds
under $p(e \cdot d)^{4\lambda} < 1$

under polynomial criterion $\lambda = O(1)$:

$\gg 2^{O(\sqrt{\log \log n})}$ rounds

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BOOTSTRAPPING:
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$$T_{LLL}(n) = 2^{O(\sqrt{\log \log n})}$$

$2^{O(\sqrt{\log \log n})}$ - round algorithm
under $p(e \cdot d)^{32} < 1$ (POLYNOMIAL)
for $d = O(\log^{1/5} \log n)$

BASE ALGORITHM

$O(d^2) + \lambda \cdot \log^{\frac{1}{\lambda}} n \cdot 2^{O(\sqrt{\log \log n})}$ rounds under $p(e \cdot d)^{4\lambda} < 1$

Shattering Technique, rooted in breakthrough LLL algorithm of Beck [RSA'91]

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I. PARTIAL SAMPLING

assign values to subset of variables
such that remainder graph

- 1) satisfies polynomial criterion
- 2) consists of “small” components

in $O(d^2 + \log^* n)$ under $p(e \cdot d)^{4\lambda} < 1$

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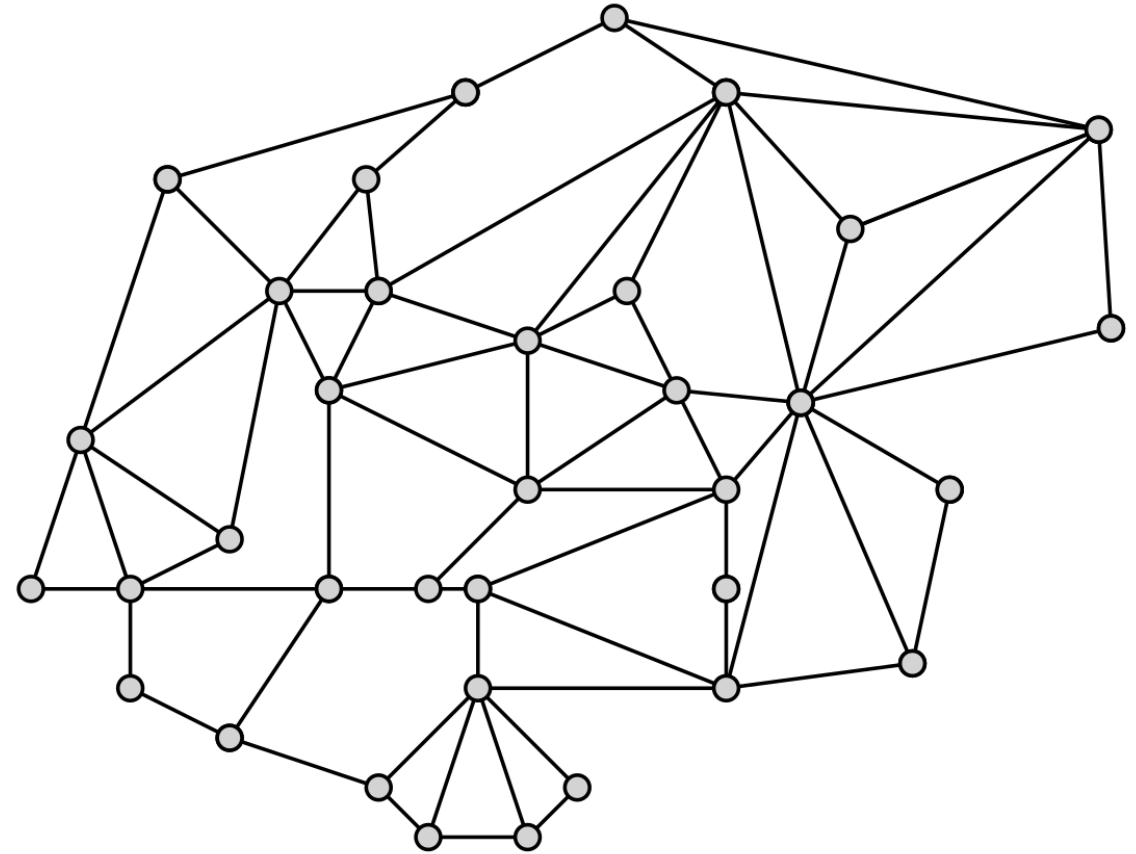
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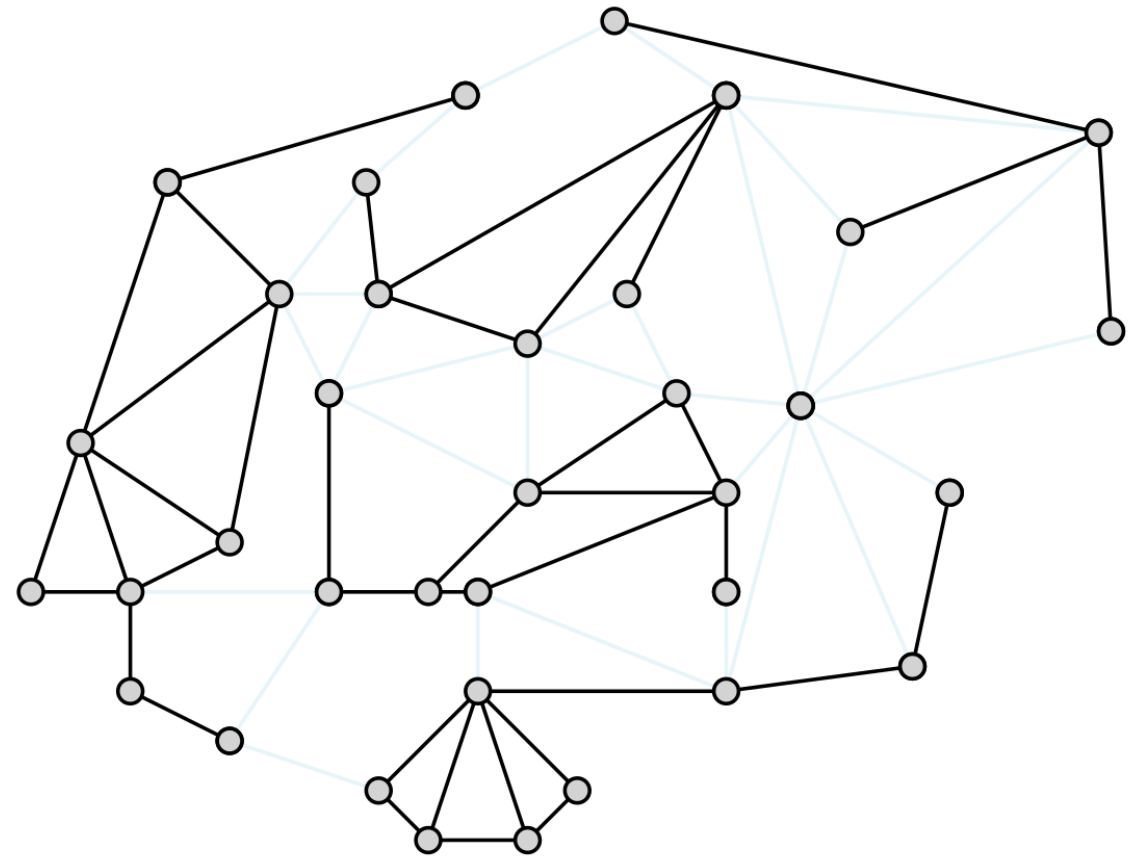
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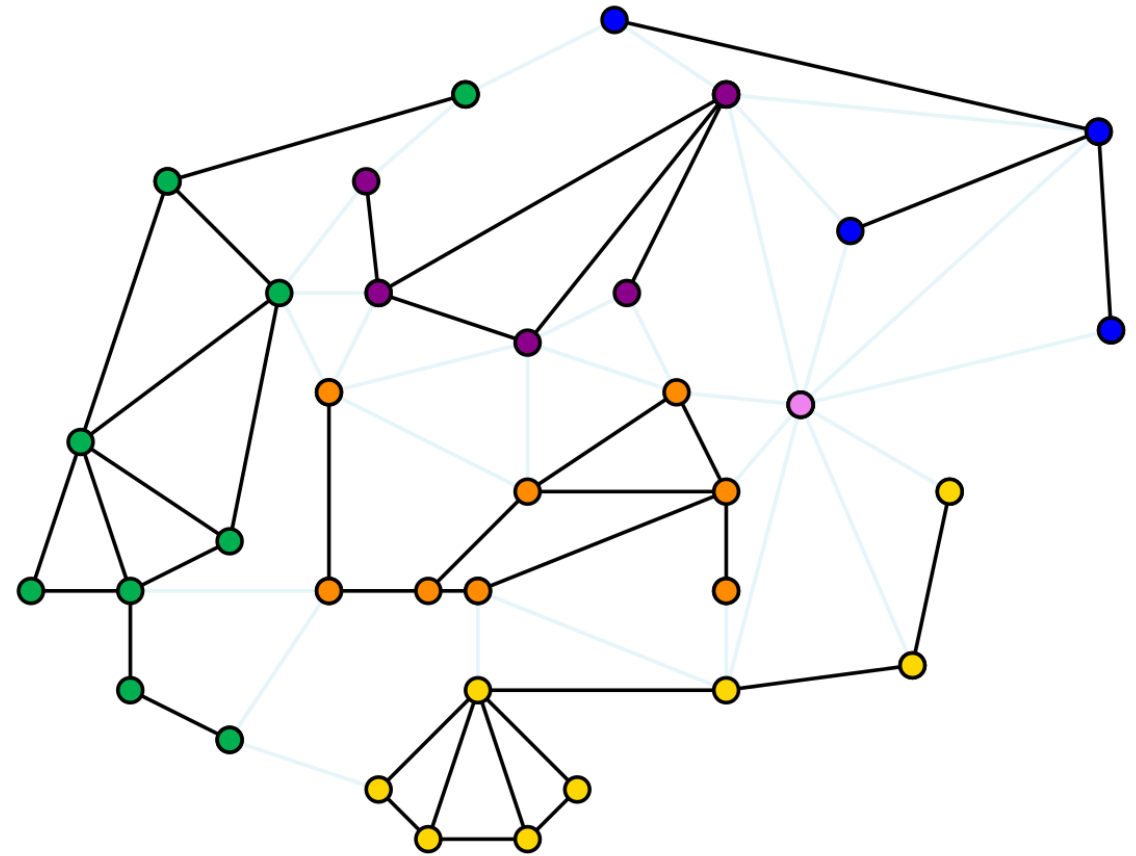
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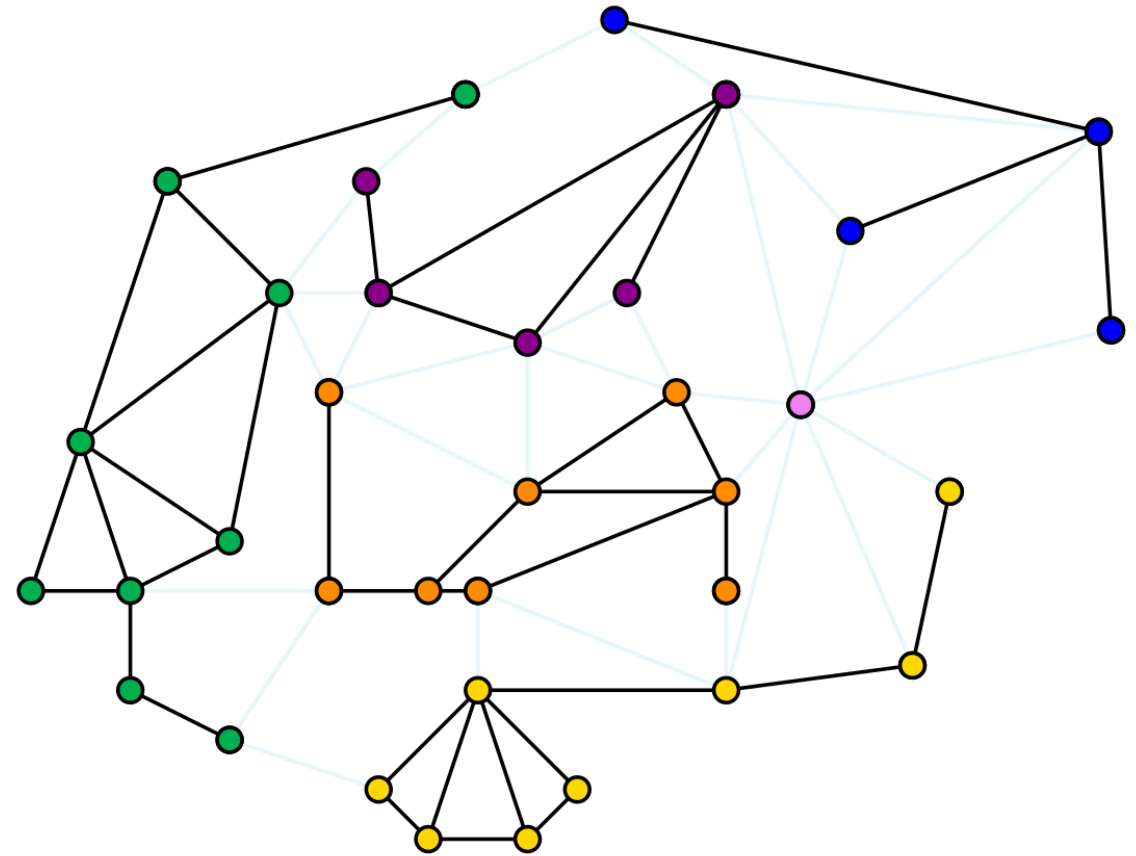
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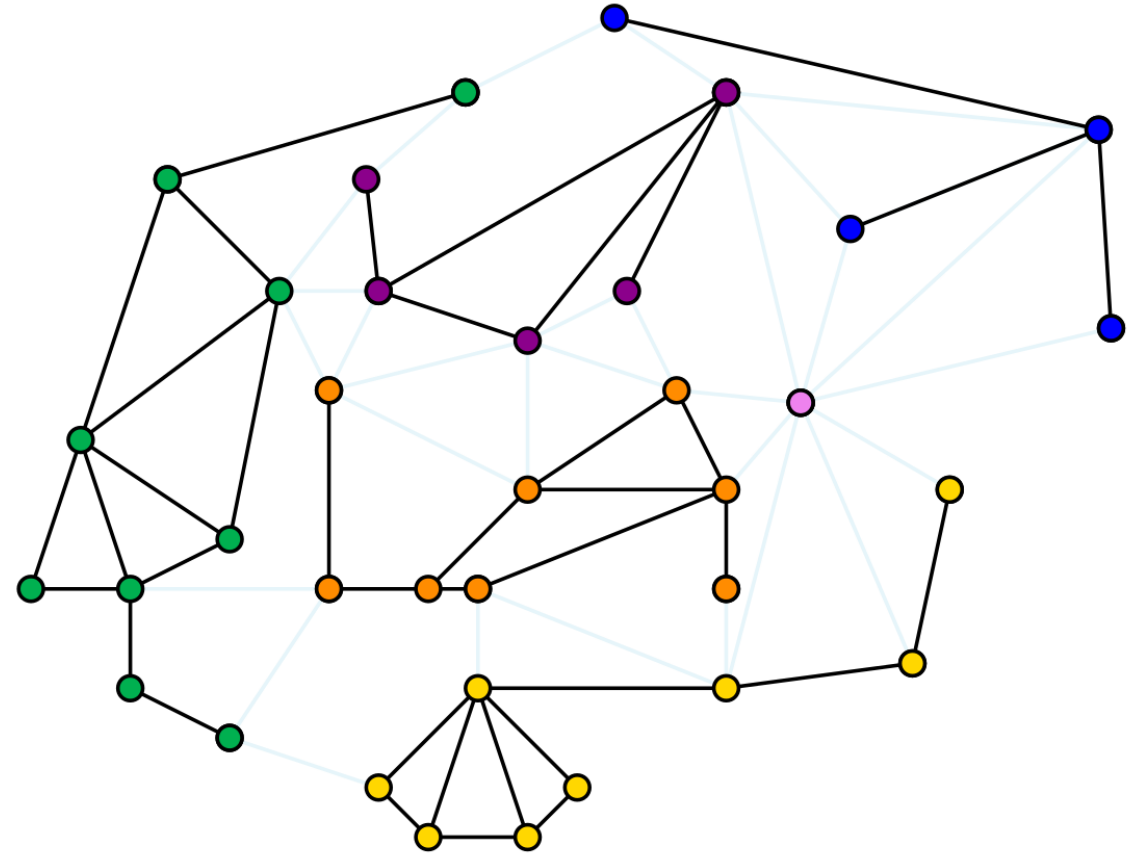
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I. PARTIAL SAMPLING in $O(d^2 + \log^* n)$ under $p(e \cdot d)^{4\lambda} < 1$ s.t. remainder

