

# LOCAL Algorithms:

## The Chasm Between Deterministic & Randomized

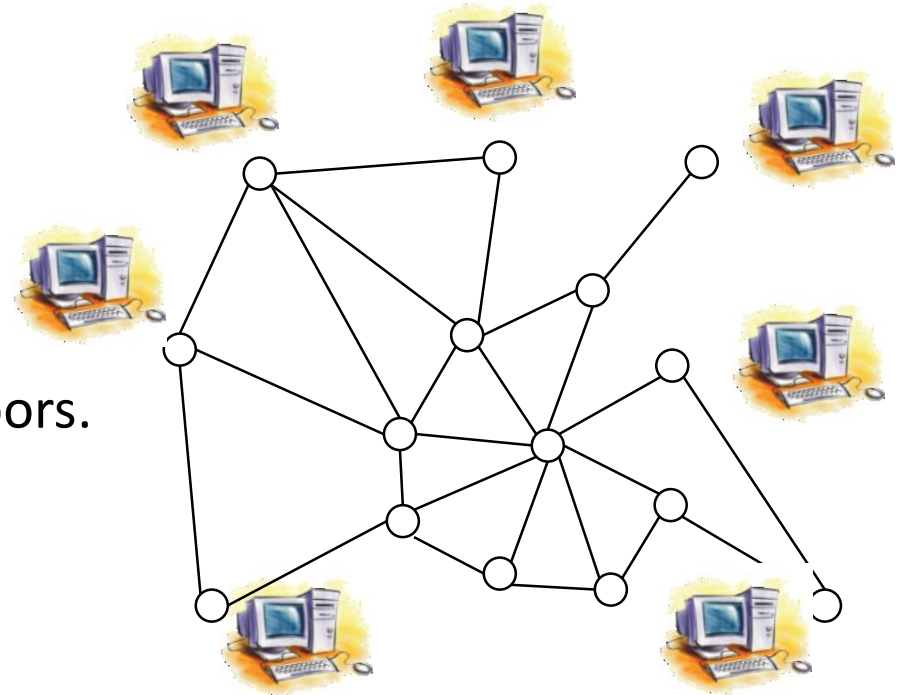
**Mohsen Ghaffari**

ETH Zurich

# The LOCAL Model of Distributed Graph Algorithms

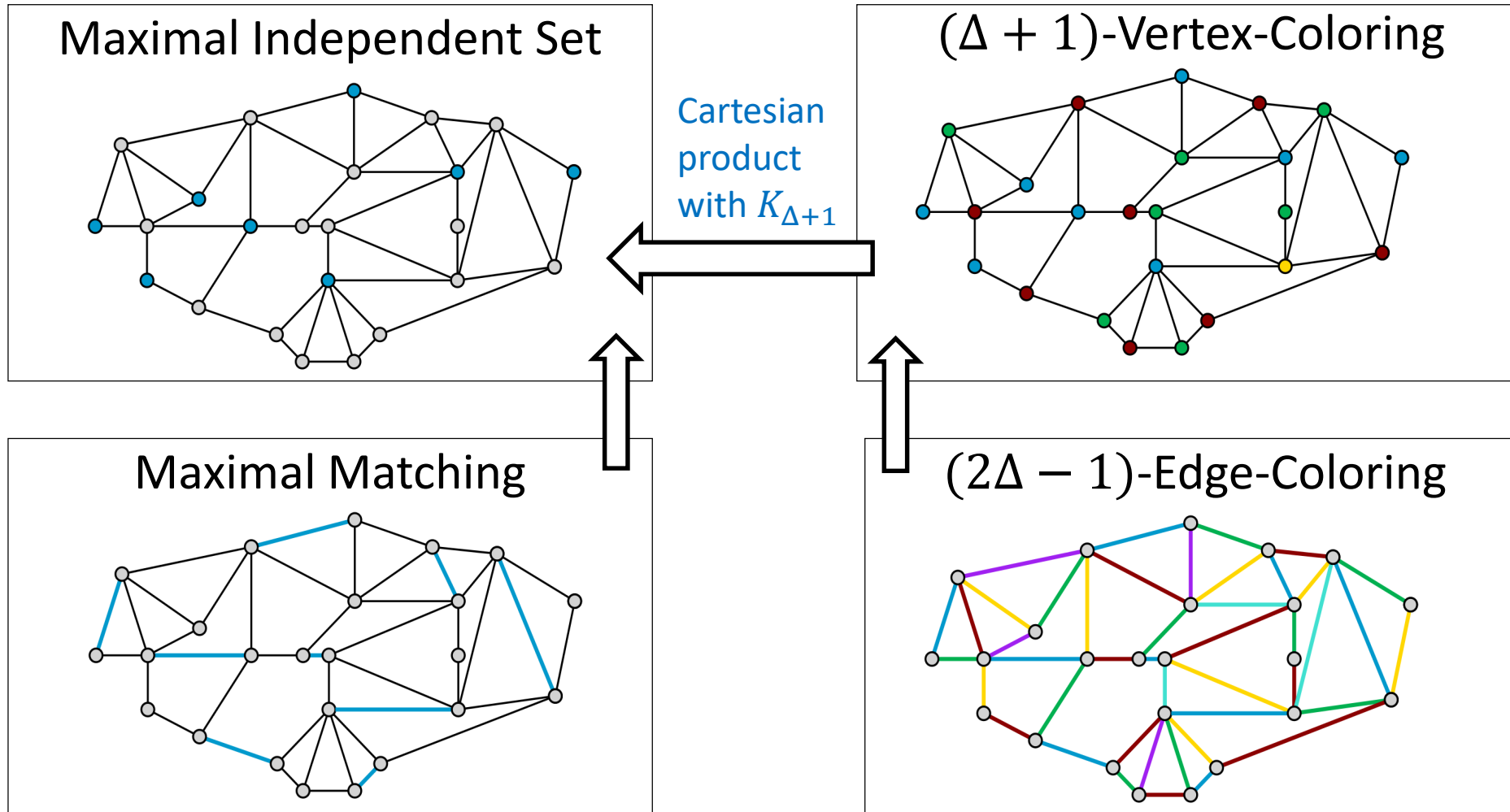
## LOCAL model [Linial FOCS'87]

- undirected graph  $G = (V, E)$ ,  $n$  nodes, max degree  $\Delta$
- one computer on each graph node,
- Synchronous message-passing rounds  $1, 2, 3, \dots$   
per round, each node sends one message to its neighbors.
- Unbounded message size & computation.
  
- Initially nodes do not know the topology
- Each node should learn its own part of the output, e.g., its color.



**Time-Complexity:** number of rounds until all nodes are done.

# Four Classic Problems (since 1980's)

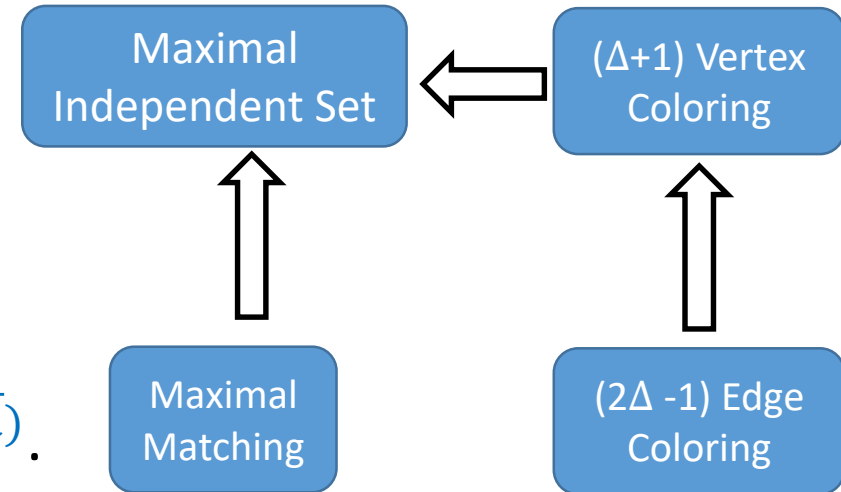


# State of the Art

# Randomized vs. Deterministic LOCAL Algorithms

First-Order Summary: Significant gap between Randomized & Deterministic

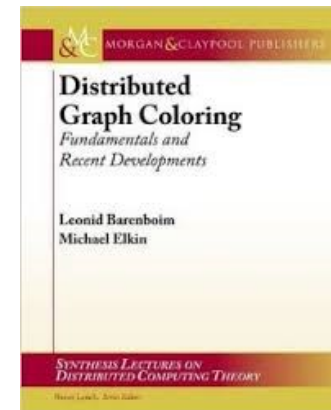
- **Randomized:** Very simple  $O(\log n)$ -round algorithms, and even some  $o(\log n)$ -round algorithms
- **Deterministic:**  $\text{polylog } n$  is the dream, for most problems, the the best known is  $2^{O(\sqrt{\log n})}$ .



**Linial's Open Question [FOCS'87, SICOMP'92]:**

“Can it [MIS] always be found [deterministically] in polylogarithmic time?”

