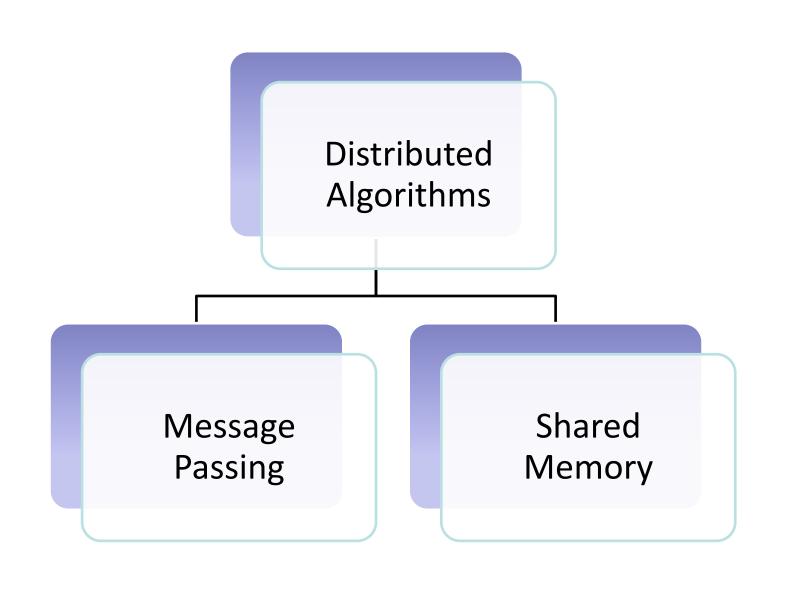
Distributed Algorithms

Tutorial



Roger Wattenhofer

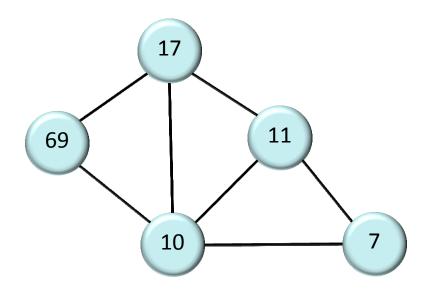




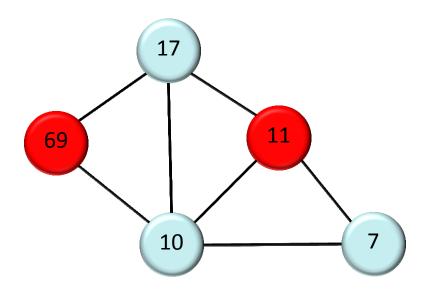
THE 1555



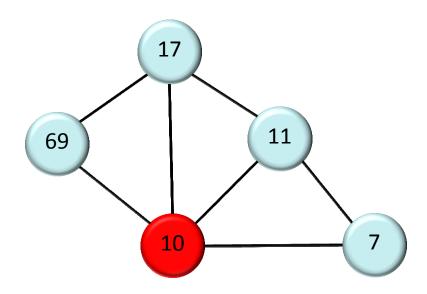
- Given a network with *n* nodes, nodes have unique IDs.
- Find a Maximal Independent Set (MIS)
 - a non-extendable set of pair-wise non-adjacent nodes



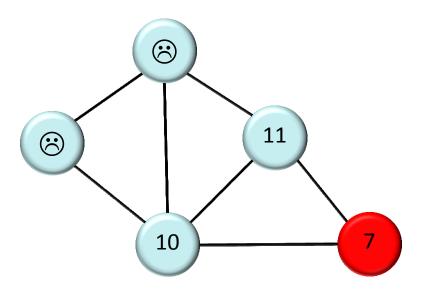
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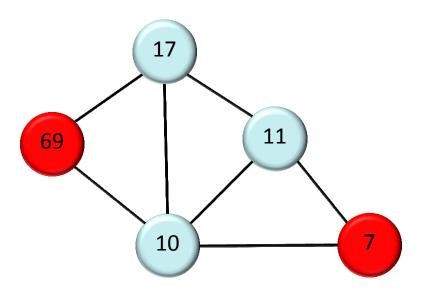
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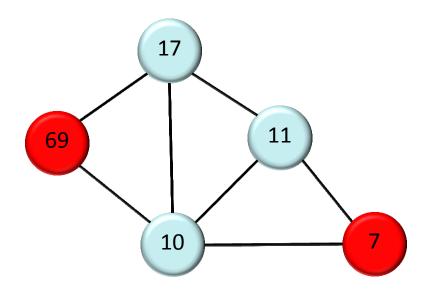
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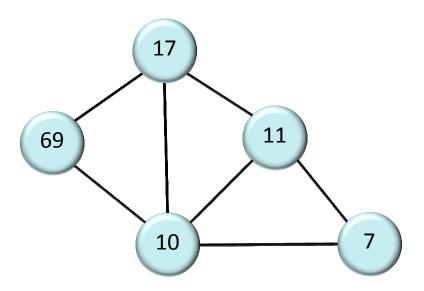


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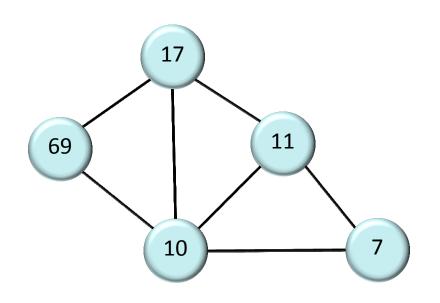


Traditional (sequential) computation:
 The simple greedy algorithm finds MIS (in linear time)

 Nodes are agents with unique ID's that can communicate with neighbors by sending messages. In each synchronous round, every node can send a (different) message to each neighbor.