- 1) satisfies polynomial LLL
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Inspired by Molloy and Reed's
Sequential Partial Sampling [STOC'98]

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SEQUENTIAL PARTIAL SAMPLING

iteratively, for unblocked $v \in \mathcal{V}$

flip coin for v

if $P[A \mid \text{flipped coins}] > \frac{\sqrt{p}}{2}$

block $vbl(A)$

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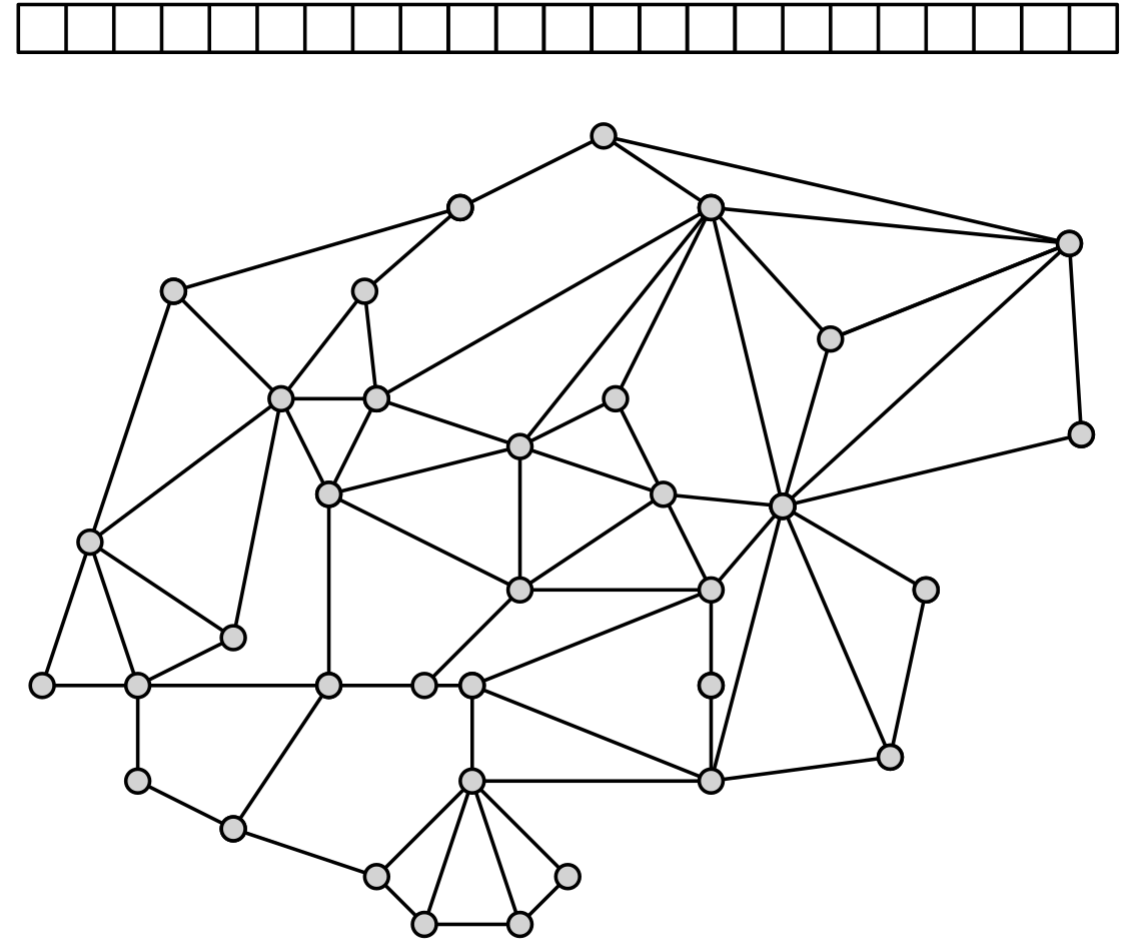
