# Algorithms in Nature

Maximal Independent Set (MIS)

#### Overview

A minimum connected dominating set (MCDS) is a useful substructure of a graph representing a network:

- routing
- access control
- coverage

Computing a MCDS in a general graph is NP-complete ... But an approximation may be much easier to compute.

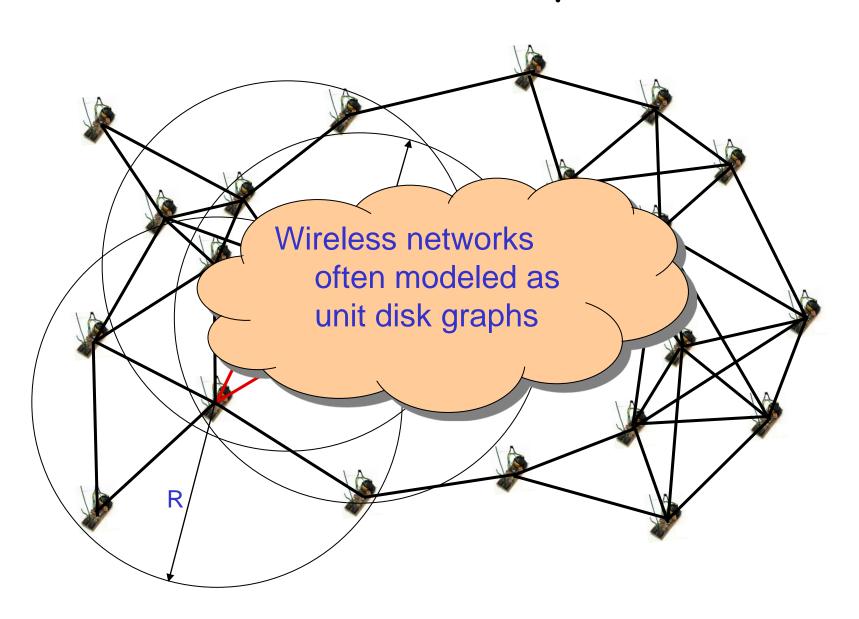
#### Overview

A

independent set is a constant approximation of a MCDS

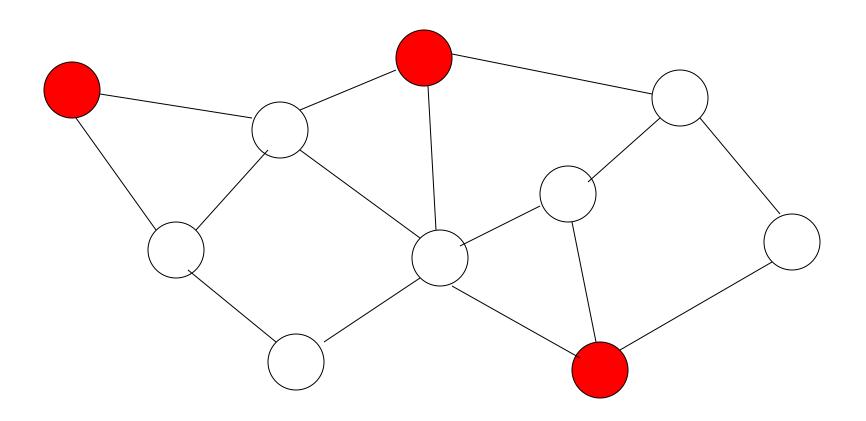
A MIS is an independent subset S of the nodes of a graph (none of the nodes in S are neighbors), and no superset of S is independent