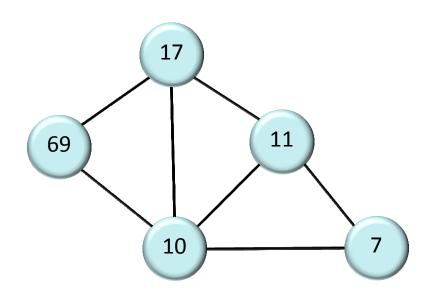
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each round:
every node:
1. send msgs
2. rcv msgs
3. compute

A Simple Distributed Algorithm

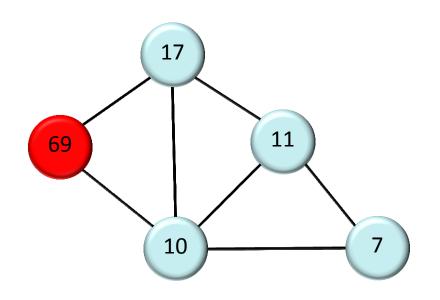
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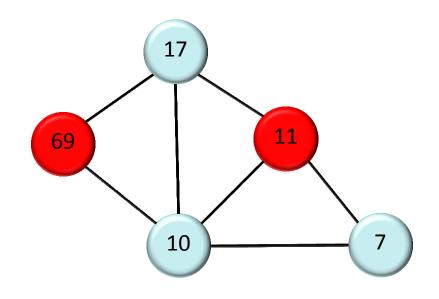
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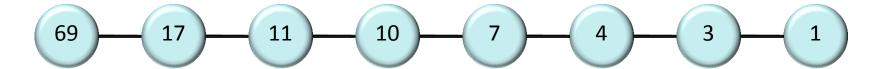
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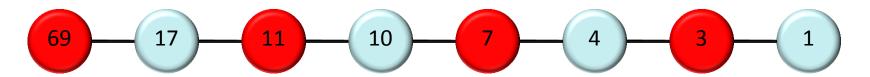
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What's the problem with this distributed algorithm?

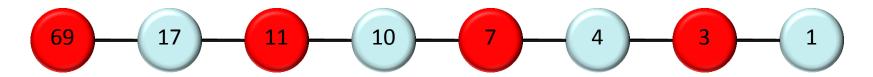
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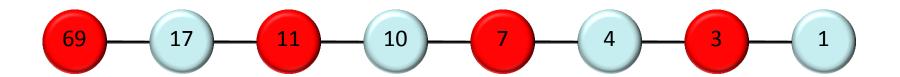


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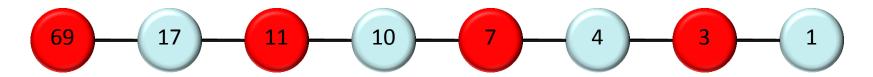


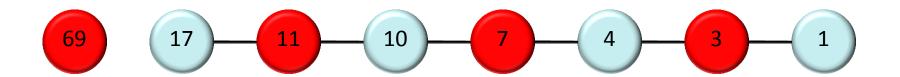
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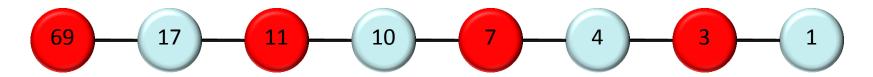


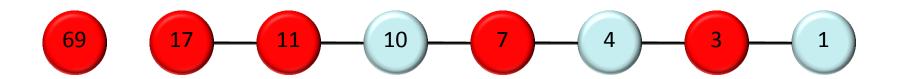
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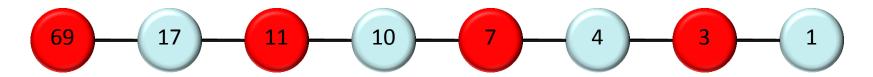


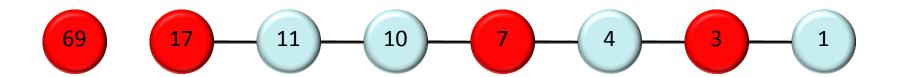
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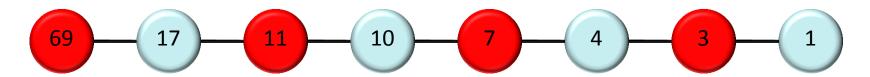


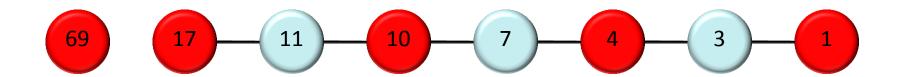
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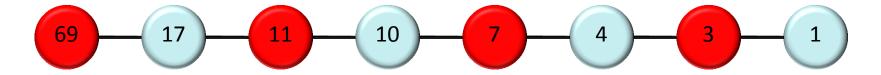


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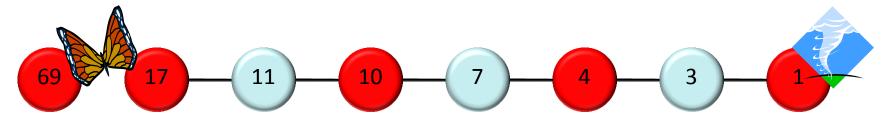




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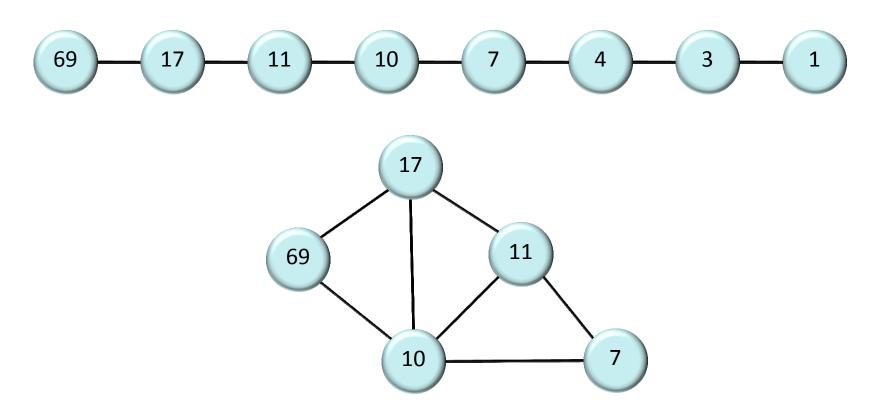


What if we have minor changes?

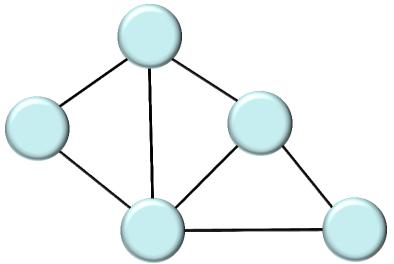


 Proof by animation: In the worst case, the algorithm is slow (linear in the number of nodes). In addition, we have a terrible "butterfly effect".