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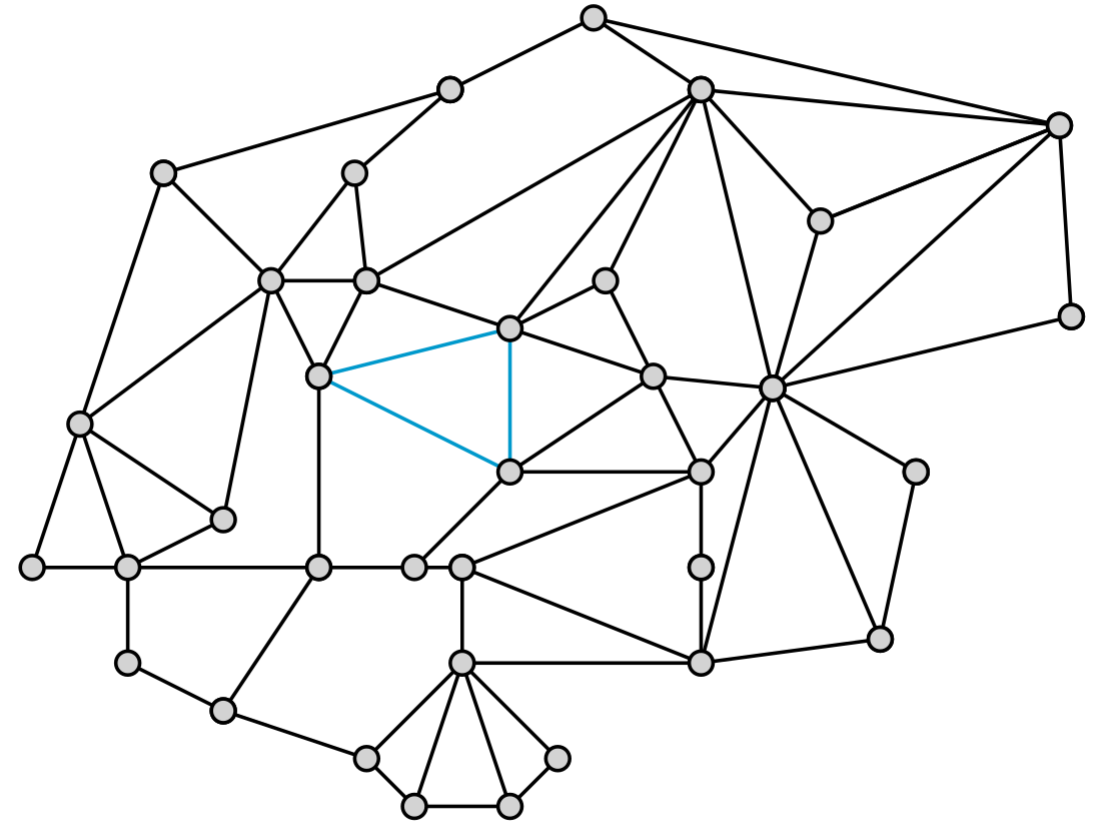
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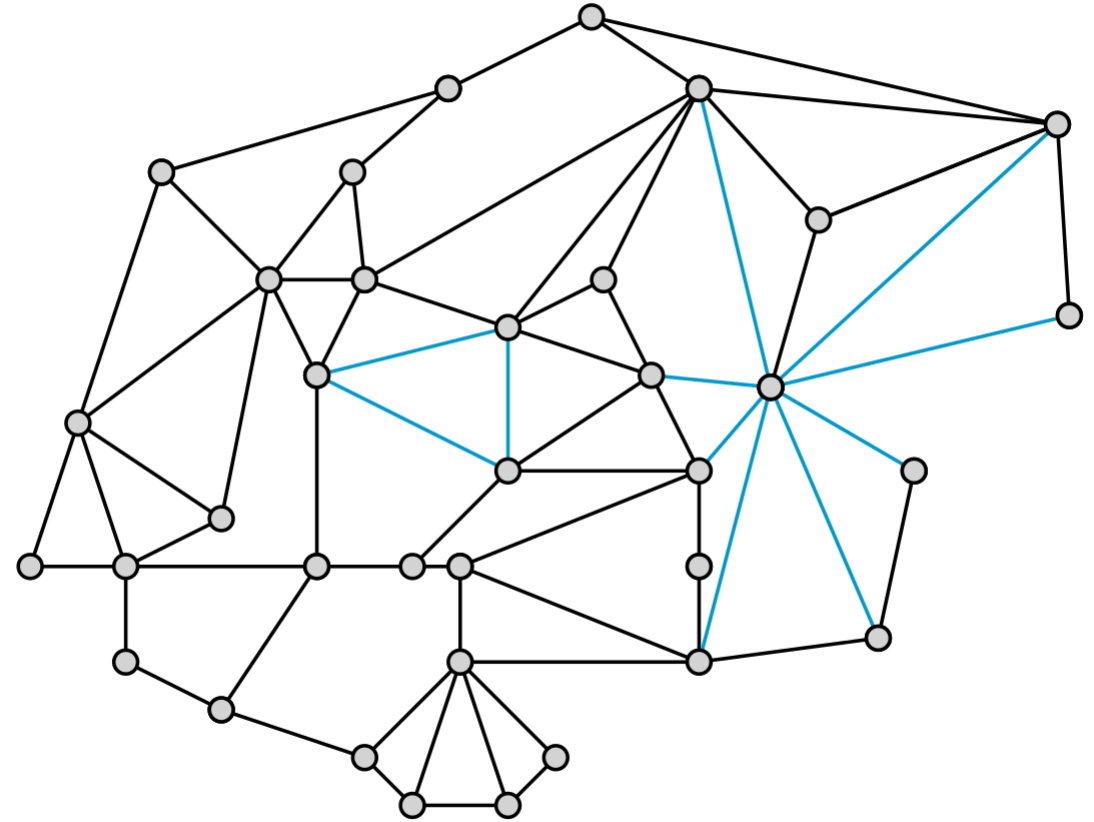
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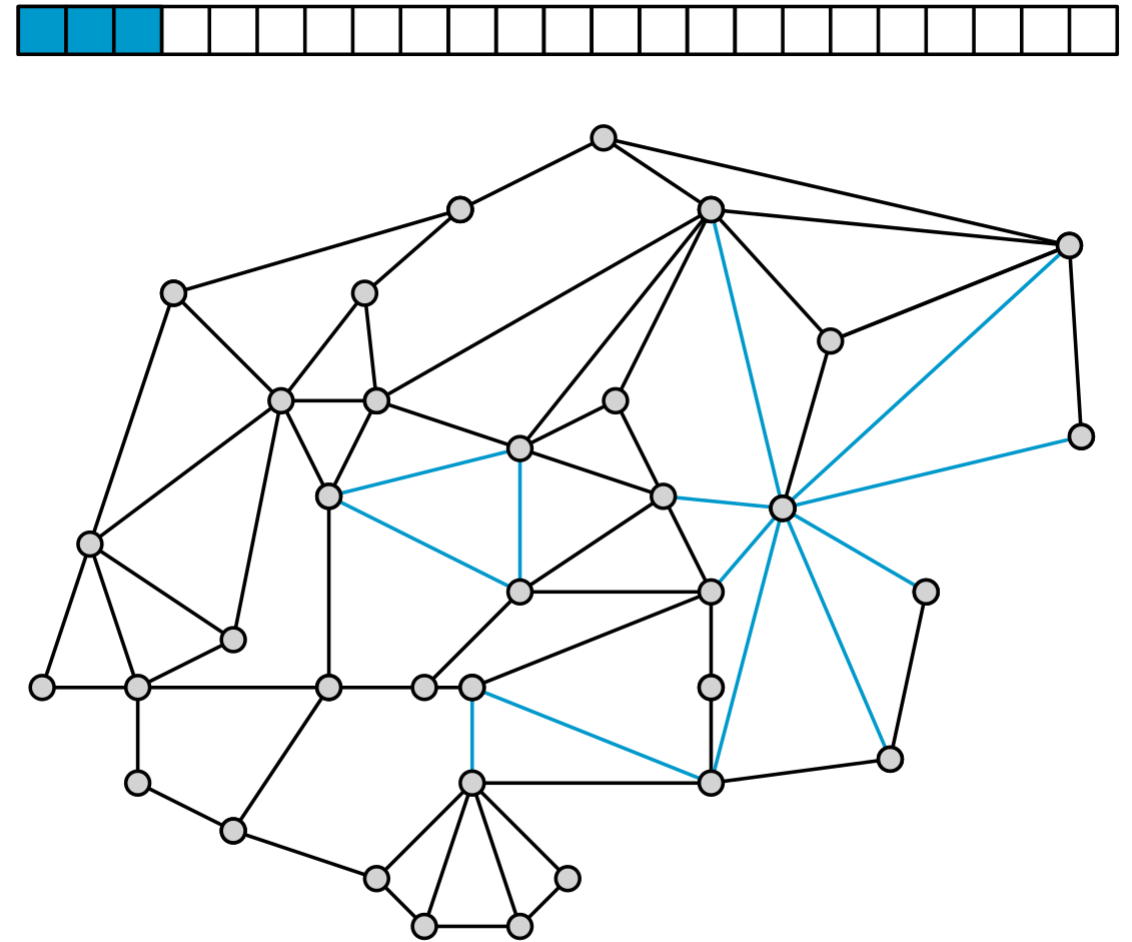
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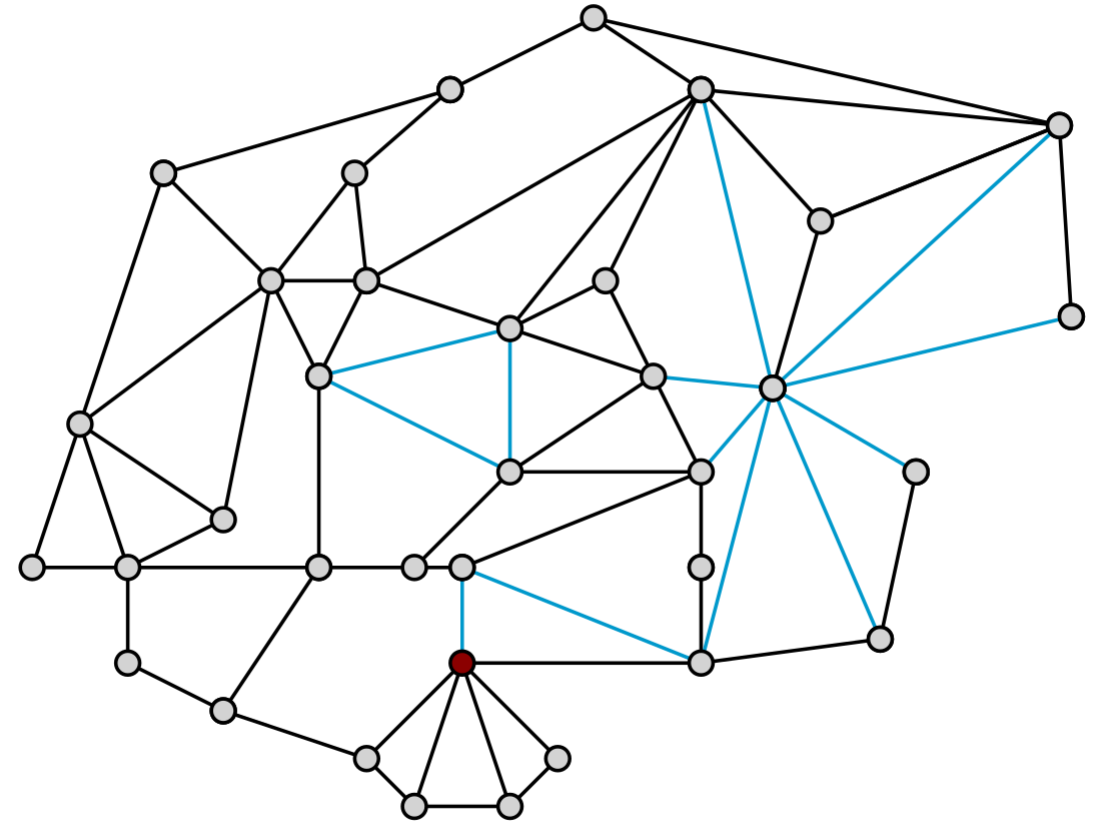