**Also the first 5 open problems of the Distributed Graph Coloring book [Barenboim & Elkin]**



# Maximal Independent Set (MIS): Current State

## Lower Bound

- $\Omega\left(\min\left\{\frac{\log \Delta}{\log \log \Delta}, \sqrt{\frac{\log n}{\log \log n}}\right\}\right)$  rounds needed [Kuhn, Moscibroda, Wattenhofer PODC'04]

## Efficient Randomized Algorithms:

- An  $O(\log^4 n)$ -time algorithm [Karp, Wigderson STOC'84]
- A truly-simple  $O(\log n)$ -time algorithm [Luby STOC'85; Alon, Isreali, Itai JALG'86]
- Best upper bound:  $O(\log \Delta) + 2^{O(\sqrt{\log \log n})}$  [G., SODA'16]

## Best Known Deterministic Algorithm

- Based on network decomposition:  $2^{O(\sqrt{\log n})}$  [Panconesi, Srinivasan STOC'92]

# $(\Delta + 1)$ -Vertex-Coloring: Current State

## Lower Bound

- $\Omega(\log^* n)$  rounds needed even on the ring

[Linial FOCS'87]

## Efficient Randomized Algorithms

- Simple randomized  $O(\log n)$ -time algorithms [Luby STOC'86; Alon, Isreali, Itai JALG'86]
- Best current upper bound:  $O\left(\sqrt{\log \Delta}\right) + 2^{O\left(\sqrt{\log \log n}\right)}$  [Harris, Schneider, Su STOC'16]

## Best Known Deterministic Algorithm

- Based on network decomposition:  $2^{O\left(\sqrt{\log n}\right)}$

[Panconesi, Srinivasan STOC'92]

# State of the Art: Deterministic vs. Randomized

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

[Panconesi, Srinivasan  
STOC'92]

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

[Panconesi, Srinivasan  
STOC'92]

$$O(\log \Delta) + 2^{O(\sqrt{\log \log n})}$$

[G., SODA'16]

$$O(\sqrt{\log \Delta}) + 2^{O(\sqrt{\log \log n})}$$

Harris, Schneider, Su [STOC'16]

Ir

- In RANDOMIZED, the  $n$ -dependencies come from deterministically solving  $\text{polylog } n$ -size graphs.
  - All are necessary, unless we improve the deterministic algorithms.
- Thm** [Chang, Kopelowitz, & Pettie FOCS'16]:

$$\text{RandomizedComplexity}(n) \geq \text{DeterministicComplexity}(\sqrt{\log n})$$

$$O(\log^4 n)$$

[Hanckowiak, Karonski,  
Panconesi SODA'98 &  
PODC'99]

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

[Panconesi, Srinivasan  
STOC'92]

$$O(\log \Delta) + O(\log^4 \log n)$$

Barenboim, Elkin, Pettie,  
& Schneider [FOCS'11]

$$O(\sqrt{\log \Delta}) + 2^{O(\sqrt{\log \log n})}$$

Harris, Schneider, Su [STOC'16]

**DETERMINISTIC**

**RANDOMIZED**



# Some Other Related Work

## Exponential Separations

[Chang, Kopelowitz, Pettie FOCS'16], [Brandt et al. STOC'16], [G., Su SODA'17]

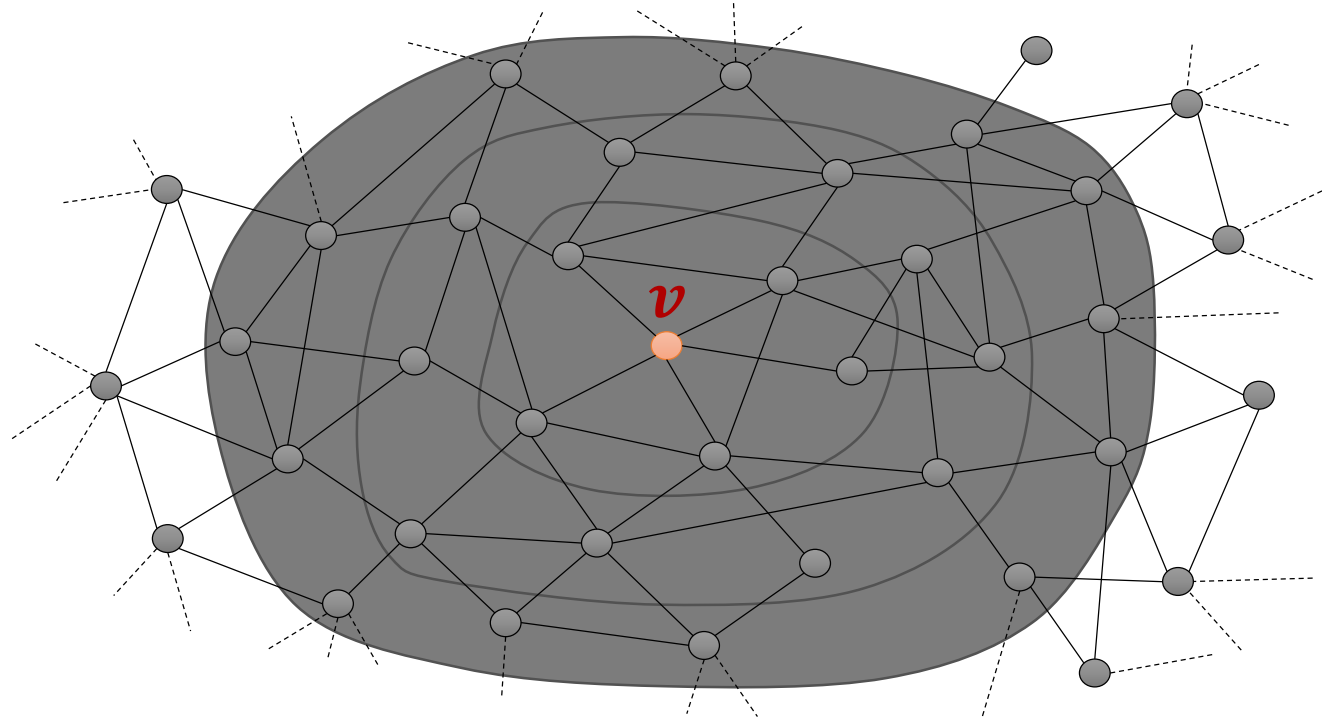
If we do not ignore log-factors, there is an exponential separation between randomized and deterministic LOCAL algorithm.

- *$\Delta$ -coloring trees* has randomized round complexity  $\Theta(\log \log n)$  and deterministic round complexity  $\Theta(\log n)$
- *Sinkless-orientation* has randomized round complexity  $\Theta(\log \log n)$  and deterministic round complexity  $\Theta(\log n)$

# Challenges in the LOCAL Model

# Challenges in the LOCAL Model

## (1) Locality:



In any  $r$ -round Algorithm, each node computes its output as a function of the initial state of its  $r$ -neighborhood.

# Challenges in the LOCAL Model

## (2) Local Coordination / Symmetry Breaking

- Nearby (symmetric) nodes need to **output different values**
    - E.g., different colors
  - Nodes **decide in parallel** based on their  $r$ -neighborhoods
  - Need **local coordination** among nearby nodes
- ✓ Randomization naturally helps
- E.g., choose random color, keep if no conflict with neighbors

SLOCAL:

A sequential variant of  
the LOCAL Model

# Sequential LOCAL Model



## SLOCAL Model

- **locality**  $r(n)$
- **sequentially** go over all nodes  $v_1, v_2, \dots, v_n$  (an arbitrary given order)
- compute **output** of each node based on the current state of its  $r(n)$ -**neighborhood**

# SLOCAL Model

The SLOCAL model is much more powerful than LOCAL model

- $(\Delta + 1)$ -coloring and MIS can easily be solved with locality 1
  - The sequential greedy algorithm is an SLOCAL-algorithm.
  - The output a node  $v$  only depends on the outputs of neighbors that were processed before  $v$ .
- SLOCAL is a generalization of sequential greedy algorithms
  - if for each node, one only looks at previous nearby nodes

# Complexity Classes

**LOCAL**( $t(n)$ ): graph problems that can be solved **deterministically** in  $t(n)$  rounds in the **LOCAL** model

**SLOCAL**( $t(n)$ ): graph problems that can be solved **deterministically** with **locality**  $t(n)$  in the **SLOCAL** model,

– e.g., MIS,  $(\Delta + 1)$ -coloring  $\in$  SLOCAL(1)

**P-LOCAL** := **LOCAL**(poly log  $n$ )

**P-SLOCAL** := **SLOCAL**(poly log  $n$ )

**Randomized classes:**

RLOCAL, RSLOCAL, P-RLOCAL, P-RSLOCAL



# Relations between the Complexity Classes

**Basic:**  $\text{LOCAL}(t(n)) \subseteq \text{SLOCAL}(t(n))$ ,  $\text{P-LOCAL} \subseteq \text{P-SLOCAL}$

**Fact 1: P-SLOCAL  $\subseteq$  P-RLOCAL**

- randomized poly  $\log n$ -round distributed alg. for all problems in P-SLOCAL

**Fact 2: P-SLOCAL  $\subseteq$  LOCAL( $2^{O(\sqrt{\log n})}$ )**

- deterministic  $2^{O(\sqrt{\log n})}$ -round distributed alg. for all problems in P-SLOCAL

**Open Problem: P-LOCAL  $\stackrel{?}{=} \text{P-SLOCAL}$**



Proofs via  
Network Decompositions

# P-SLOCAL Completeness

Problems in P-SLOCAL that if proven to be in P-LOCAL, imply **P-SLOCAL = P-LOCAL**.