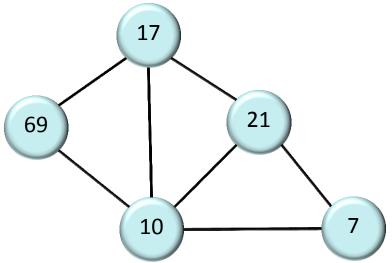
 Can you find a distributed algorithm that is polylogarithmic in the number of nodes n, for any graph?



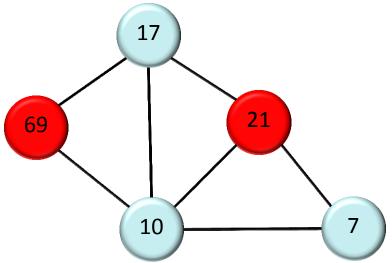
- Surprisingly, for deterministic distributed algorithms, this is an problem!
- However, randomization helps! In each synchronous round, nodes should choose a random value. If your value is larger than the value of your neighbors, join MIS!



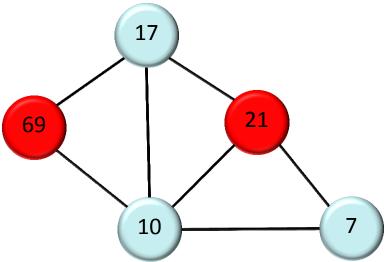
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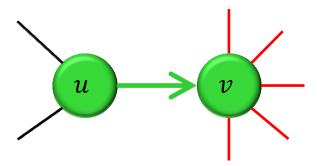
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How many synchronous rounds does this take in expectation (or whp)?

Analysis

- Event $(u \rightarrow v)$: node u got largest random value in combined neighborhood $N_u \cup N_v$.
- We only count edges of v as deleted.



- Similarly event $(v \rightarrow u)$ deletes edges of u.
- We only double-counted edges.
- Using linearity of expectation, in expectation at least half of the edges are removed in each round.
- In other words, who it takes $O(\log n)$ rounds to compute an MIS.

Results: MIS

General Graphs, Randomized [Alon, Babai, and Itai, 1986] [Israeli and Itai, 1986] [Luby, 1986] [Métivier et al., 2009]

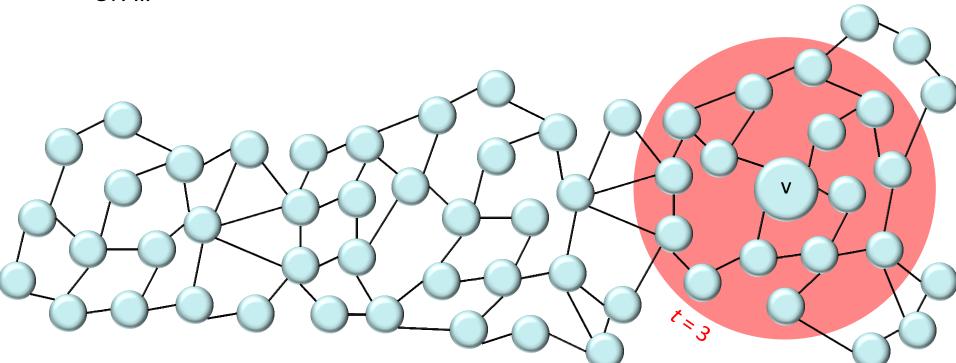
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Naïve Algo

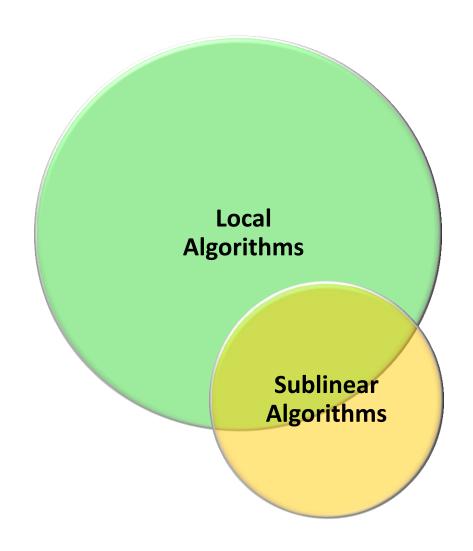
 $1 \qquad \qquad \log^* n \qquad \qquad \log n \qquad \qquad n^{\epsilon} \qquad \qquad n$

Local Algorithms

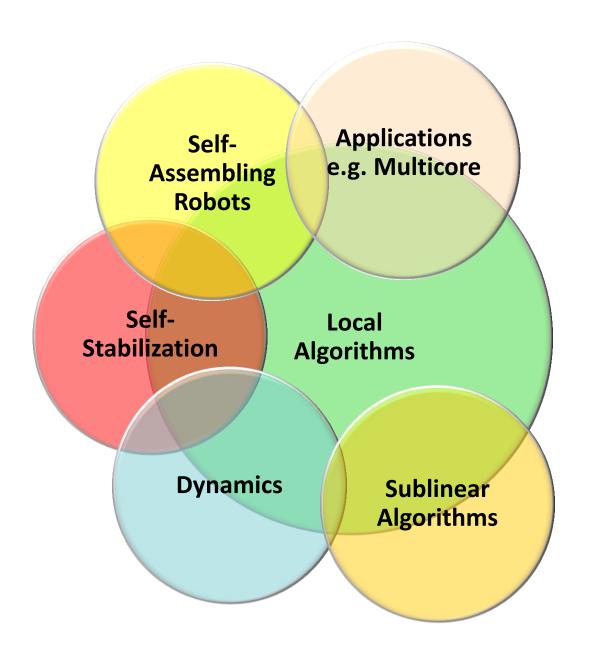
- Each node can exchange a message with all neighbors, for t communication rounds, and must then decide.
- Or: Given a graph, each node must determine its decision as a function of the information available within radius t of the node.
- Or: Change can only affect nodes up to distance t.
- Or: ...