# Unit Disk Graphs



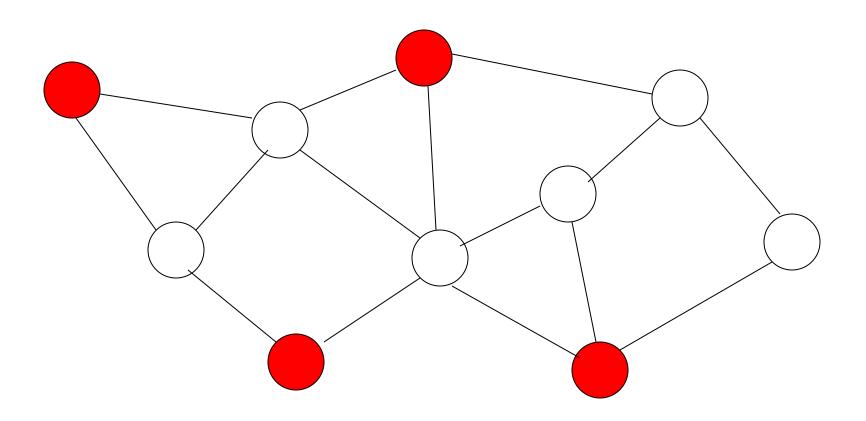
### Independent Set (IS):

Any set of nodes that are not adjacent



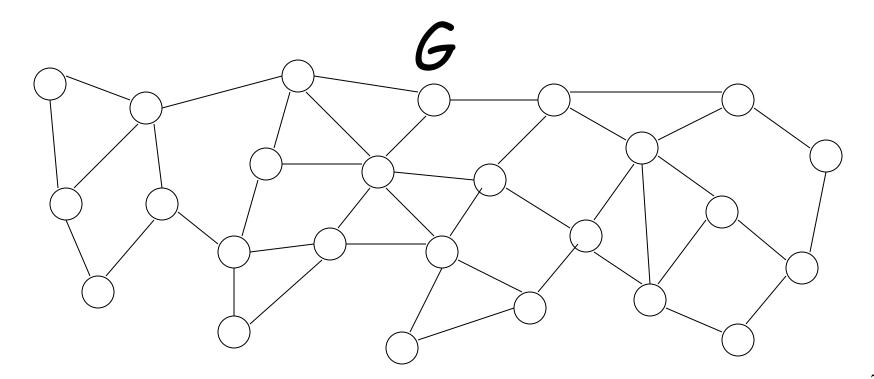
## Maximal Independent Set (MIS):