# P-SLOCAL Completeness

**Local Reduction:** We say that a distr. graph problem  $P_1$  is **polylog  $n$ -reducible** to  $P_2$  if a deterministic **polylog  $n$ -round** distr. algorithm for  $P_2$  implies a deterministic **polylog  $n$ -round** distr. algorithm for  $P_1$ .

**P-SLOCAL Completeness:** A problem  $P$  in P-SLOCAL is called **P-SLOCAL-complete** if **every problem  $P'$**  in P-SLOCAL is **polylog  $n$ -reducible** to  $P$

**Example:  $(O(\log n), O(\log n))$ -decomposition is P-SLOCAL-complete**

- $(O(\log n), O(\log n))$ -decomp is in  $\text{SLOCAL}(O(\log^2 n))$
- polylog  $n$  round decomposition alg.  $\implies$  polylog  $n$  round P-SLOCAL alg.

# Local Splitting: A Simple Yet Complete Problem

## $\lambda$ -Local Splitting for $\lambda \in (0, \frac{1}{2})$ :

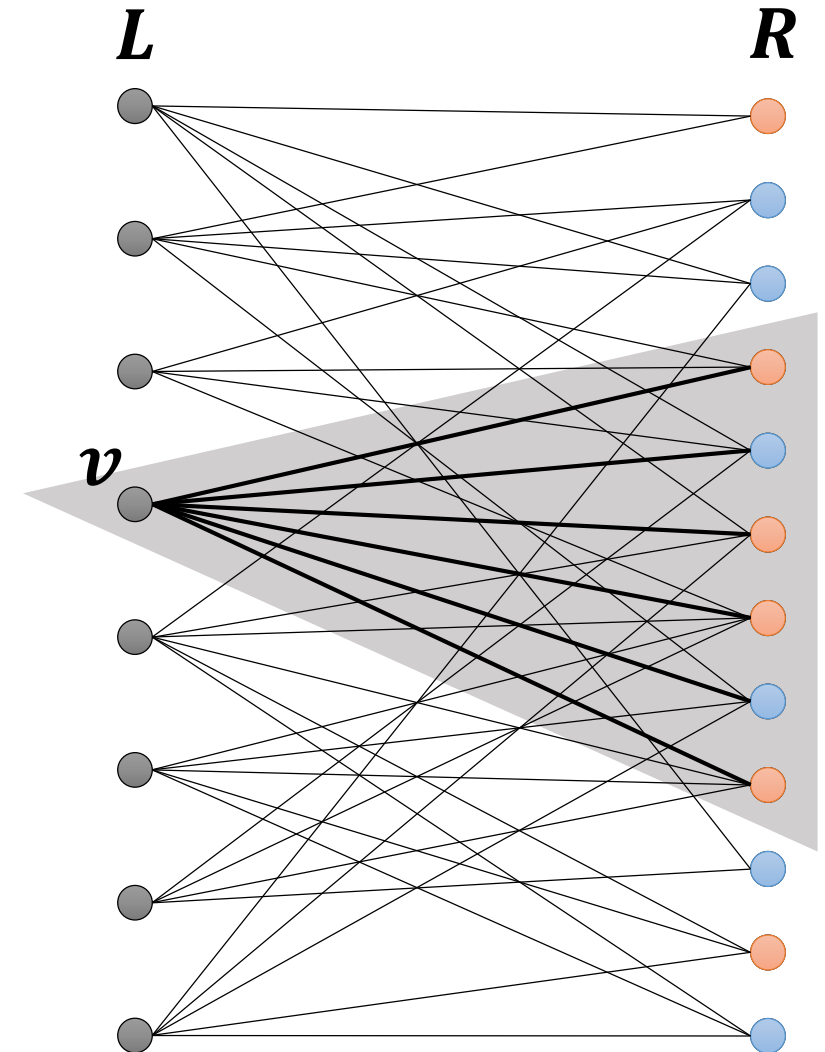
- Color R red/blue s.t. each  $v \in L$  with  $\deg(v) = \Omega(\log n)$  has at least  $\lfloor \lambda \deg(v) \rfloor$  neighbors in each color.

## Weak Local Splitting:

- Every  $v \in L$  with  $\deg(v) = \Omega(\log n)$  has at least one neighbor in each color.

## Trivial Randomized Solution:

- Independently color each node red/blue with probability  $1/2$ .

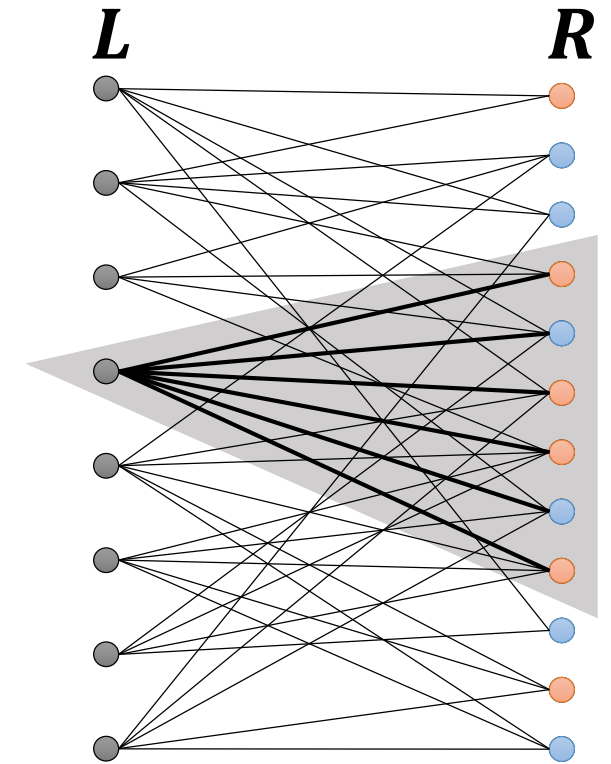


# (Weak) Local Splitting is PSLOCAL-Complete

**Theorem:** Weak local splitting for bipartite graph where all nodes in  $L$  have a large polylogarithmic degree --- say  $\Theta(\log^{10} n)$  --- is P-SLOCAL-complete.

- It can be seen as a rounding fractional (1/2) values to integer values (0 or 1), while preserving some linear constraints.

**Take-Home Message:** Rounding fractional values to integer values, while coarsely preserving some linear constraints, is the only obstacle to obtaining efficient (polylog  $n$ -time) deterministic LOCAL algorithms.



# (Weak) Local Splitting is PSLOCAL-Complete

**Theorem:** Weak local splitting for bipartite graph where all nodes in  $L$  have a large polylogarithmic degree --- say  $\Theta(\log^{10} n)$  --- is P-SLOCAL-complete.

Proof.

network decomposition

polylog-reducible

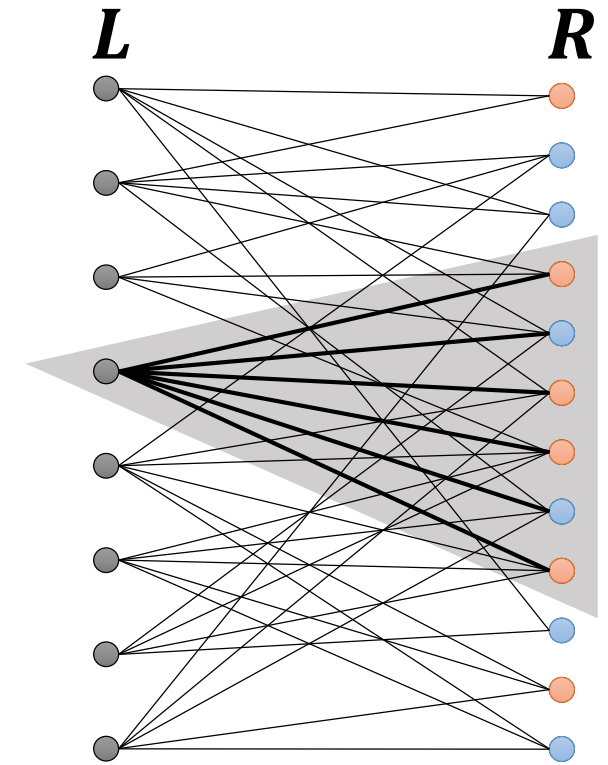
conflict-free coloring

polylog-reducible

$\lambda$ -local splitting

polylog-reducible

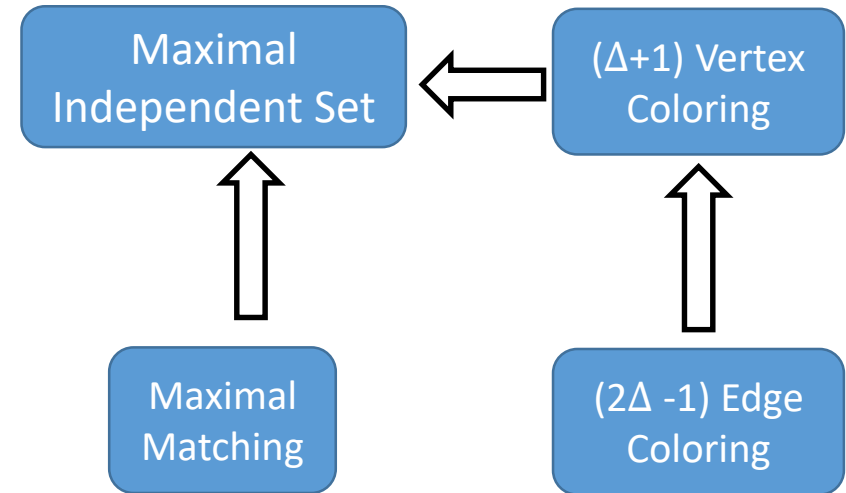
weak local splitting



# DET LOCAL Algorithms

## via Rounding

1. Maximal Matching
2.  $(2\Delta - 1)$  edge coloring



# Maximal Matching via Rounding

## THEOREM [Fischer'17]

There is a  $O(\log^2 \Delta \cdot \log n)$ -round deterministic algorithm for maximal matching.