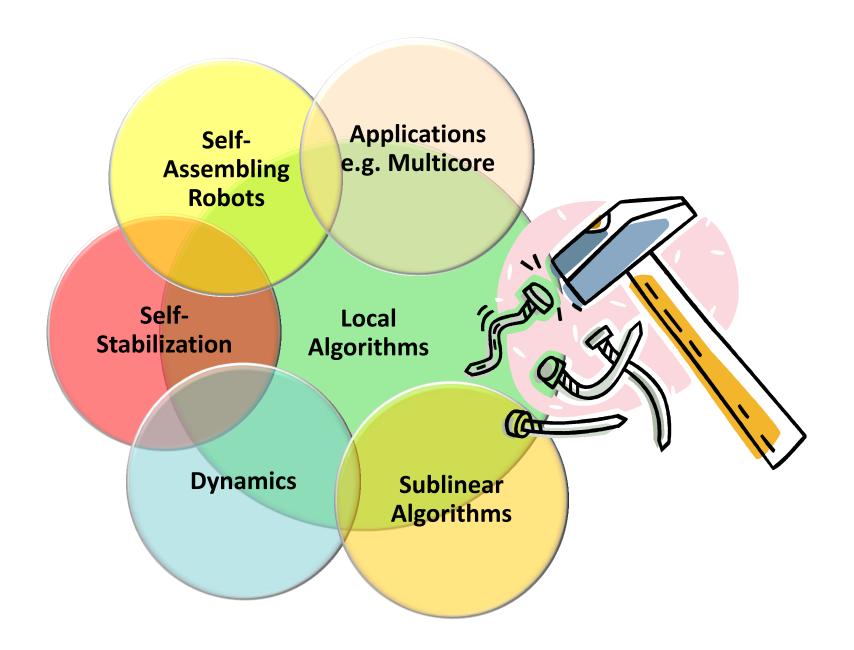
Locality

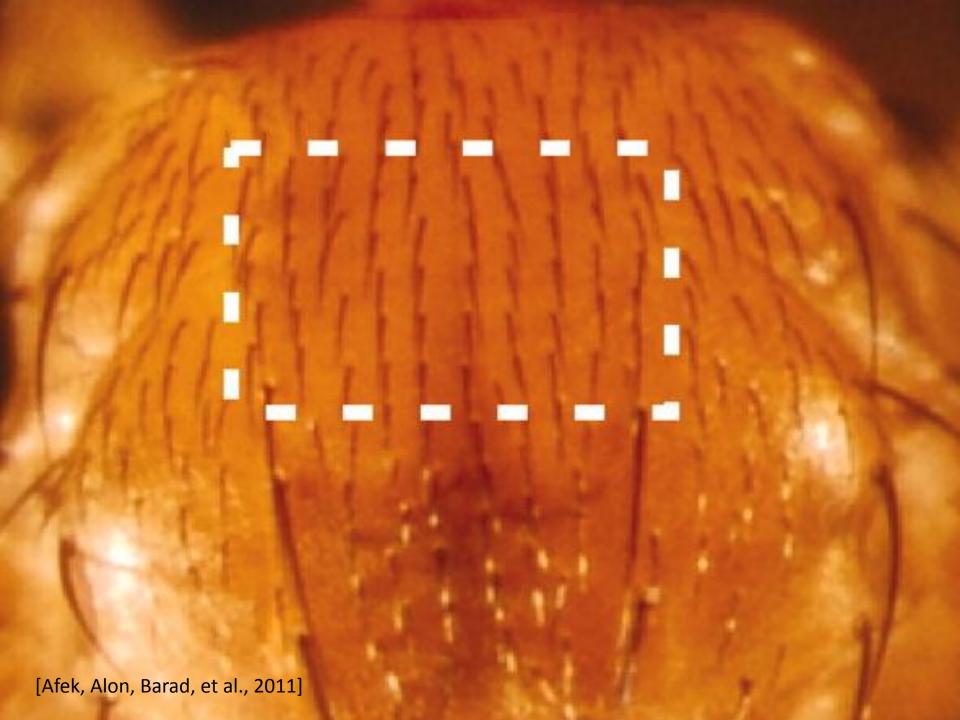


Locality is Everywhere!

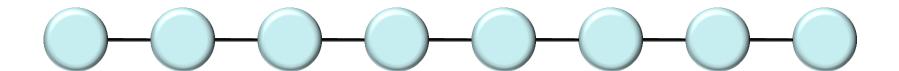


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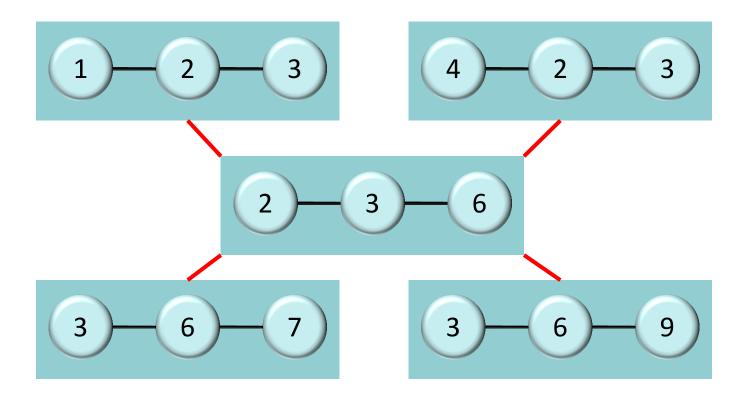


- Since the 1980s, nobody was able to improve this simple algorithm.
- What about lower bounds?
- There is an interesting lower bound, essentially using a Ramsey theory argument, that proves that an MIS needs at least $\Omega(\log^* n)$ time.
 - log* is the so-called iterated logarithm how often you need to take the logarithm until you end up with a value smaller than 1.
 - This lower bound already works on simple networks such as the linked list