An independent set that is no subset of any other independent set



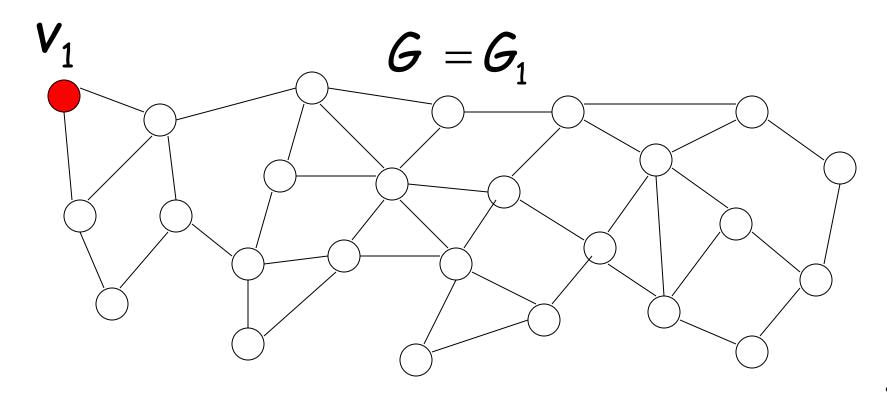
## A Sequential Greedy algorithm

Suppose that I will hold the final MIS Initially  $I=\varnothing$ 

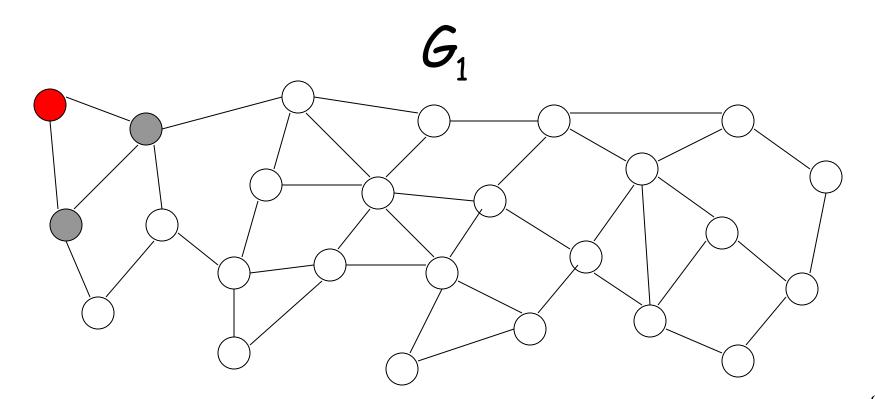


#### Phase 1:

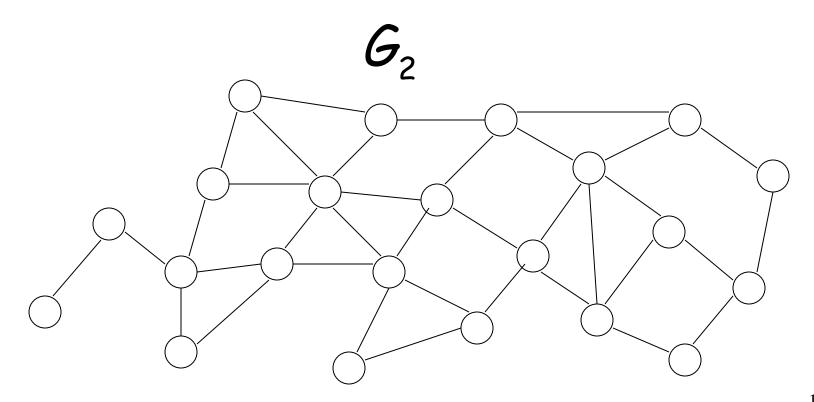
Pick a node  $V_1$  and add it to I



# Remove $V_1$ and neighbors $N(v_1)$

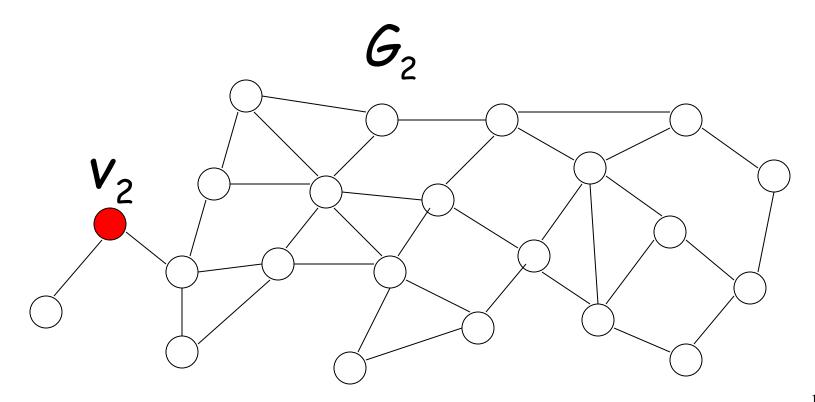


# Remove $V_1$ and neighbors $N(v_1)$

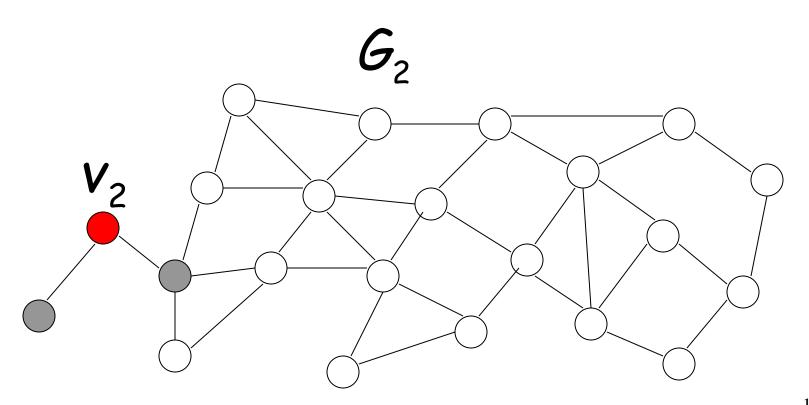