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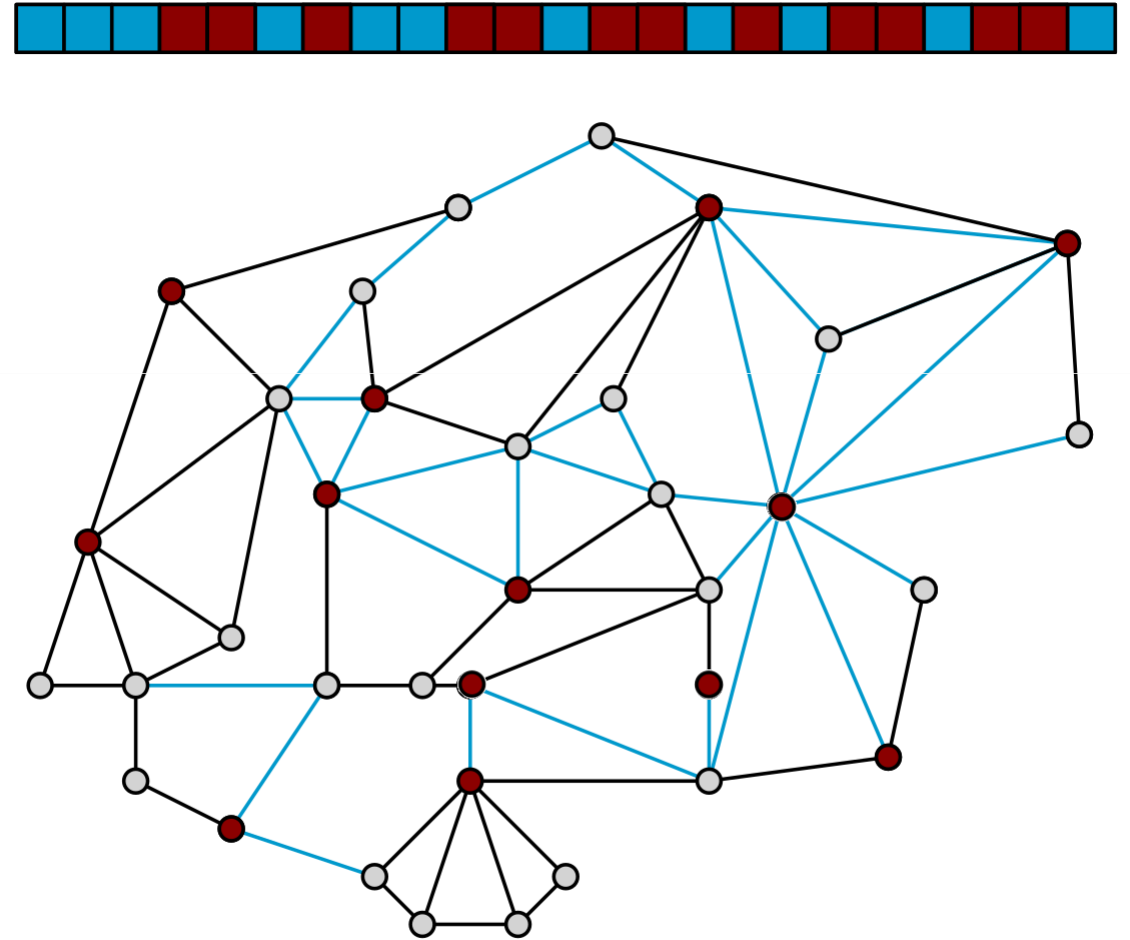
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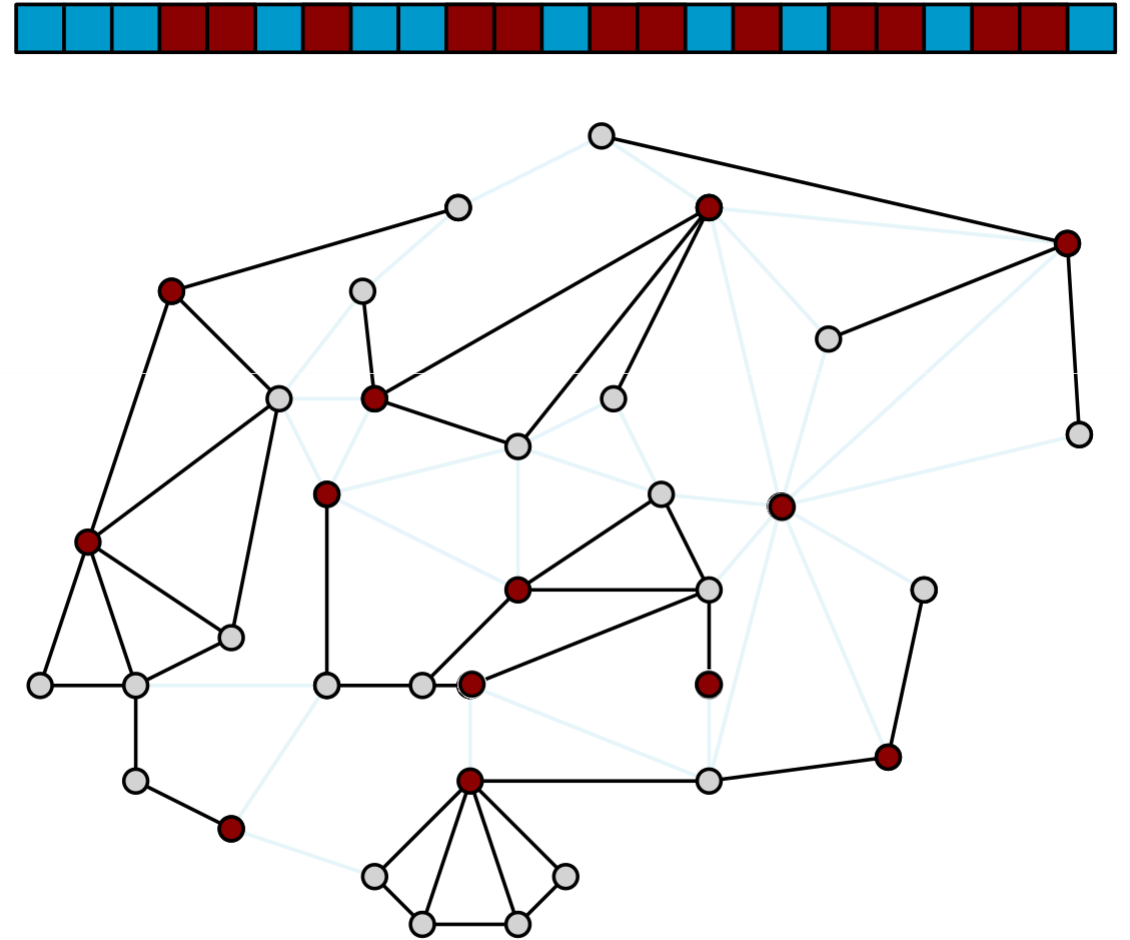
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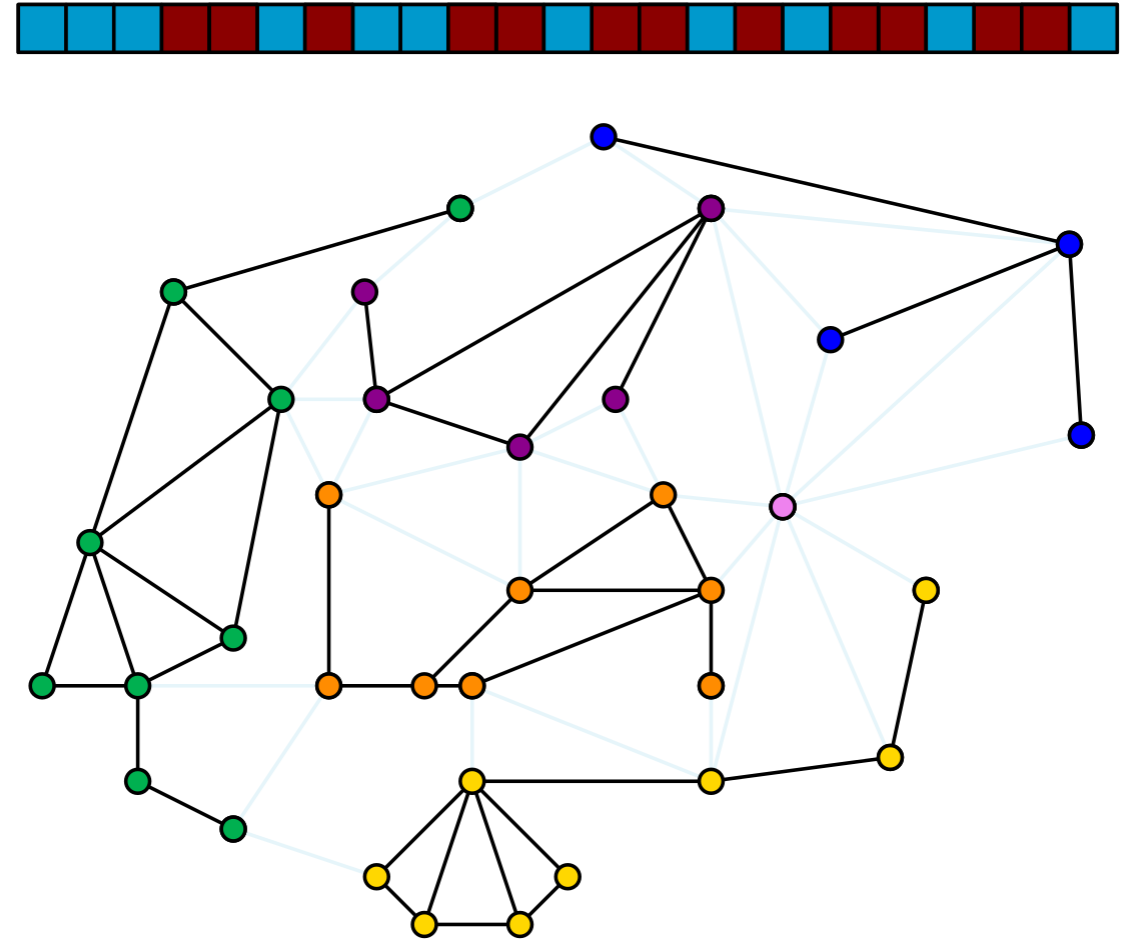
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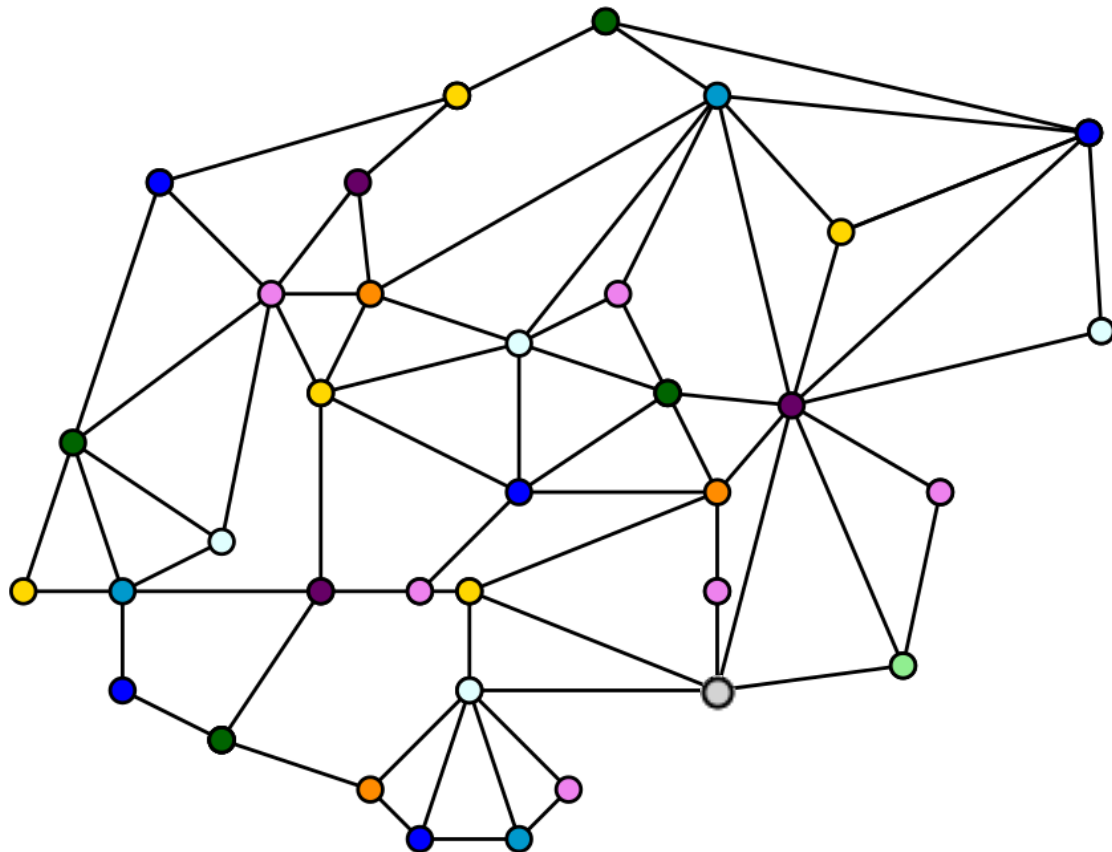
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OBSERVATIONS:

- $P[A \mid \text{flipped coins}] \leq \sqrt{p}$
satisfies polynomial LLL
- $P[A \text{ dangerous}] \leq \frac{p}{\frac{\sqrt{p}}{2}} = 2\sqrt{p}$
- $P[A \text{ remains}] = O(d)P[A \text{ dangerous}] \leq O(d\sqrt{p}) = \frac{1}{\text{poly } d}$
- A and B at distance > 2 remain independently

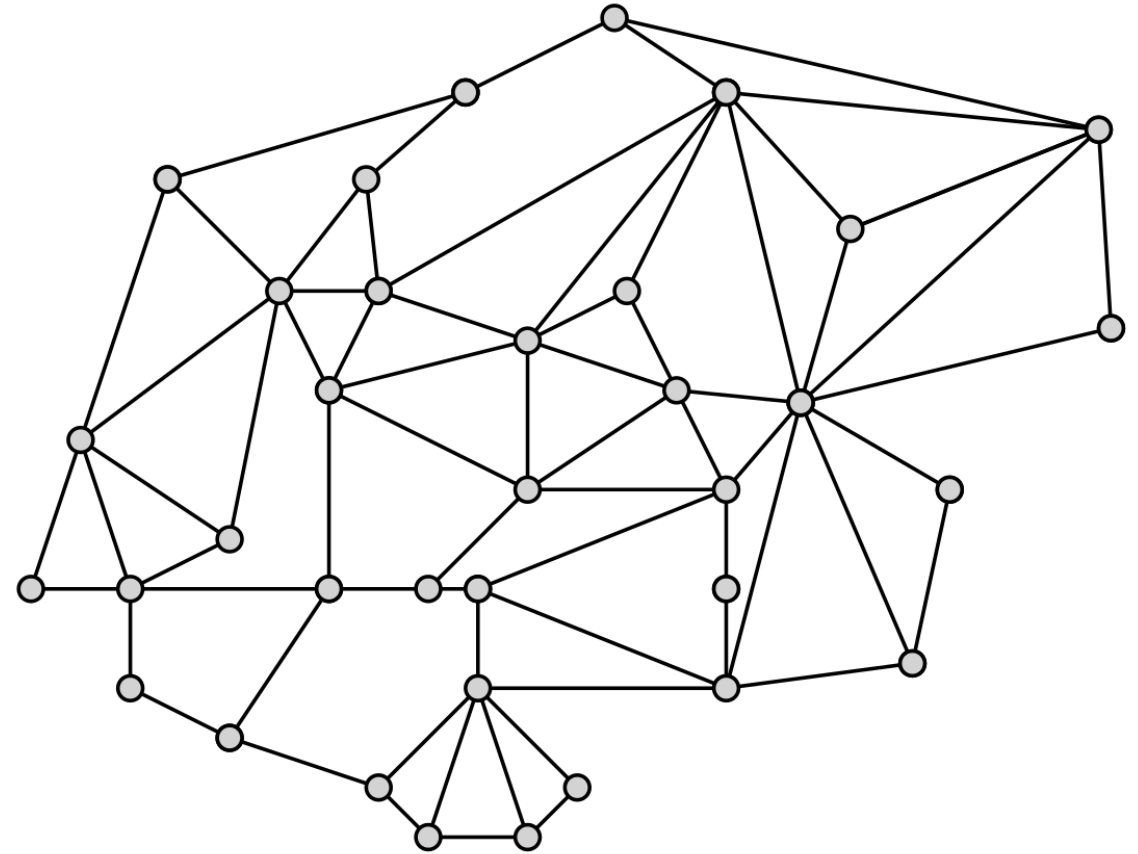
by Shattering Lemma: small components
(see, e.g., Barenboim, Elkin, Pettie, Schneider [FOCS'12])

- can be parallelized using a $(d^2 + 1)$ - coloring of G^2
consistent with one sequential global order

BASE ALGORITHM

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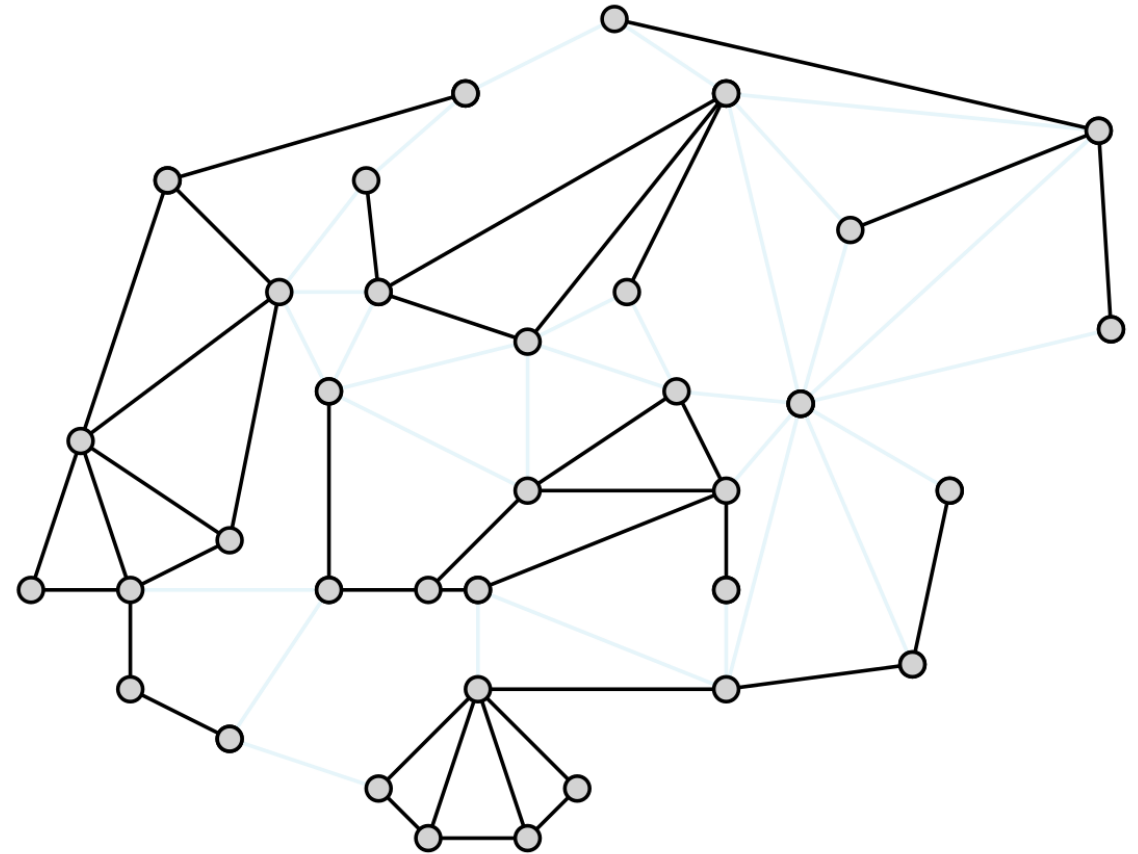
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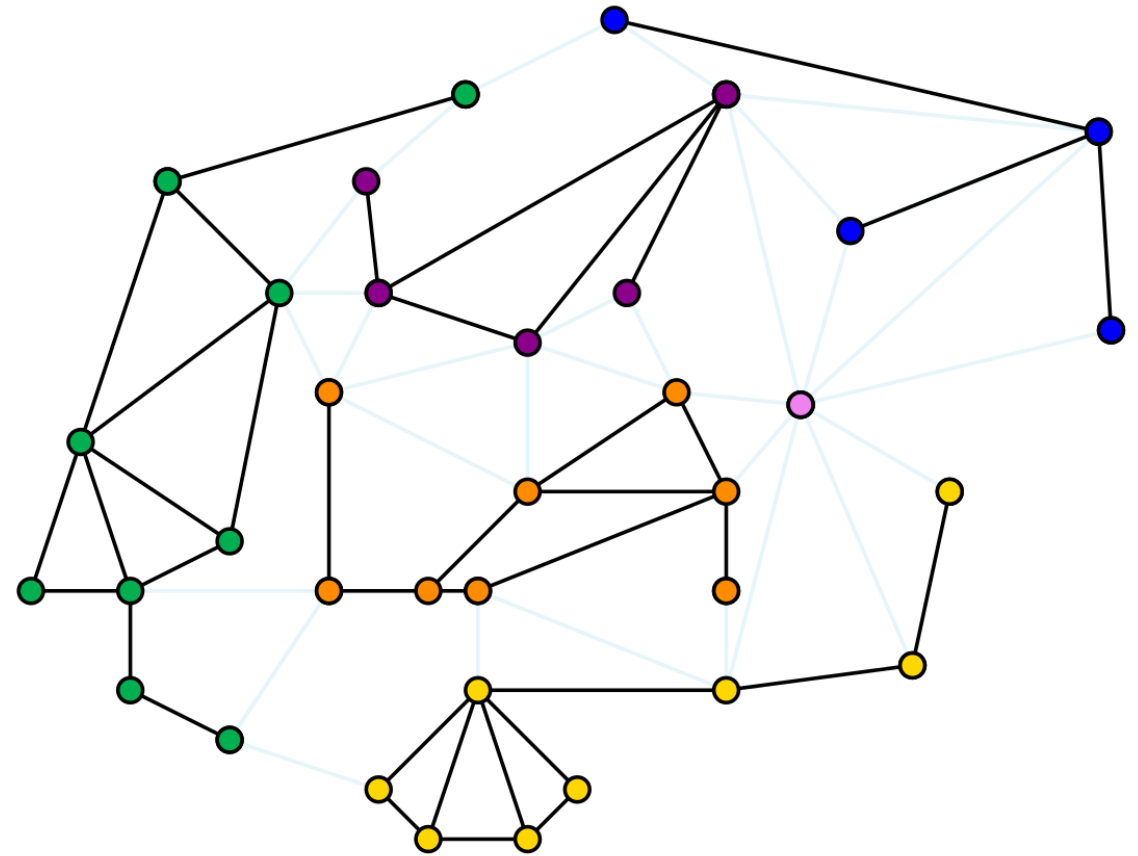
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II. DETERMINISTIC LLL ALGORITHM