$O(\log^4 n)$

Hańkowiak, Karoński, Panconesi [SODA'98, PODC'99]

## Algorithm Outline (Core Part):

$O(1)$  - Approximate Bipartite Matching

$O(\log^2 \Delta)$  rounds

- **4 - Approximate Fractional Matching**

$O(\log \Delta)$  rounds

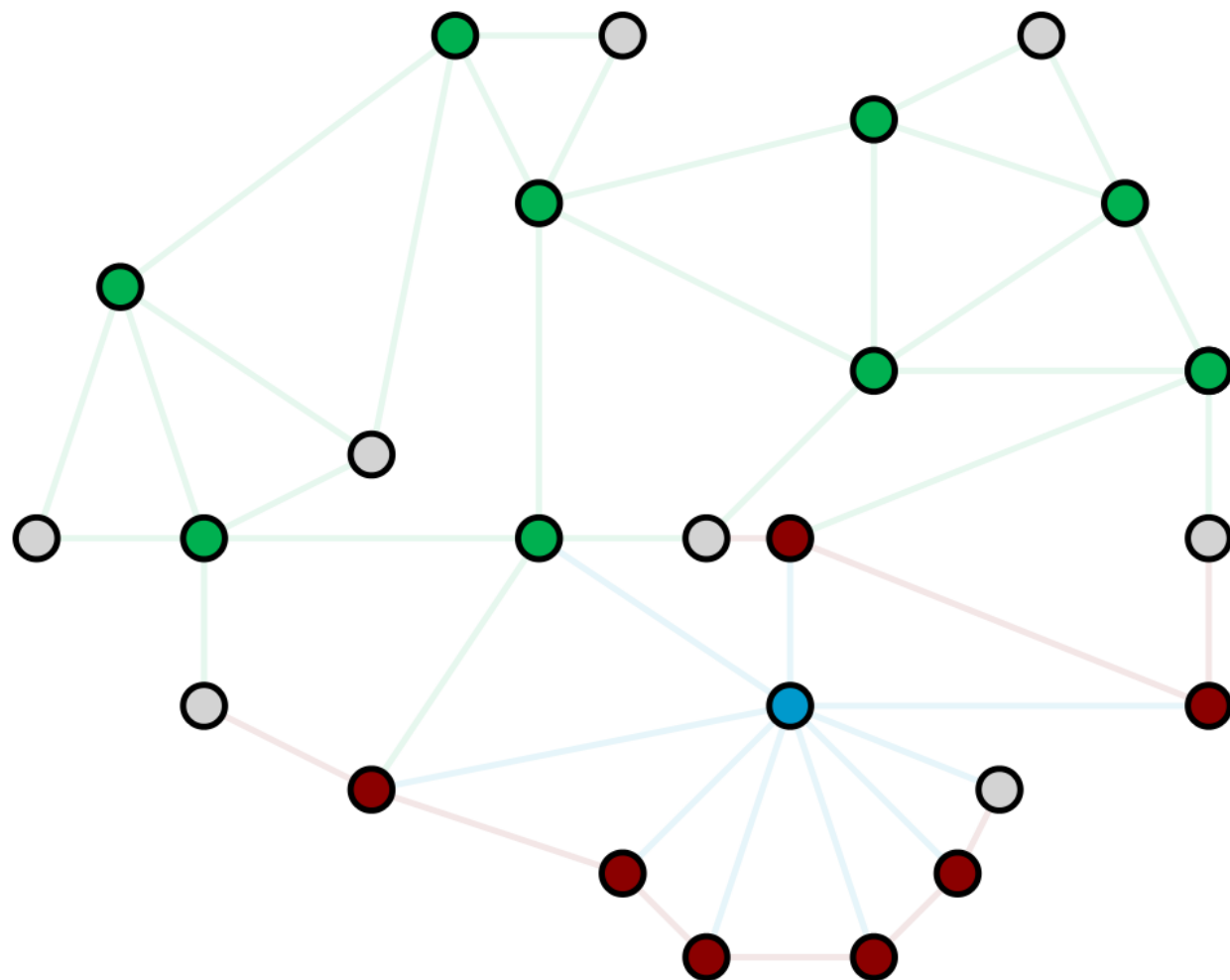
- **Rounding Fractional Bipartite Matching**

$O(\log^2 \Delta)$  rounds,  $O(1)$  los



# I) 4- Approximate Fractional Matching

$O(\log \Delta)$  rounds



$$\frac{1}{16} \quad \frac{1}{8} \quad \frac{1}{4}$$

## Fractional Maximum Matching

$$\begin{aligned} & \max \sum_{e \in E} x_e \\ & \text{s.t. } \sum_{e \in E(v)} x_e \leq 1 \quad \text{for all } v \in V \\ & \quad x_e \in [0, 1] \quad \text{for all } e \in E \end{aligned}$$

**v is half-tight if its value is  $\geq \frac{1}{2}$**

## LOCAL GREEDY ALGORITHM

- $x_e = 2^{-\lceil \log \Delta \rceil}$  for all  $e \in E$
- repeat until all edges are blocked
- mark half-tight nodes
- block its edges
- double value of unblocked edges

## Constant - Approximate Bipartite Matching

$O(\log^2 \Delta)$  rounds

### I) 4-Approximate Fractional Matching

$O(\log \Delta)$  rounds

### II) Rounding Fractional Bipartite Matching

$O(\log^2 \Delta)$  rounds,  $O(1)$  loss

## II) Rounding Fractional Bipartite Matching

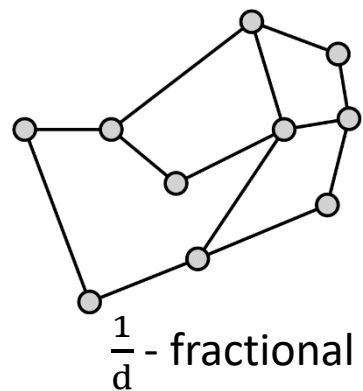
$O(\log^2 \Delta)$  rounds,  $O(1)$  loss

### Fractionality of a Matching

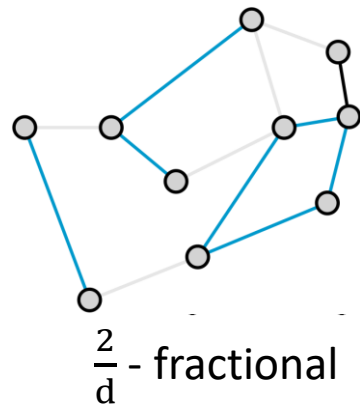
Matching is  $\frac{1}{d}$ -fractional if all non-zero values are  $\geq \frac{1}{d}$