Coloring Lower Bound on Oriented Ring

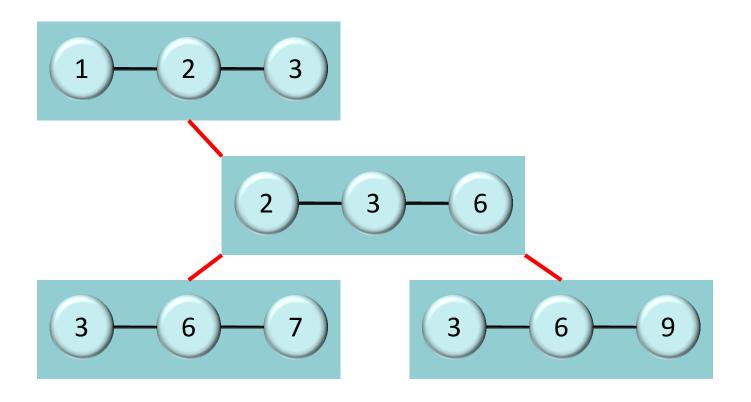
- Build graph G_t , where nodes are possible views of nodes for distributed algorithms of time t. Connect views that could be neighbors in ring.
- Here is for instance of G_1 :



• Chromatic number of G_t is exactly minimum possible colors in time t.

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Linked List [Linial, 1992]

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 $|IS(N_2)| \in O(1)$

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Other problems e.g., [Kuhn et al., 2006]

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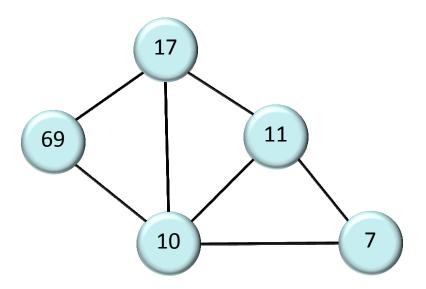
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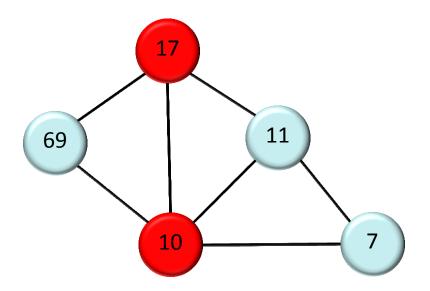
Example: Minimum Vertex Cover (MVC)

- Given a network with *n* nodes, nodes have unique IDs.
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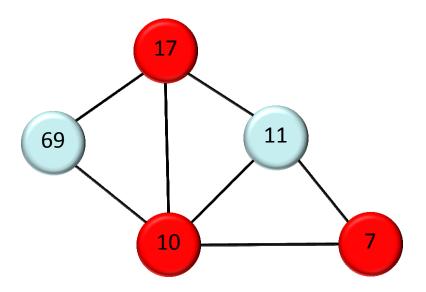
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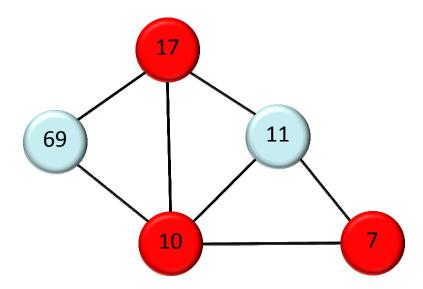
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Differences between MIS and MVC

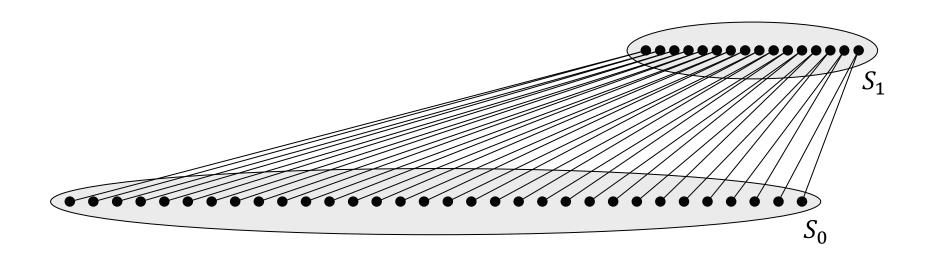
- Central (non-local) algorithms: MIS is trivial, whereas MVC is NP-hard
- Instead: Find an MVC that is "close" to minimum (approximation)
- Trade-off between time complexity and approximation ratio



- MVC: Various simple (non-distributed) 2-approximations exist!
- What about distributed algorithms?!?

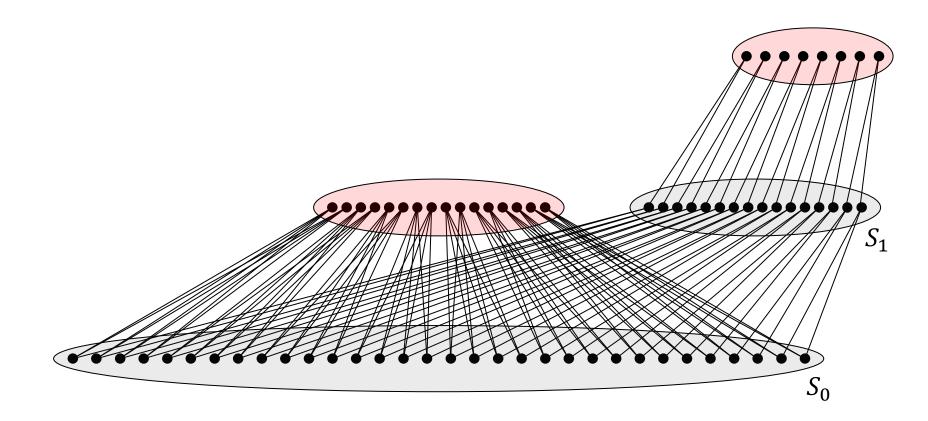
Finding the MVC (by Distributed Algorithm)

- Given the following bipartite graph with $|S_0| = \delta |S_1|$
- The MVC is just all the nodes in S_1
- Distributed Algorithm...