$$\lambda \cdot n^{\frac{1}{\lambda}} \cdot 2^{O(\sqrt{\log n})} \text{ rounds under } p(e \cdot d)^{4\lambda} < 1$$

1. $(\lambda, \tilde{O}(n^{1/\lambda}))$ -network decomposition of G^2 in $\lambda \cdot n^{\frac{1}{\lambda}} \cdot 2^{O(\sqrt{\log n})}$

combining approaches by *Awerbuch and Peleg* [FOCS'90], *Panconesi and Srinivasan* [STOC'92], and *Awerbuch, Luby, Goldberg, Plotkin* [FOCS'89]

2. Iterative Assignment

Inductively, $P[A \mid \mathcal{V}_{\leq i}] \leq p(e \cdot d)^i$

Eventually, $P[A \mid \mathcal{V}_{\leq \lambda}] = P[A \mid \mathcal{V}] \leq p(e \cdot d)^\lambda < 1$

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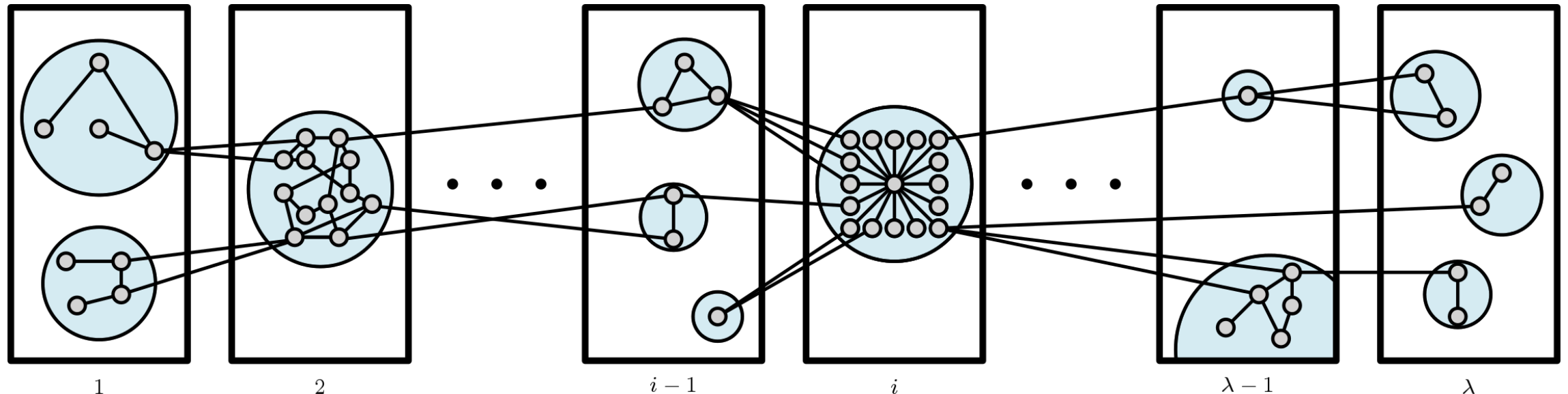
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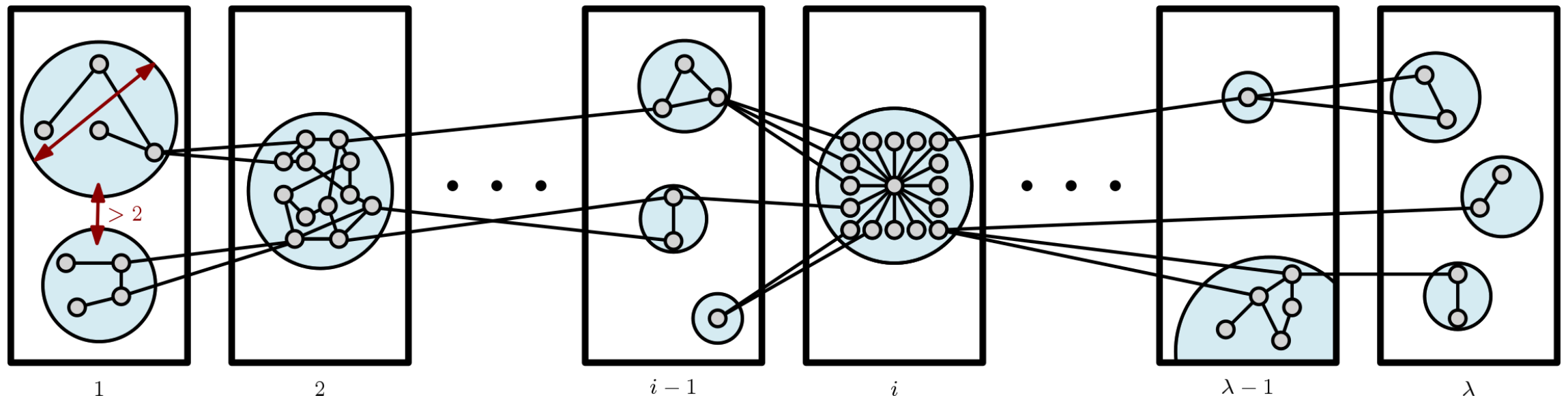
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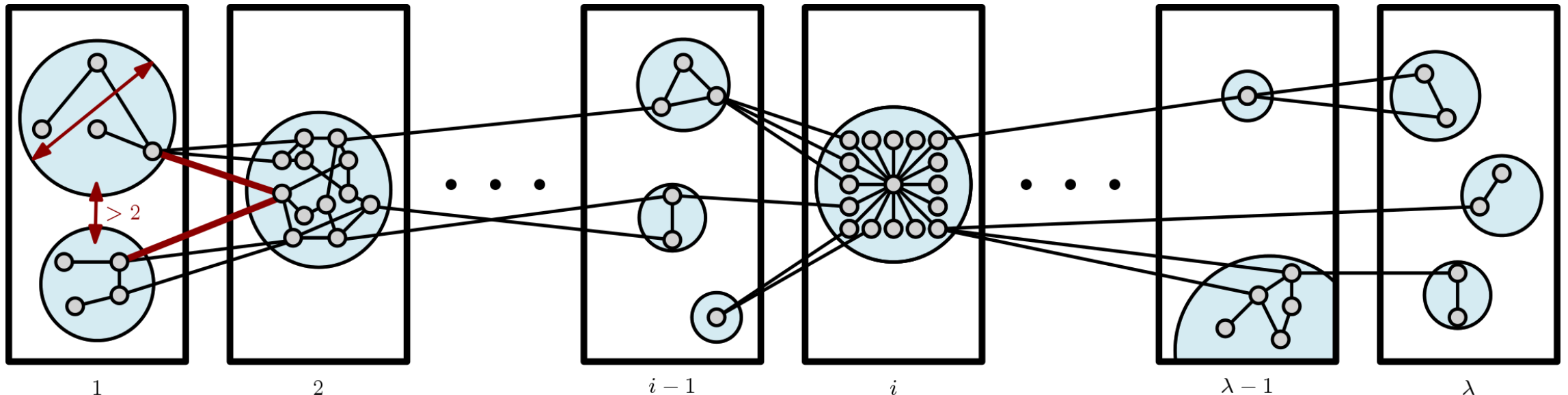
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combining approaches by *Awerbuch and Peleg* [FOCS'90], *Panconesi and Srinivasan* [STOC'92], and *Awerbuch, Luby, Goldberg, Plotkin* [FOCS'89]

2. Iterative Assignment

Inductively, $P[A \mid \mathcal{V}_{\leq i}] \leq p(e \cdot d)^i$

Eventually, $P[A \mid \mathcal{V}_{\leq \lambda}] = P[A \mid \mathcal{V}] \leq p(e \cdot d)^\lambda < 1$