1-fractional means integral

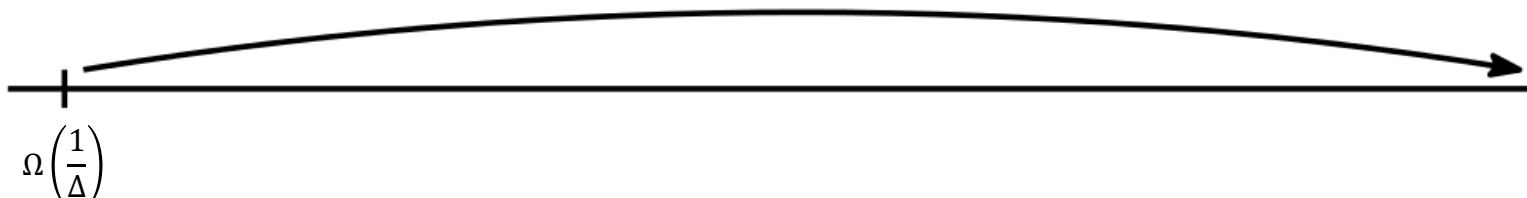
### Factor-2-Rounding



$\cup$



DIRECT ROUNDING



GRADUAL ROUNDING



## II) Rounding Fractional Bipartite Matching

$O(\log^2 \Delta)$  rounds,  $O(1)$  loss

### Iterated Factor-2-Rounding using Locally Balanced Splitting

#### Locally Balanced Splitting:

2-edge-coloring so that  
every node roughly balanced

for  $i = \lceil \log \Delta \rceil, \dots, 1$

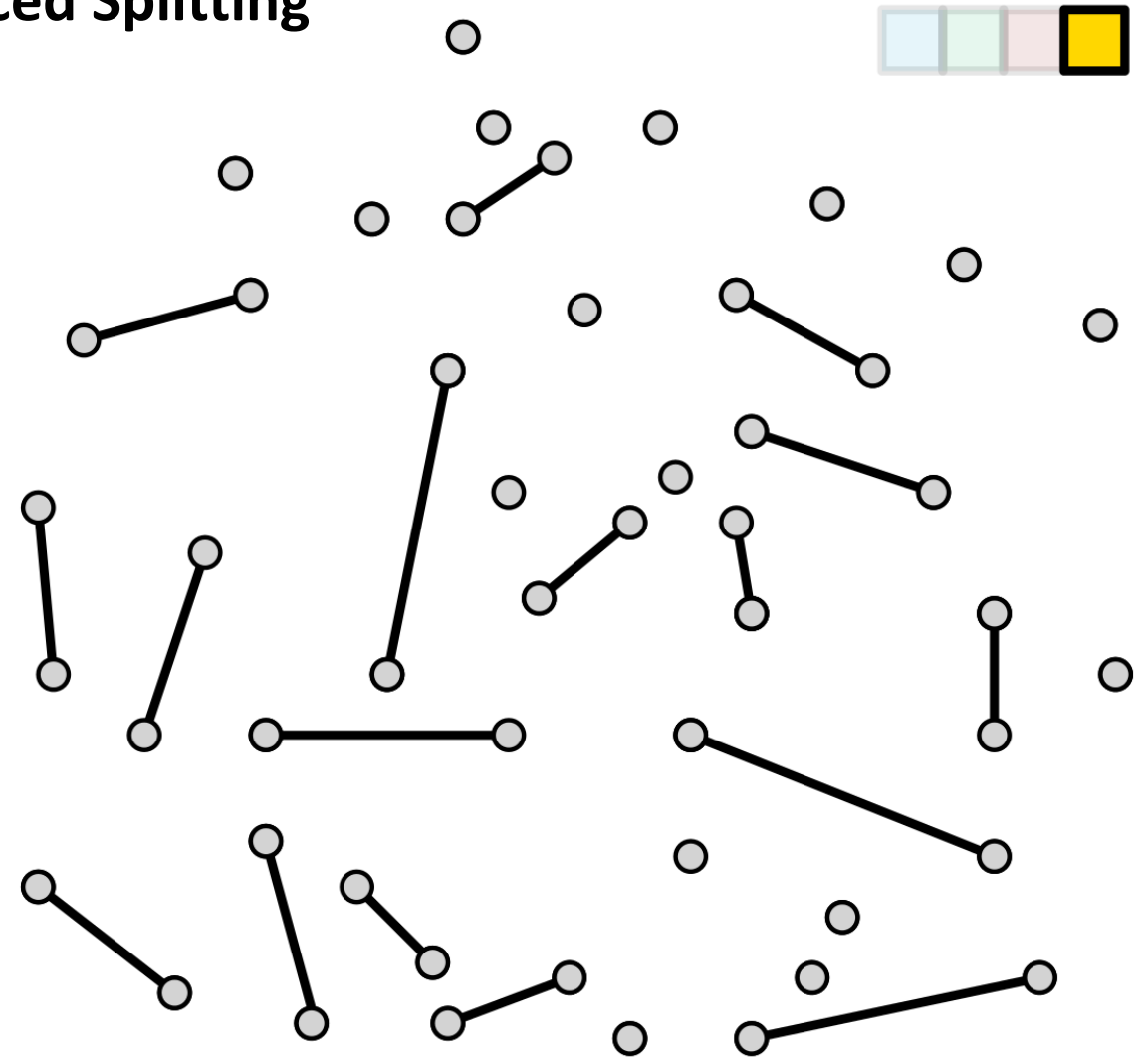
$$E_i = \{e \in E : x_e = 2^{-i}\}$$

splitting of  $E_i$  into  

increase  to  $2^{-i+1}$

decrease  to 0

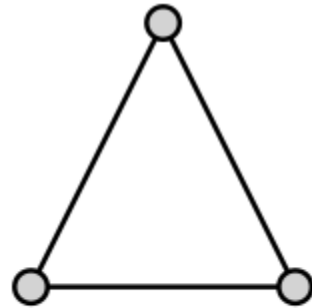
In case of perfect locally balanced splitting:  
no constraint violated & no loss in total value



## II) Rounding Fractional Bipartite Matching

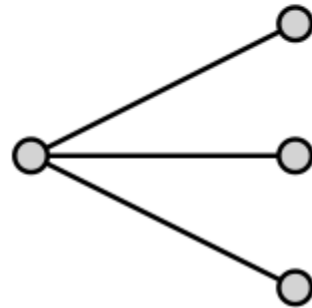
$O(\log^2 \Delta)$  rounds,  $O(1)$  loss

perfect splitting not possible in case of...



... odd cycles

bipartite graph!



... odd-degree vertices

small technicality.

Suppose that bipartite even-degree graph!

## II) Rounding Fractional Bipartite Matching

$O(\log^2 \Delta)$  rounds,  $O(1)$  loss

### LOCAL Almost-Perfect Splitting

Decompose into cycle

In parallel, for all cycles

A) Short cycles of length  $O(\log \Delta)$

alternate  

B) Long cycles

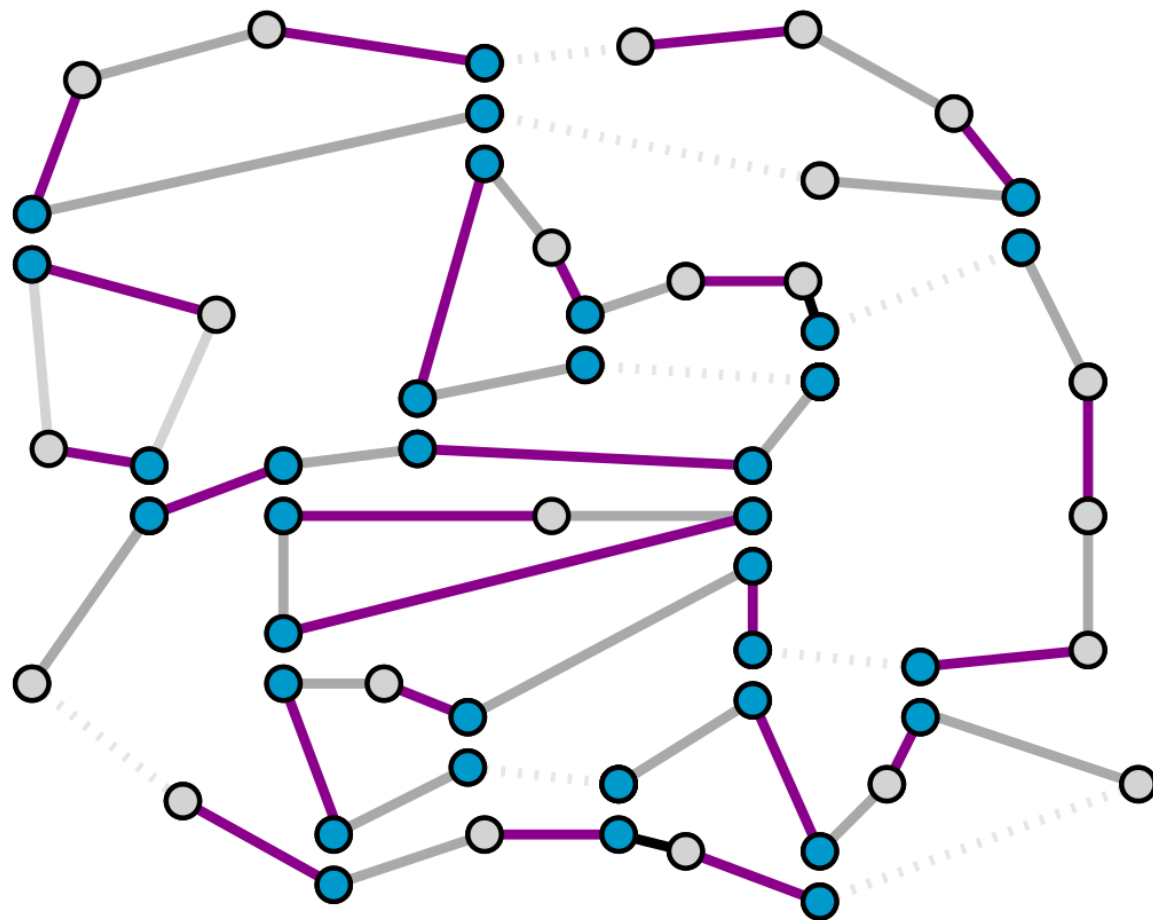
chop at length  $\Theta(\log \Delta)$

set boundary to 0

alternate   in between

$\Theta\left(\frac{1}{\log \Delta}\right)$  loss

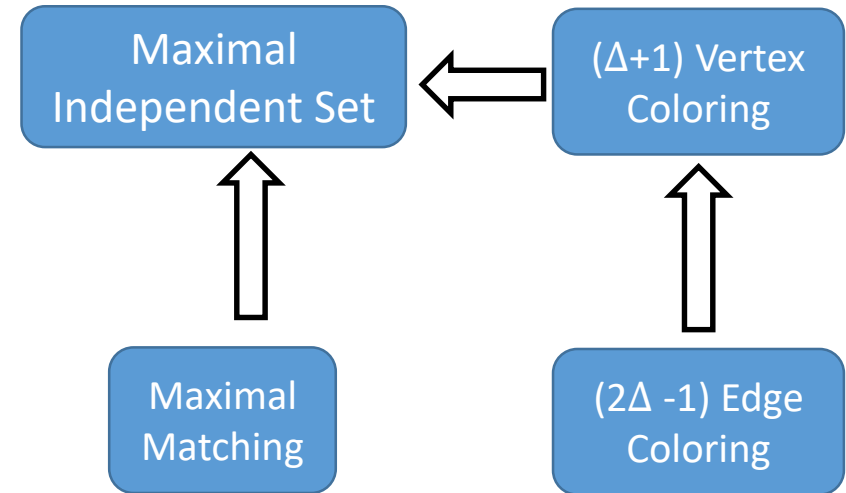
Over all  $O(\log \Delta)$  rounding iterations,  
the overall loss still a constant!