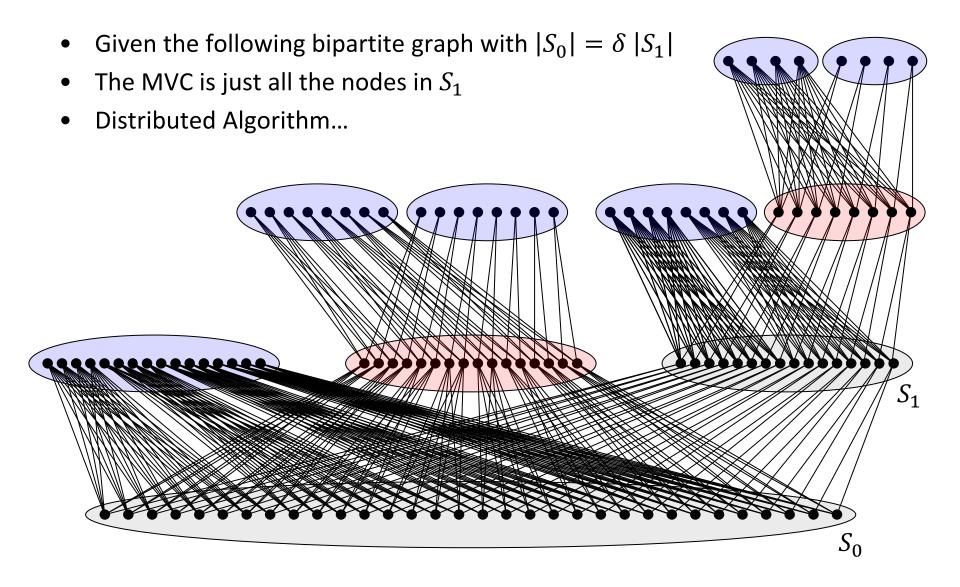


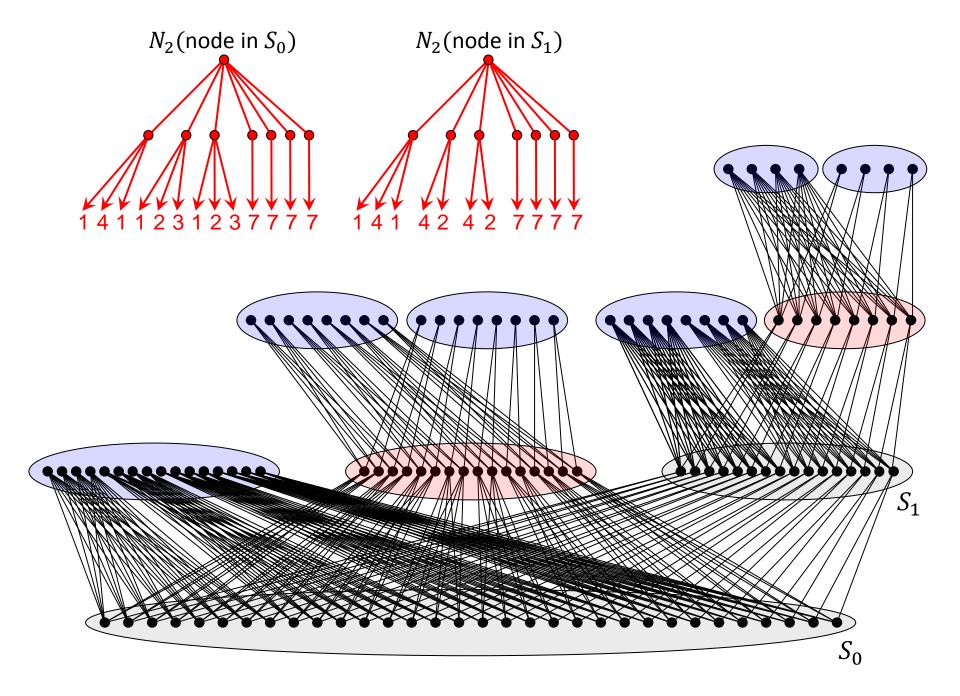
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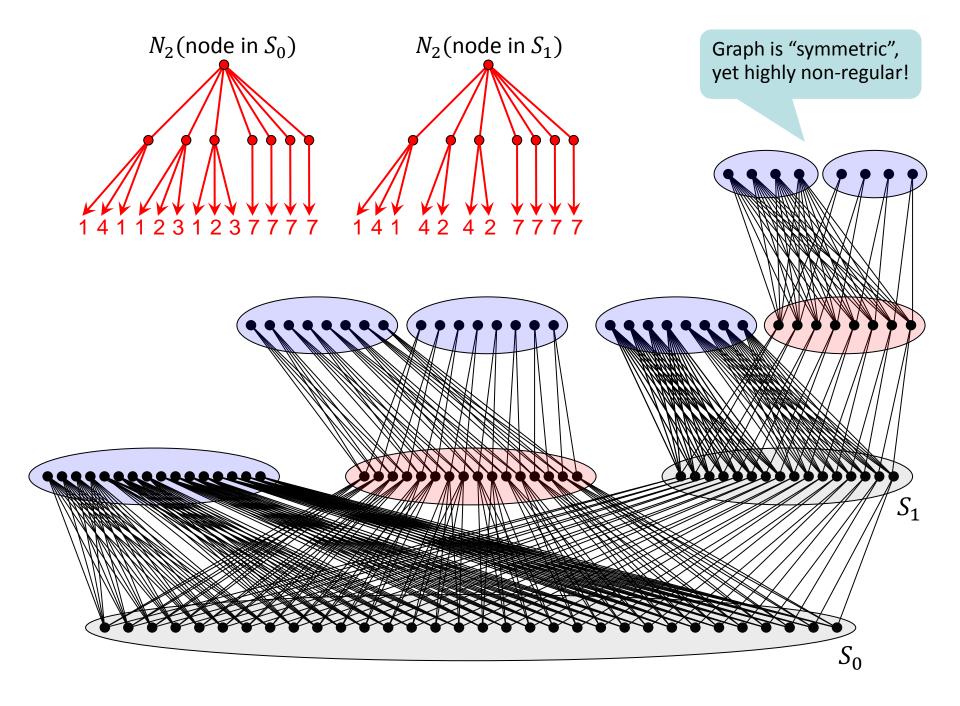
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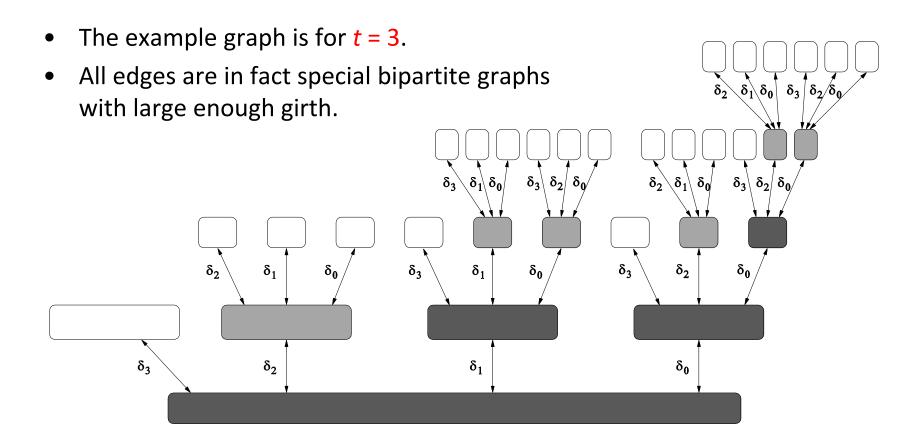
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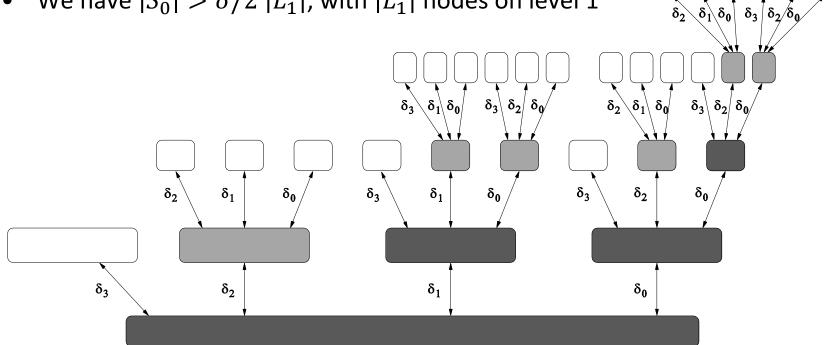
Lower Bound: The Argument