II. DETERMINISTIC LLL ALGORITHM

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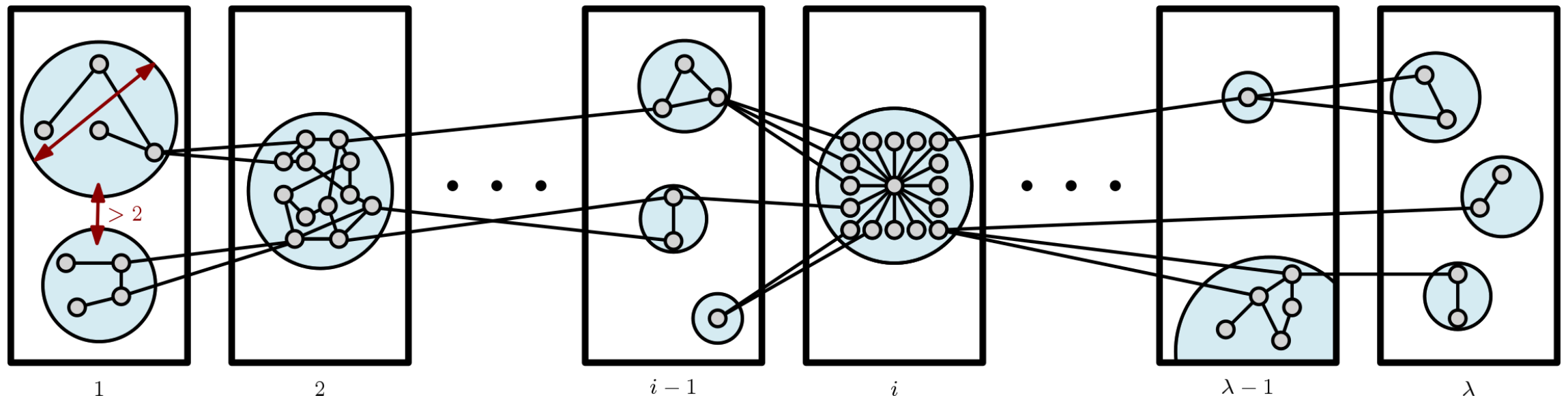
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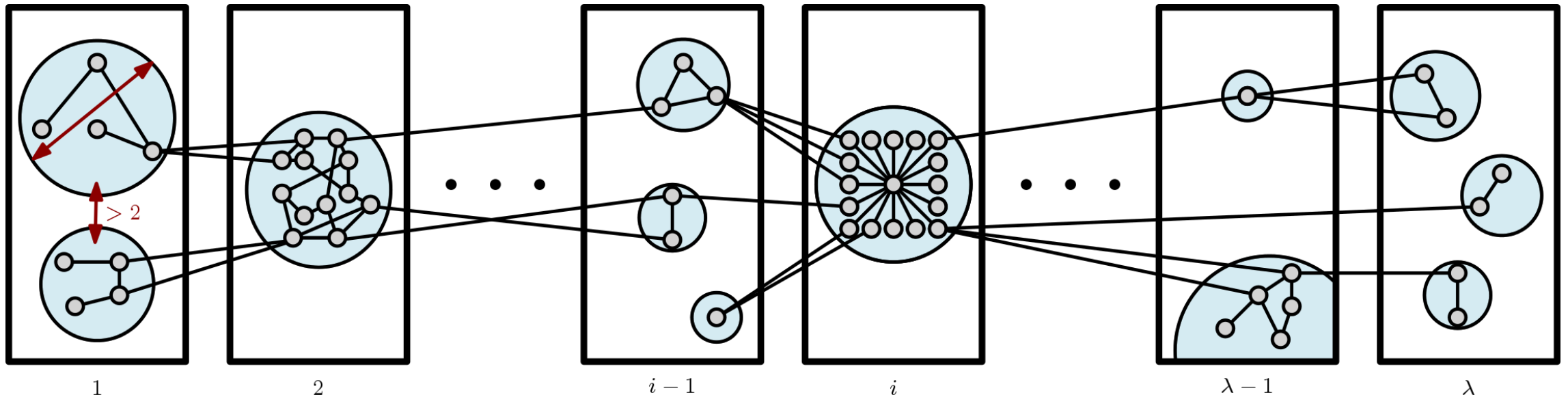
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LLL for step i

event $B_{A,i}$: $P[A \mid \mathcal{V}_{\leq i}] > p(e \cdot d)^i$

probability: $\leq \frac{1}{e \cdot d}$



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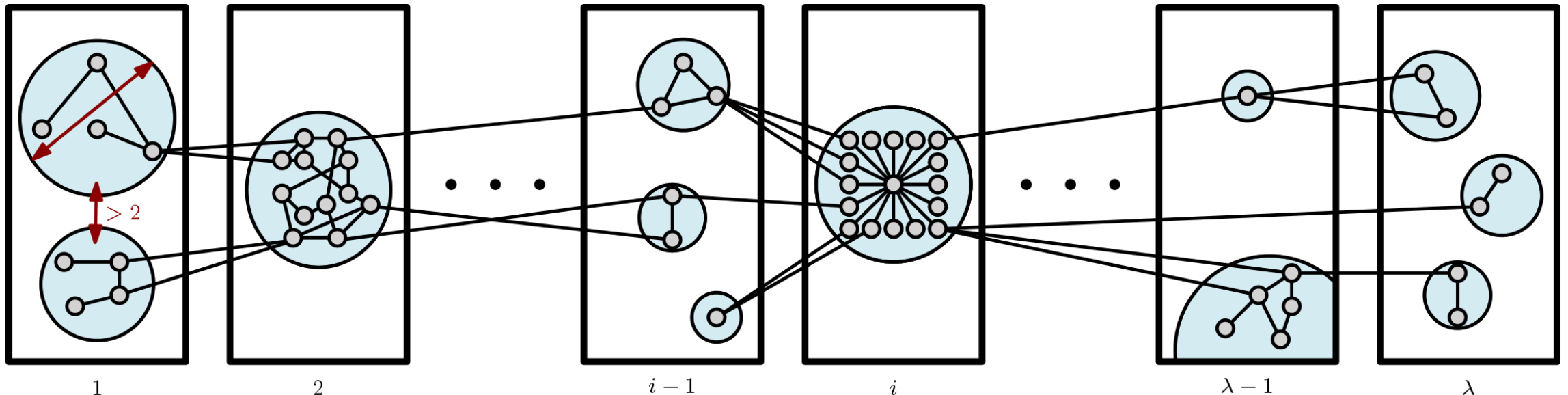
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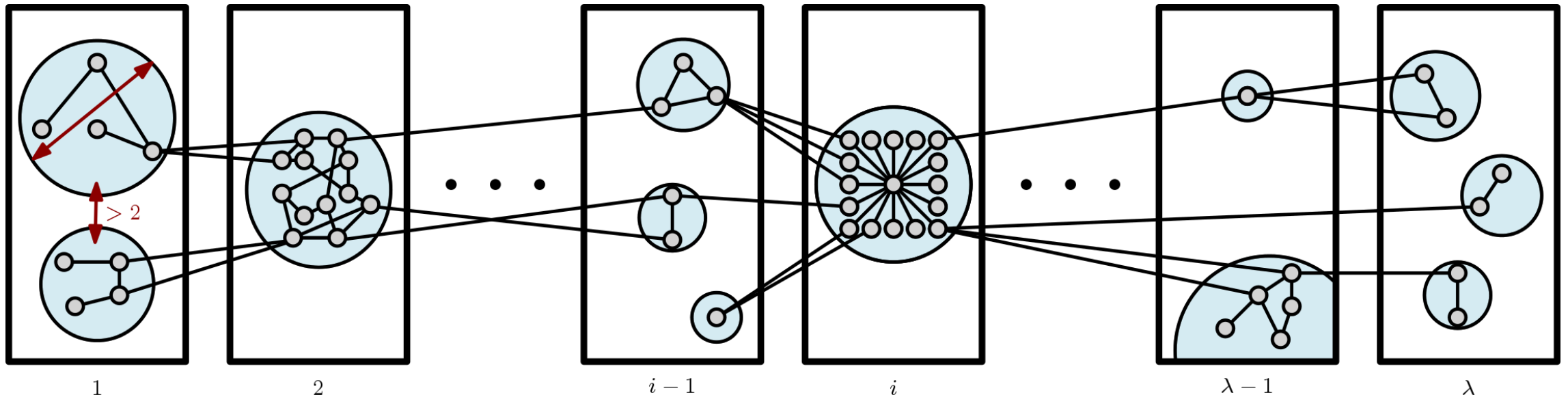
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LLL for step i satisfies standard LLL criterion

event $B_{A,i}$: $P[A \mid \mathcal{V}_{\leq i}] > p(e \cdot d)^i$

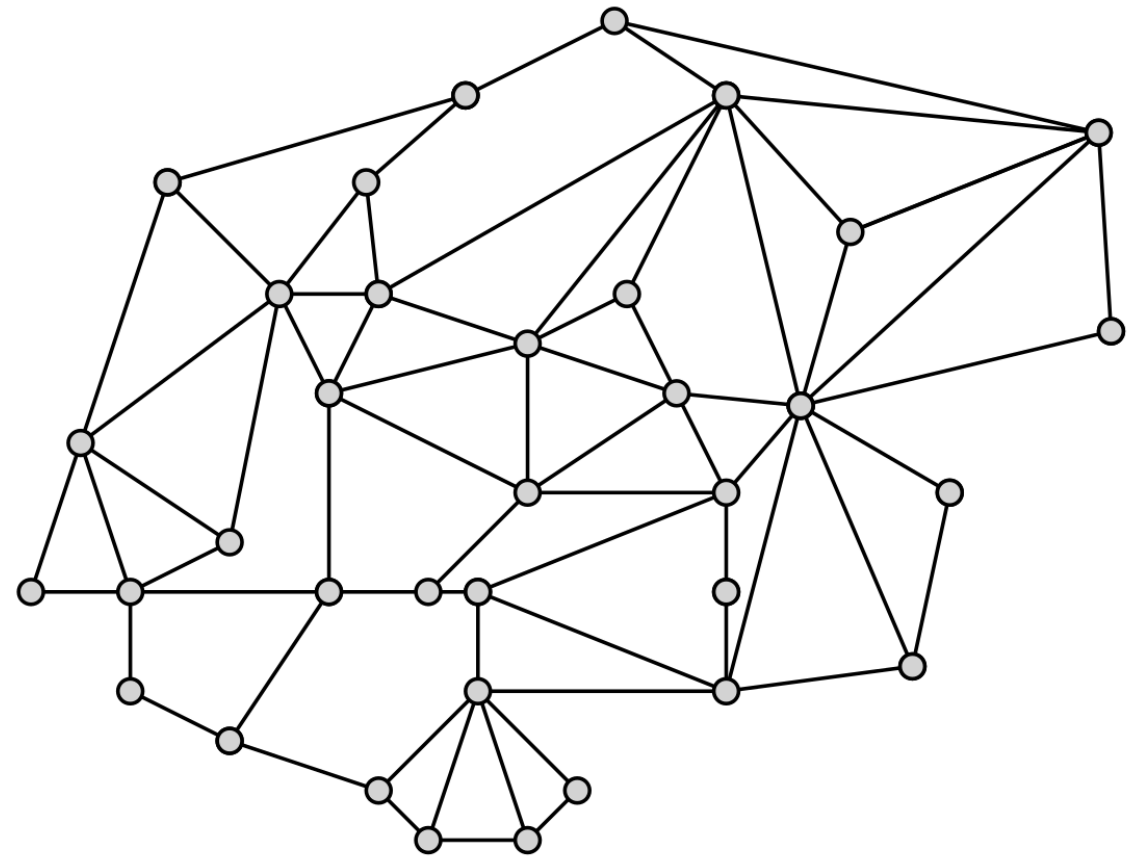
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BASE ALGORITHM