# DET LOCAL Algorithms

## via Rounding

1. Maximal Matching
2.  $(2\Delta - 1)$  edge coloring



# $(2\Delta - 1)$ edge coloring

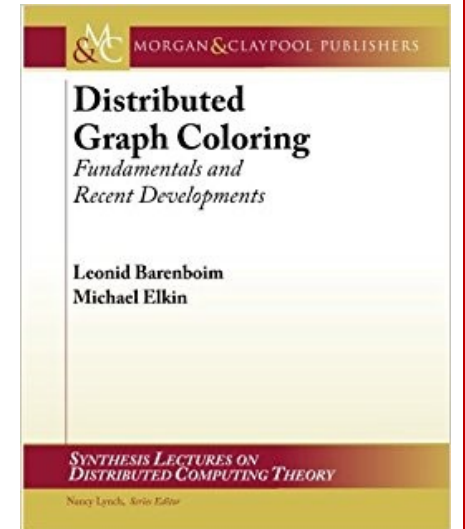
"While maximal matchings can be computed in polylogarithmic time [...], it is a decade old open problem whether the same running time is achievable for the remaining three structures."

Panconesi, Rizzi '01

## **Open Problem 11.4:**

Devise or rule out a deterministic  $(2\Delta - 1)$ -edge-coloring algorithm that runs in polylogarithmic time.

Barenboim, Elkin '13



- We resolve this problem and give a polylog  $n$  round algorithm for it.
- The solution goes via hypergraph maximal matching.



# Hypergraph Maximal Matching & Implications

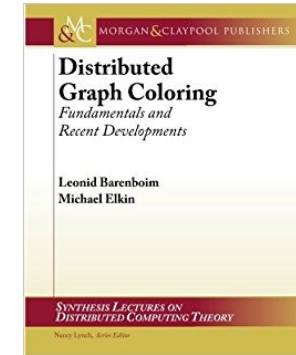
**Rank- $r$ -Hypergraph Maximal Matching**  
in  $\text{poly } r \cdot \log^{O(\log r)} \Delta \cdot \log n$  rounds



**$(2\Delta - 1)$ -Edge-Coloring**  
in  $O(\log^8 n)$  rounds

**Randomized**  
 **$(2\Delta - 1)$ -Edge-Coloring**  
in  $O(\log^8 \log n)$  rounds

**Maximal Independent Set and  $(\Delta + 1)$ -Vertex-Coloring**  
for graphs with bounded neighborhood independence



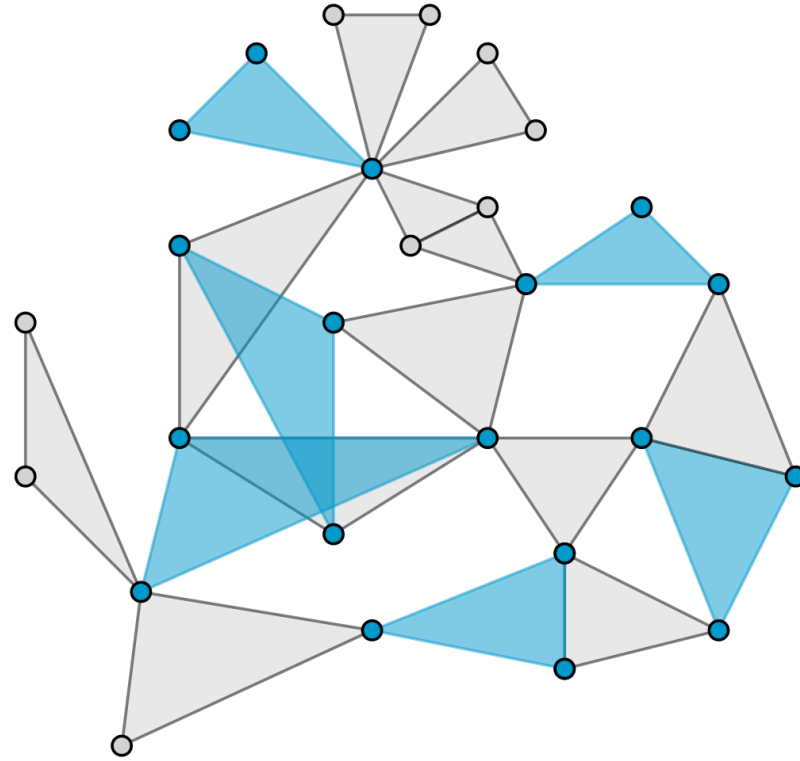
**Open Problem 11.4**

**Open Problem 11.5**

**I) Formulation as Hypergraph Maximal Matching**

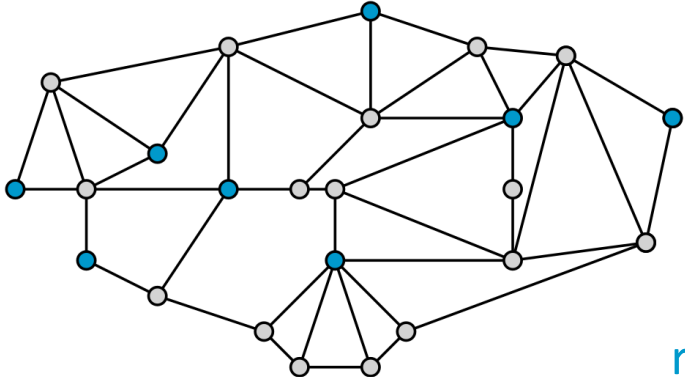
**II) Hypergraph Maximal Matching Algorithm**

# I) Formulation as Hypergraph Maximal Matching



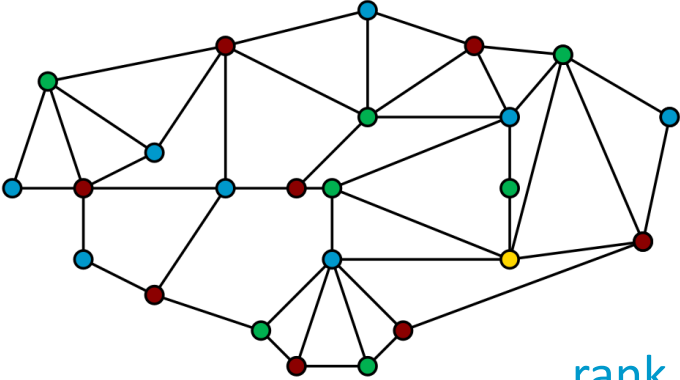
# Unified Formulation as Hypergraph Maximal Matching Problem