 If you use the graph of recursion level t, then a distributed algorithm cannot find a good MVC approximation in time t.

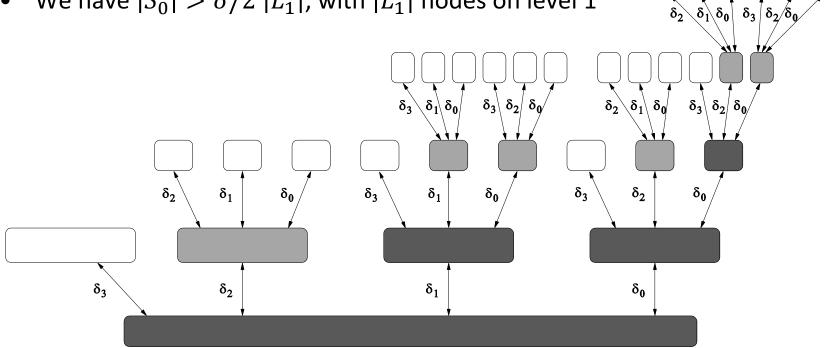
Lower Bound: The Math

- Choose degrees δ_i such that $\delta_{i+1}/\delta_i = 2^i \delta$.
- We have $|S_0| > \delta/2 |L_1|$, with $|L_1|$ nodes on level 1



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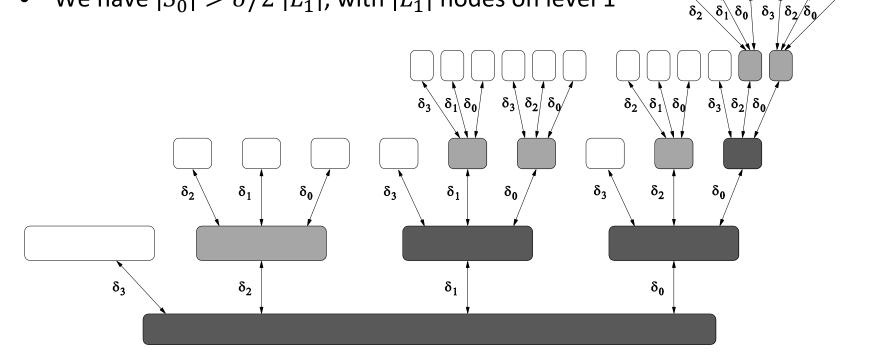


- By induction we have a $(1 \Theta(1/\delta))$ fraction of the nodes is in S_0 .
- Now δ , n, Δ are depending on the recursion level t.

Lower Bound: The Math

Graph useful for proving lower bounds in sublinear algos?

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Lower Bound: Results

• We can show that for $\epsilon > 0$, in t time, the approximation ratio is at least

$$\Omega\left(n^{rac{1/4-arepsilon}{t^2}}
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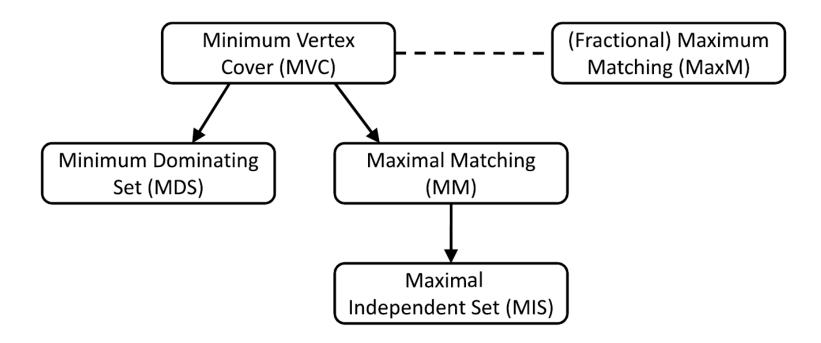
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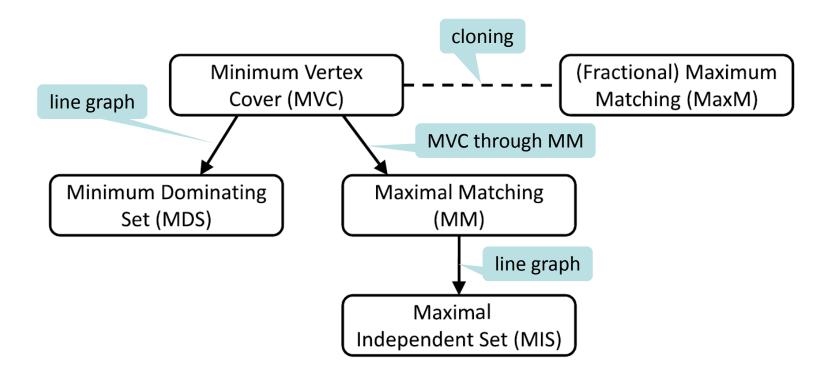
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• Many "local looking" problems need non-trivial t, in other words, the bounds $\Omega(\log \Delta)$ and $\Omega(\sqrt{\log n})$ hold for a variety of classic problems.