I. PARTIAL SAMPLING

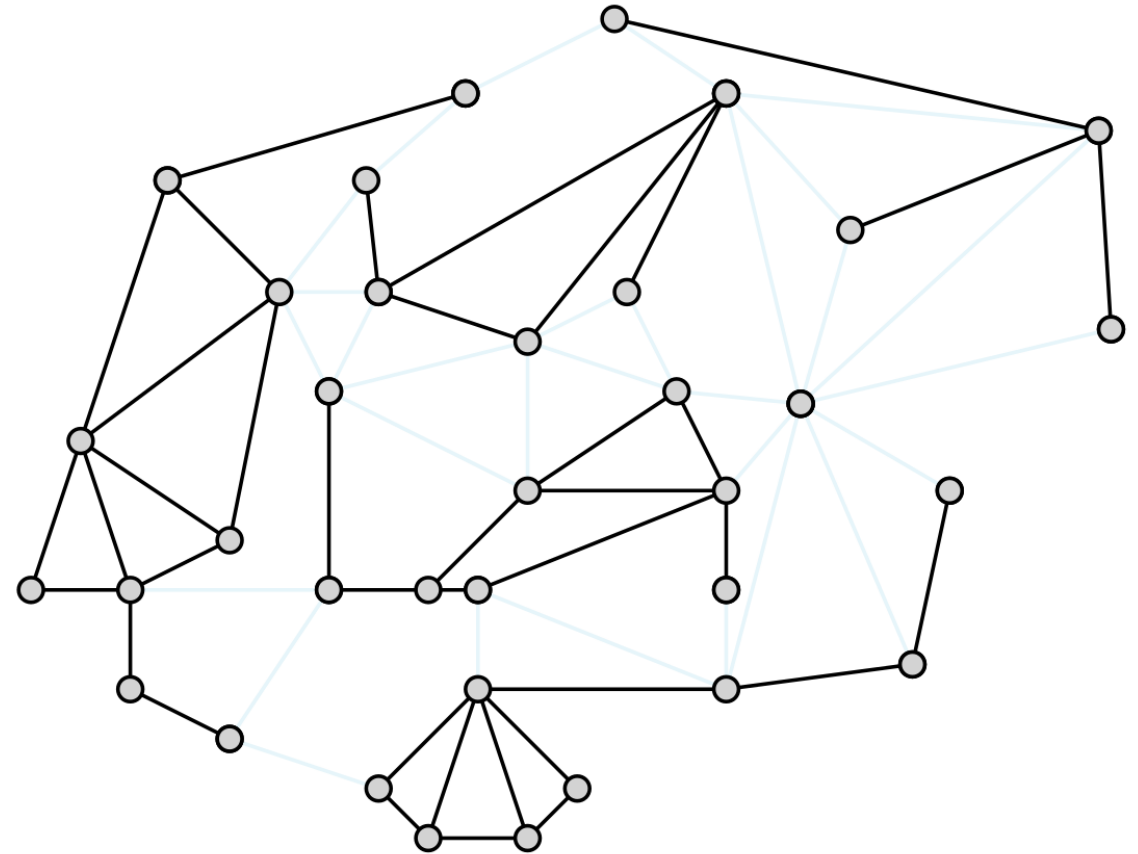
II. DETERMINISTIC LLL ALGORITHM



BASE ALGORITHM

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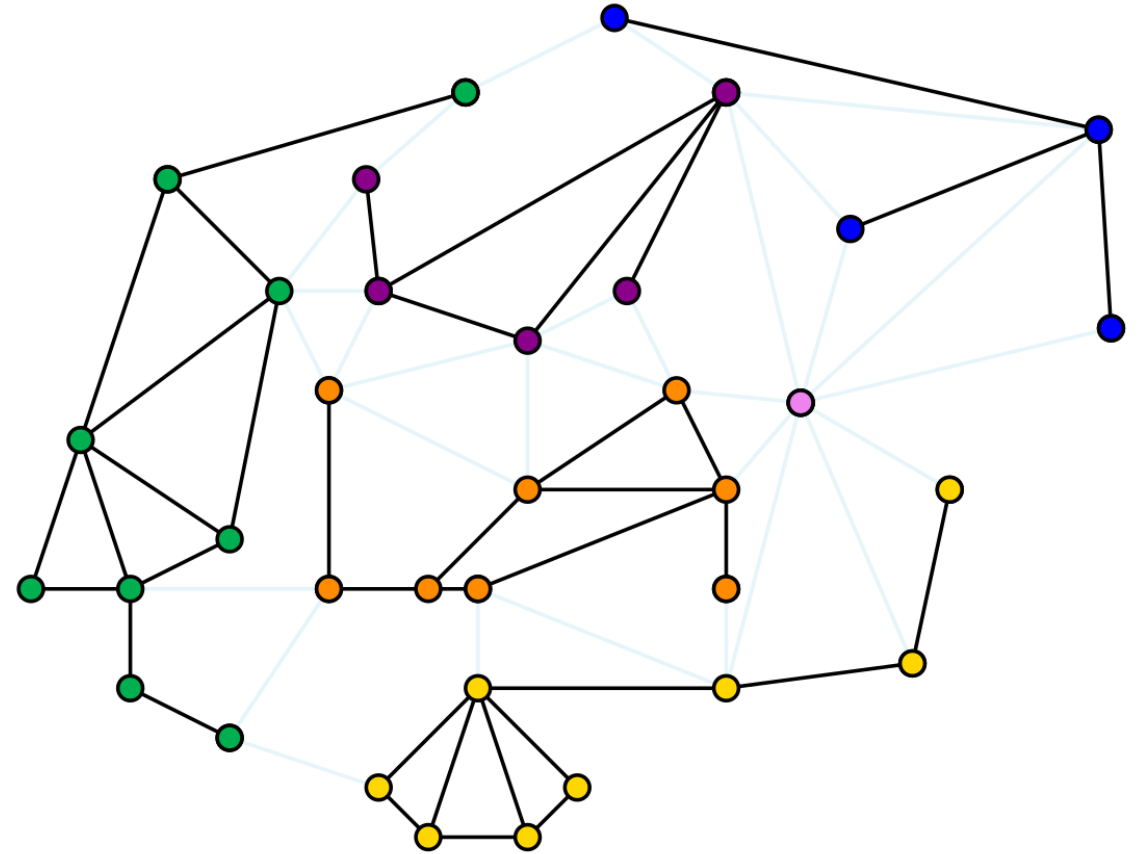
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BASE ALGORITHM

I. PARTIAL SAMPLING

II. DETERMINISTIC LLL ALGORITHM



Summary and Open Problems

BASE ALGORITHM

BOOTSTRAPPING
(Speed-Up)



$$T_{LLL}(n) = 2^{O(\sqrt{\log \log n})}$$

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CONJECTURE:

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Chang, Pettie [FOCS'16]

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$$T_{LLL}(n) = 2^{O(\sqrt{\log \log n})}$$

CONJECTURE:

$$T_{LLL}(n) = O(\text{poly log log } n)$$

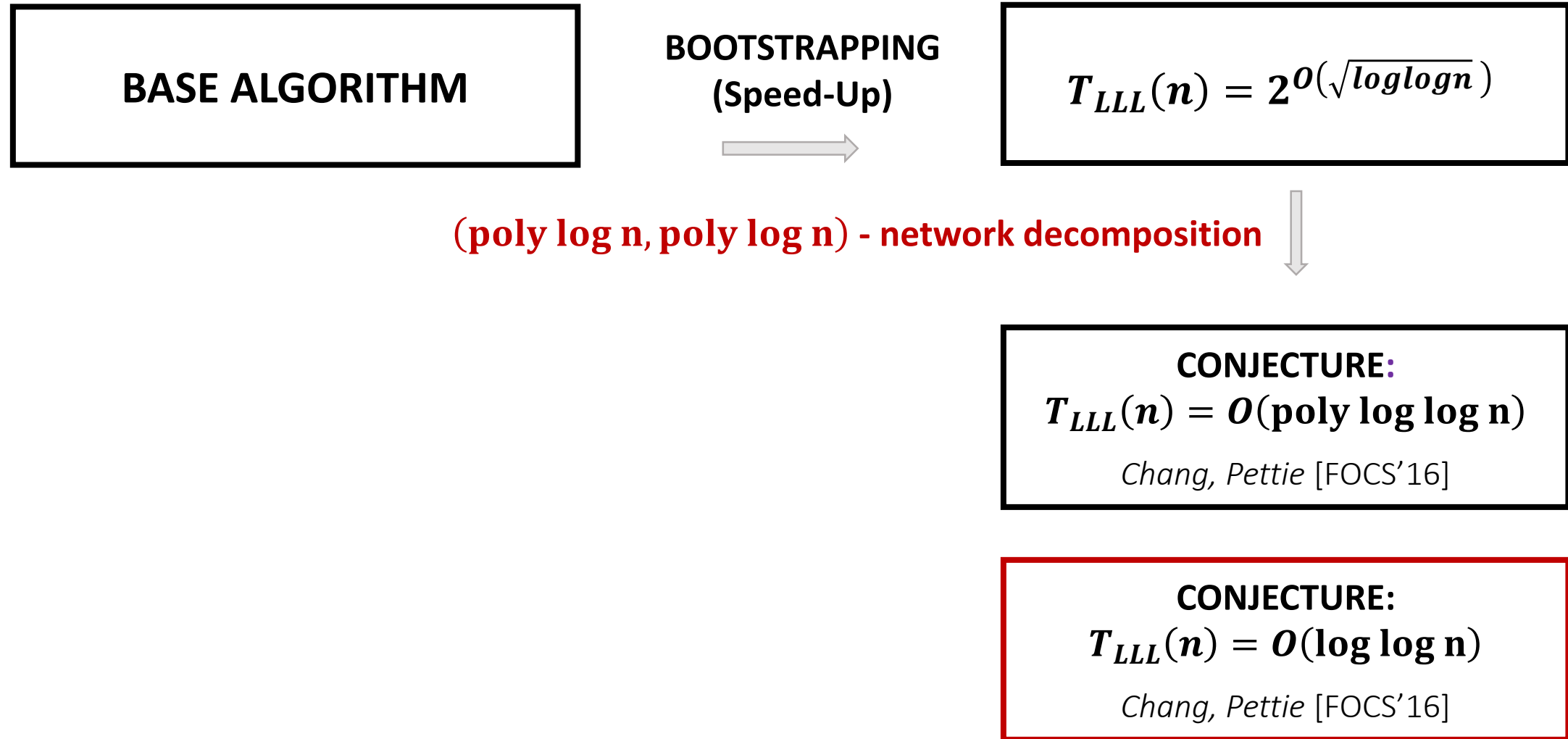
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Summary and Open Problems



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(Speed-Up)**



$$T_{LLL}(n) = 2^{O(\sqrt{\log \log n})}$$

(poly log n, poly log n) - network decomposition



- **Devise a faster deterministic algorithm.**
- **Devise a faster algorithm under weaker LLL condition.**

CONJECTURE:

$$T_{LLL}(n) = O(\text{poly log log } n)$$

Chang, Pettie [FOCS'16]

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Summary and Open Problems