Maximal Independent Set



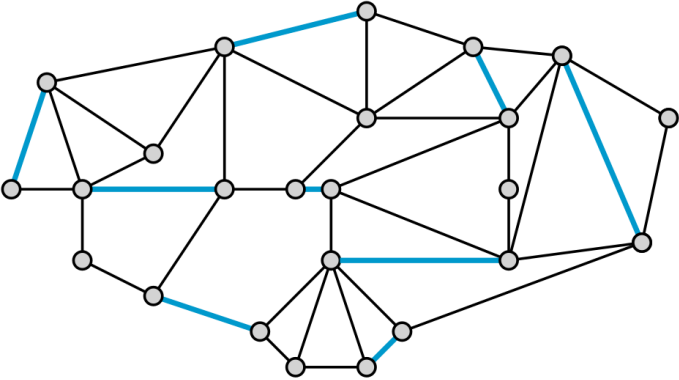
rank  $\Delta$

$(\Delta + 1)$ -Vertex-Coloring



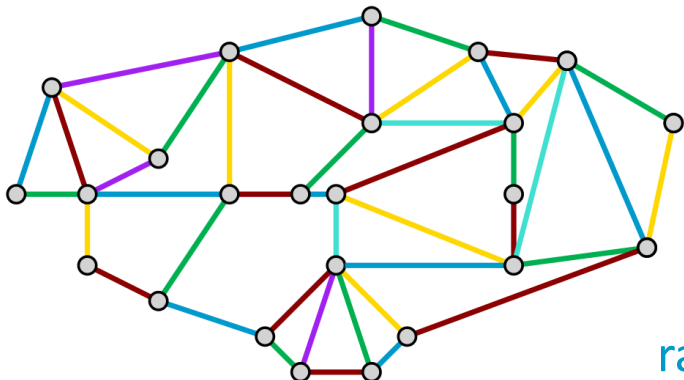
rank  $\Delta + 1$

Maximal Matching



rank 2

$(2\Delta - 1)$ -Edge-Coloring

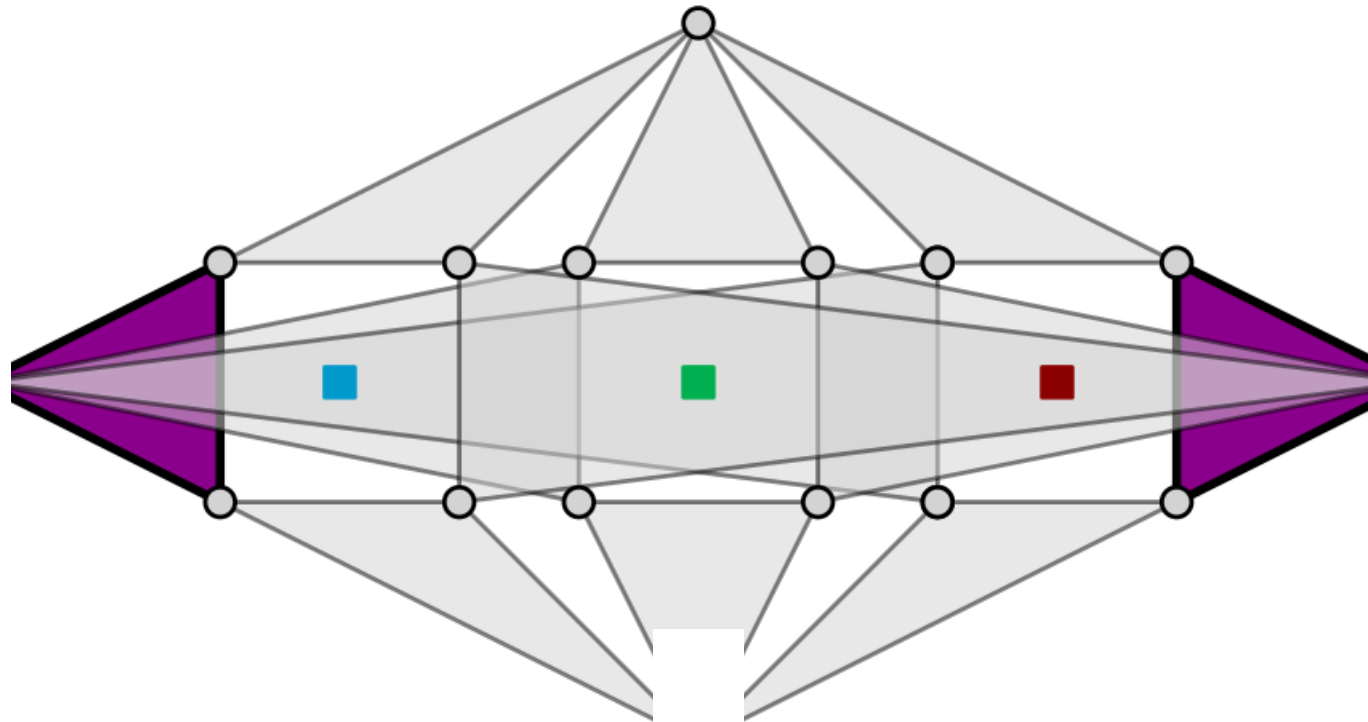


rank 3

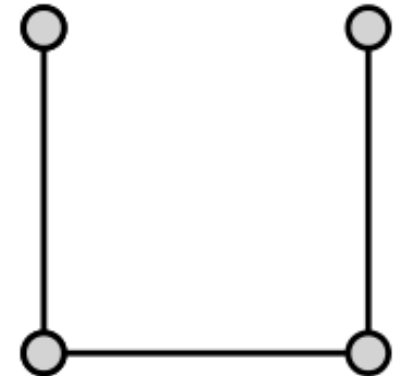
cast classic LOCAL graph problems as hypergraph maximal matching problems (LOCAL reductions)

# $(2\Delta - 1)$ -Edge-Coloring as Rank-3-Hypergraph Maximal Matching

## Rank-3-Hypergraph Maximal Matching



## Edge-Coloring of Graph



at the most one color per edge

## **$O(r^2)$ -Approximate Maximum Matching**

**$O(r)$ -Approximate Maximum Fractional Matching**

**Rounding Fractional Matching**

# Basic Rounding

## Sequential Greedy Factor-L-rounding

from  $\geq \frac{1}{d}$  to  $\geq \frac{L}{d}$

for all unblocked edges with value  $< \frac{L}{d}$

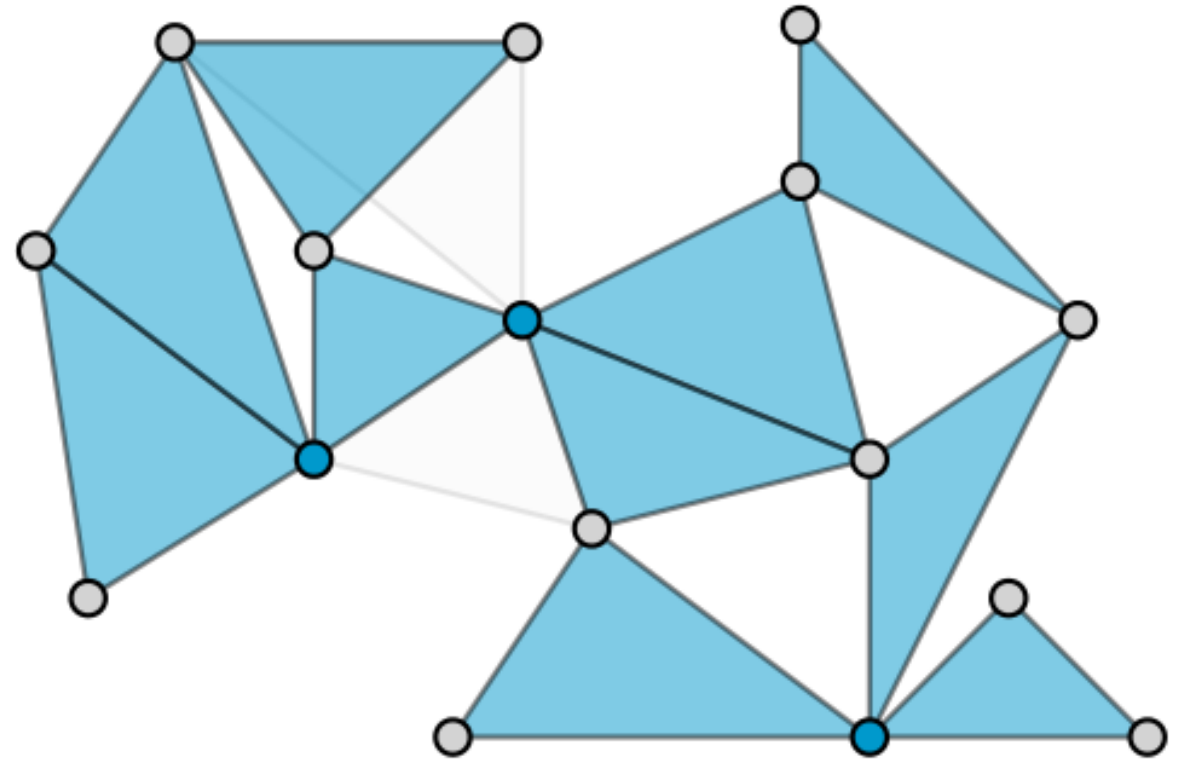
set value to  $\frac{L}{d}$

mark tight nodes

block their edges

for all blocked edges with value  $< \frac{L}{d}$

set value to 0



$O(r)$  loss

# Basic Rounding

Factor-L-Rounding in  $O(\log \Delta + (L \cdot r)^2)$  rounds with  $O(r)$  loss