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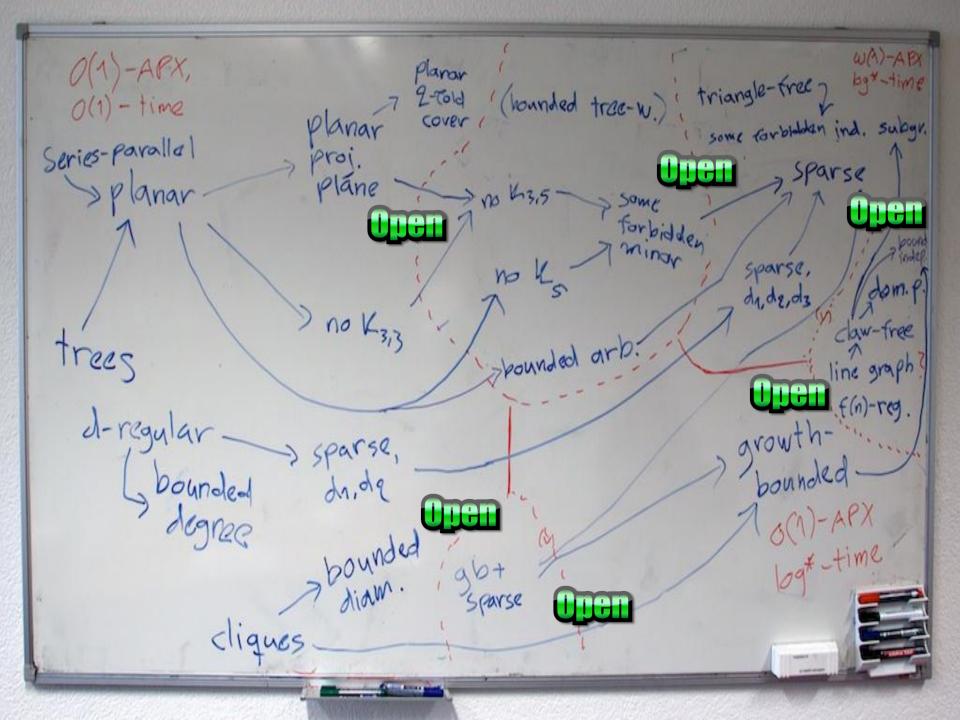
 $\sqrt{\log n \dots \log n}$

 n^{ϵ}

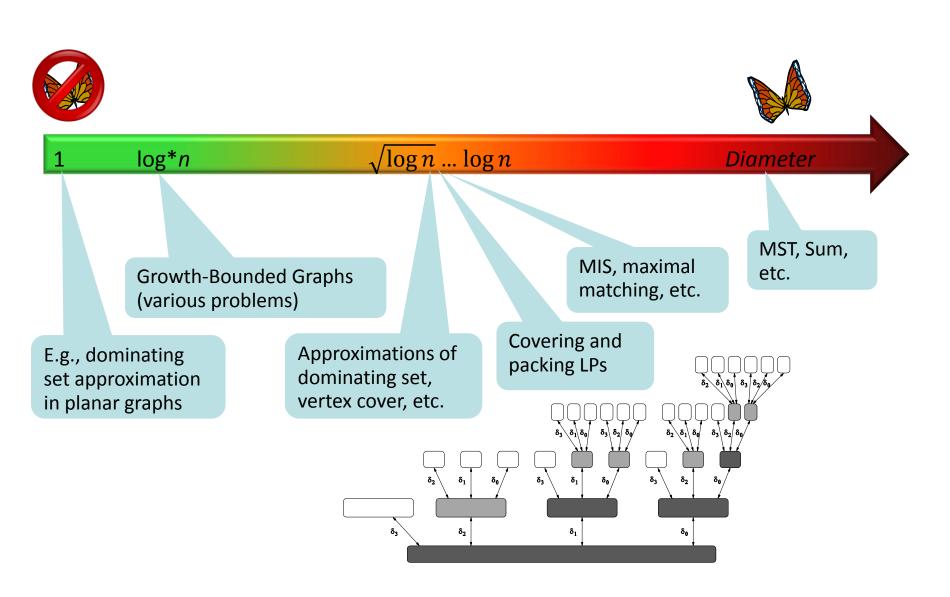
n

Linked List [Linial, 1992] **General Graphs** [Kuhn et al., 2004, 2006]

WA)-ABX 0(1)-APX, Planar triangle-free 7 2-Told 0(1) - time (bounded trae-w.) some forbladen ind. subgr. planar COVEY Series-Parallal proj. Sparse plane > planar Some forbidden sparse, d1, d2, d3 no K313 claw-free shounded arb. trees line graph f(n)-reg. d-regular growthsparse bounded y bounded degree drida 0(1)-APX bounded log* -time 96+ diam. Sparse cliques



Summary



Thank You!

Questions & Comments?



Thanks to my co-authors
Fabian Kuhn
Thomas Moscibroda
Johannes Schneider

Open Problems

- Close the gap between $\sqrt{\log n}$ and $\log n$ (for randomized algorithms)!
- Find a fast deterministic MIS algorithm (or strong det. lower bound)!
- Where are the boundaries between constant, log*, log, and diameter?
- What about algorithms that cannot even exchange messages?
- Can the lower bound graph be used in the context of sublinear algorithms?