## LOCAL Greedy Factor-L-Rounding

from  $\geq \frac{1}{d}$  to  $\geq \frac{L}{d}$

$\frac{d}{2L}$  - Defective  $O(L^2 r^2)$ -Edge-Coloring

for each color class

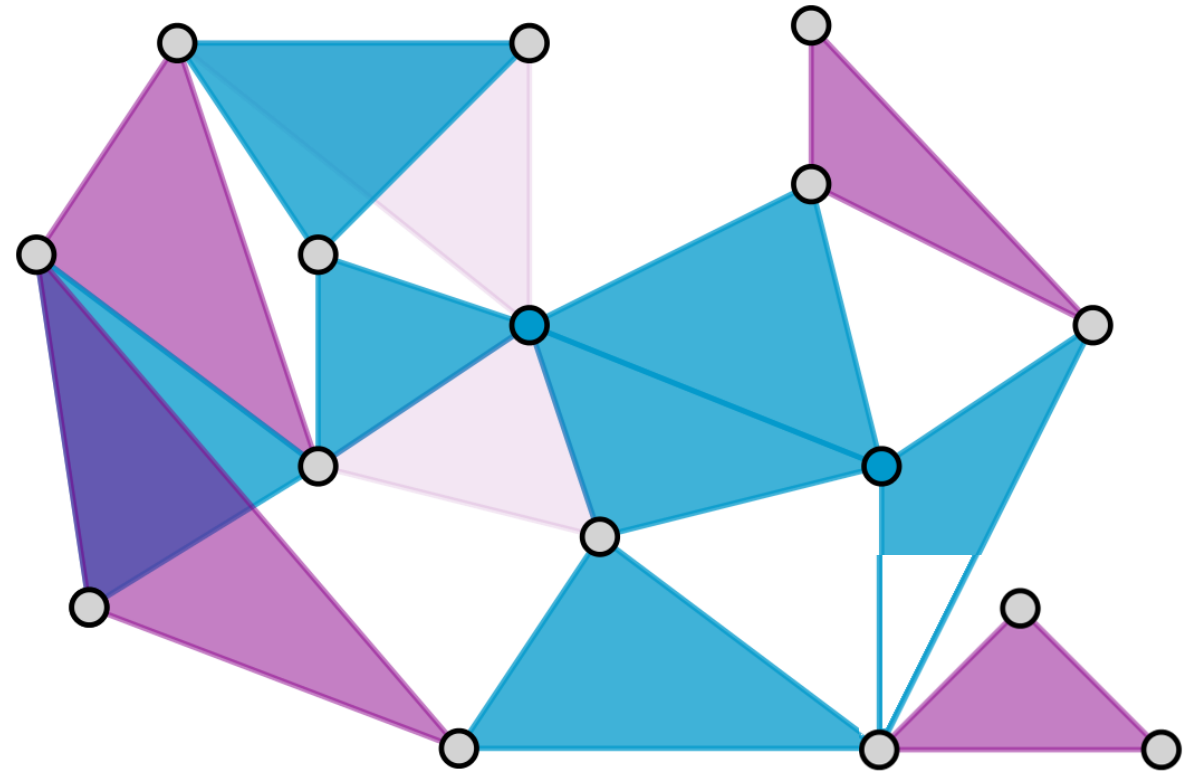
mark half-tight nodes

block their edges

set value of edges in color class to  $\frac{L}{d}$

for all blocked edges with value  $< \frac{L}{d}$

set value to 0



In each step, value of a node increased by at most  $+\frac{d}{2L} \cdot \frac{L}{d} = \frac{1}{2}$

$O(\log \Delta + (L \cdot r)^2)$  rounds using Defective-Coloring Algorithm by Kuhn [SPAA'09]



# Rounding

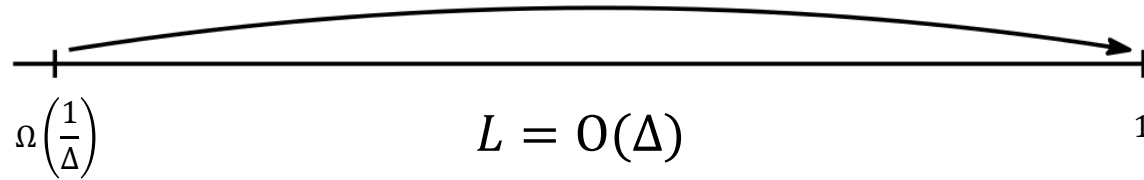
## Basic Rounding:

Factor-L-Rounding in  $O(\log \Delta + (L \cdot r)^2)$  rounds with  $O(r)$  loss

### Direct Basic Rounding

$O(r)$  loss

$O(\log \Delta + (\Delta \cdot r)^2)$  rounds

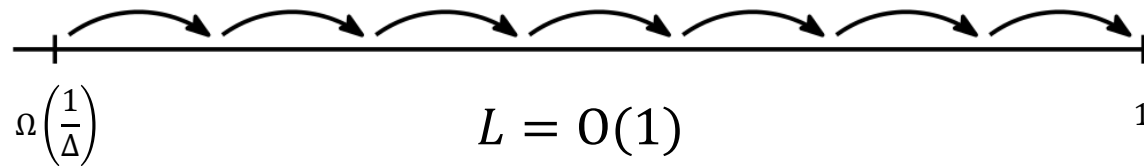


too slow!

### Iterative Basic Rounding

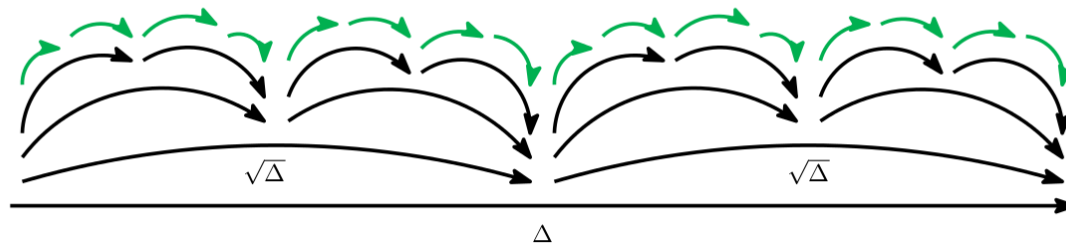
$r^{\Theta(\log \Delta)}$  loss

$O(\log \Delta \cdot (\log \Delta + r^2))$  rounds



too lossy!

### Recursive Rounding & Iterative Refilling



Basic Rounding

Invariant:  $O(r)$  loss, maintained via refilling iterations

# Further Improvements & Open Problems

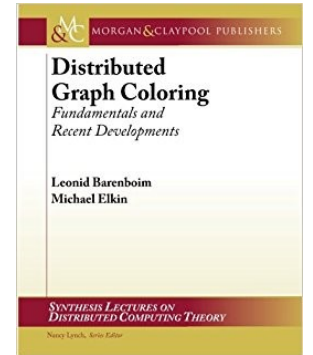
# Further Improvements (a general derandomization recipe)

- **Rank- $r$ -Hypergraph Maximal Matching** in poly  $(r \cdot \log n)$  rounds

✓  **$(a(1 + \epsilon))$  Out-Degree Orientation** in poly  $(\log n/\epsilon)$

✓  **$(1 + \epsilon)$ -Approximation of Matching** in poly  $(\log n/\epsilon)$

**Open Problem 11.10**



- **$((1 + \epsilon)\Delta)$ -Edge Coloring** in poly  $(\log n/\epsilon)$  rounds, assuming  $\Delta = \Omega_\epsilon(\log n)$

- Faster algorithms for the Lovasz Local Lemma

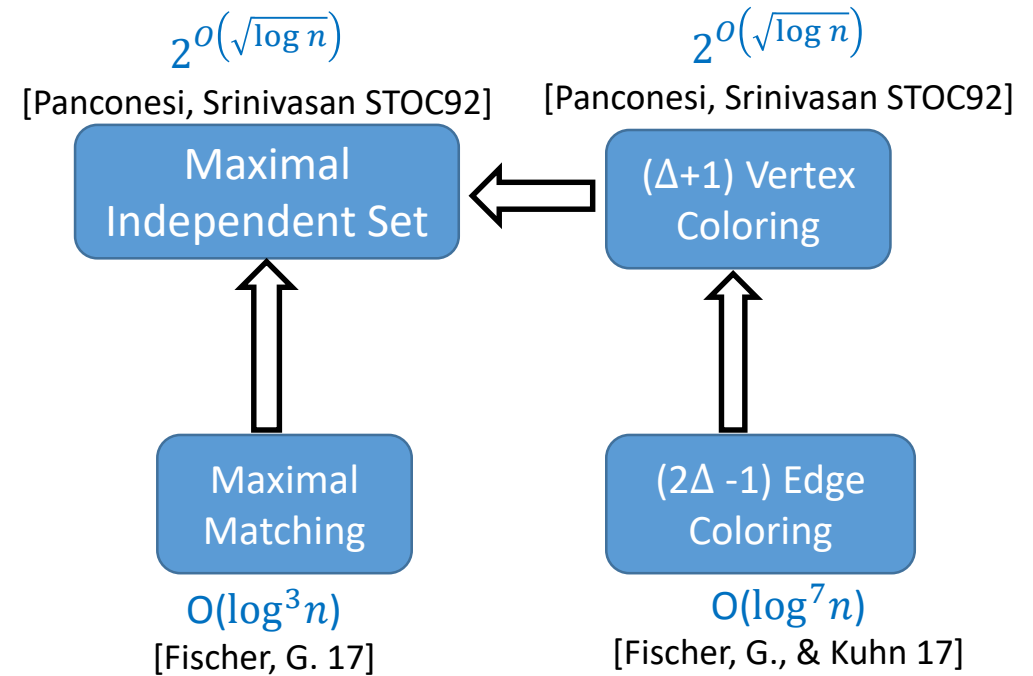
- For LCL problems, **P-SLOCAL = P-RSLOCAL**

- ...

# Open Problems

The SLOCAL model & Rounding as keys towards efficient DET LOCAL Algo.

- Linial's Q.: Is either of MIS or  $(\Delta + 1)$ -vertex-coloring in P-LOCAL?
- Are they P-SLOCAL-complete?
- Solve splitting/rounding for  $r = \log^{\omega(1)} n$



Thanks!