#### Phase 2:

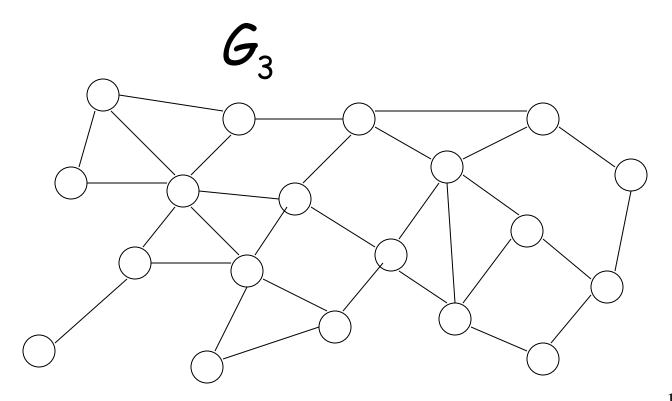
Pick a node  $V_2$  and add it to I



# Remove $V_2$ and neighbors $N(v_2)$

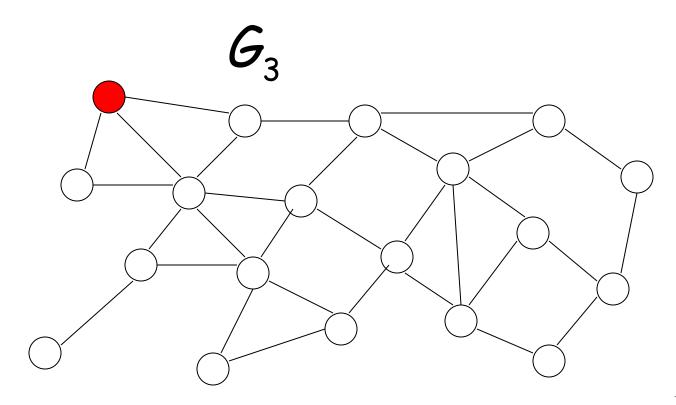


# Remove $V_2$ and neighbors $N(v_2)$



## Phases 3,4,5,...:

## Repeat until all nodes are removed



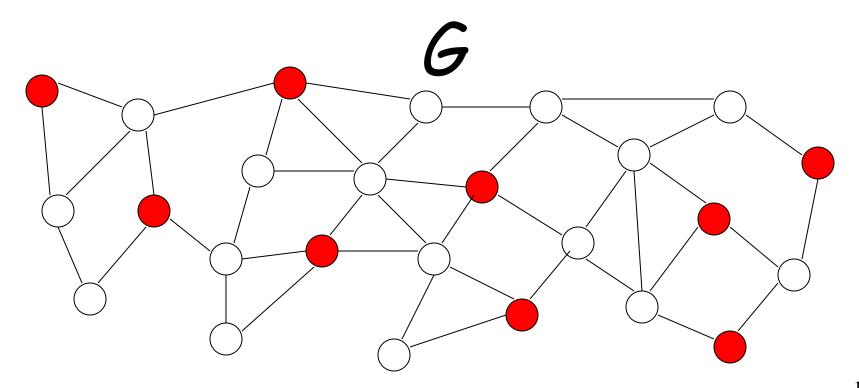
Phases 3,4,5,...,x:

Repeat until all nodes are removed

$$G_{x+1}$$

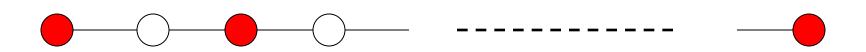
No remaining nodes

# At the end, set I will be an MIS of G



## Running time of algorithm: O(n)

Worst case graph:



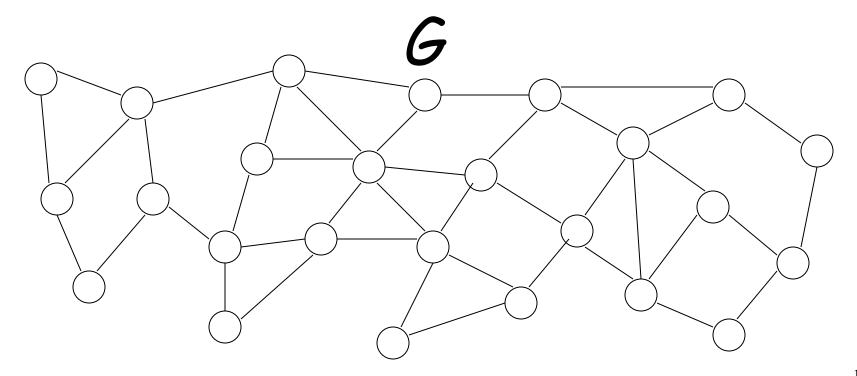
n nodes

## A General Algorithm For Computing MIS

Same as the sequential greedy algorithm, but at each phase we may select any independent set (instead of a single node)

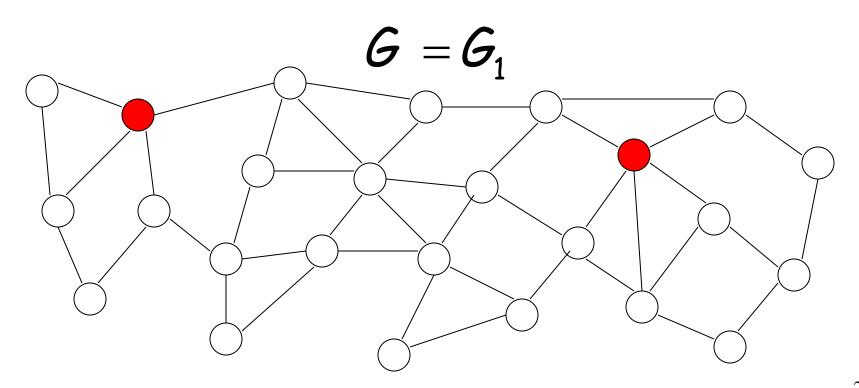
## Example:

Suppose that I will hold the final MIS Initially  $I=\varnothing$ 

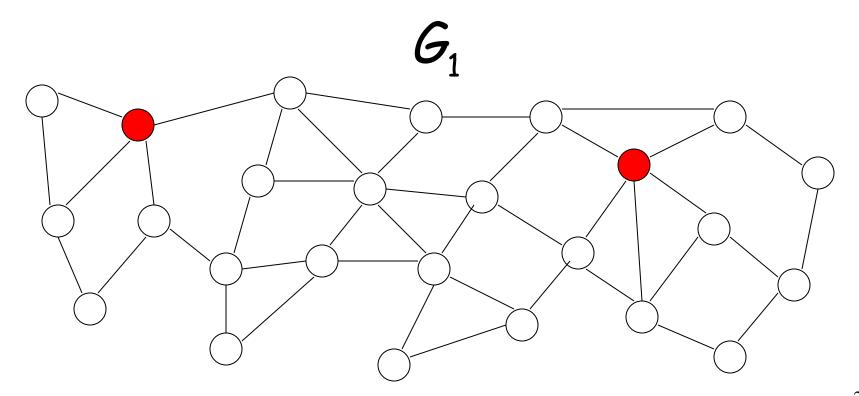


#### Phase 1:

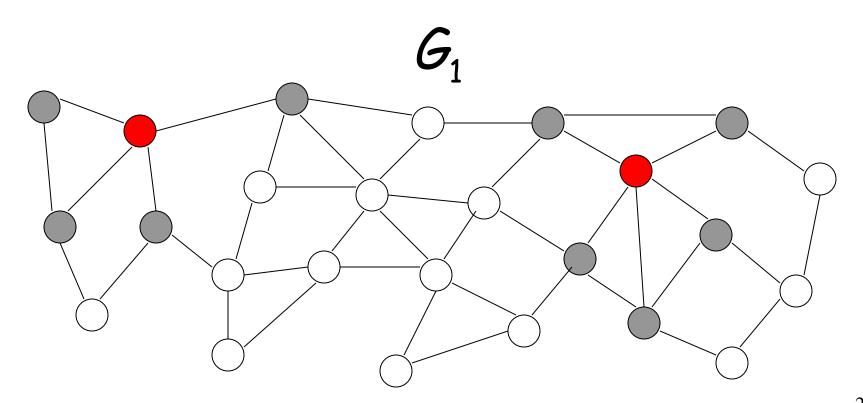
Find any independent set  $I_1$ And insert  $I_1$  to  $I: I \leftarrow I \cup I_1$ 



# remove $I_1$ and neighbors $N(I_1)$



# remove $I_1$ and neighbors $N(I_1)$



# remove $I_1$ and neighbors $N(I_1)$

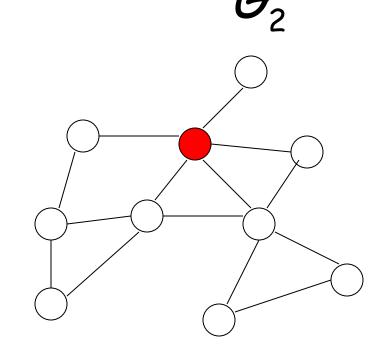
$$G_2$$

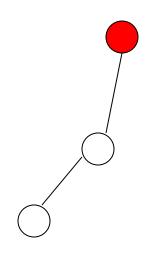
#### Phase 2:

## On new graph

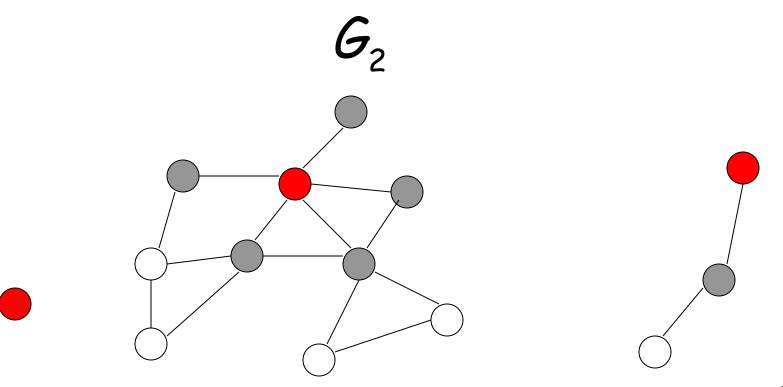
Find any independent set  $I_2$ 

And insert  $I_2$  to  $I: I \leftarrow I \cup I_2$ 



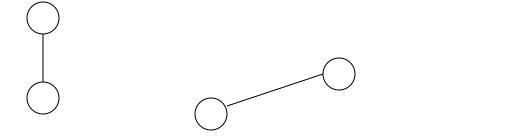


# remove $I_2$ and neighbors $N(I_2)$



# remove $I_2$ and neighbors $N(I_2)$

 $G_3$ 



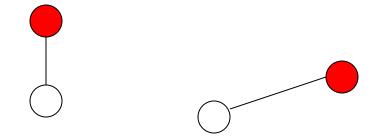
#### Phase 3:

## On new graph

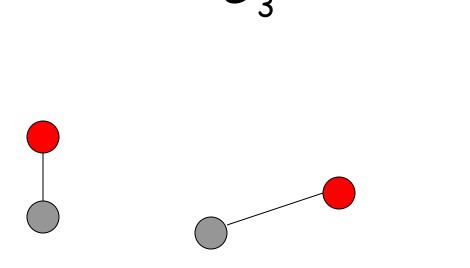
Find any independent set  $I_3$ 

And insert  $I_3$  to  $I: I \leftarrow I \cup I_3$ 





# remove $I_3$ and neighbors $N(I_3)$

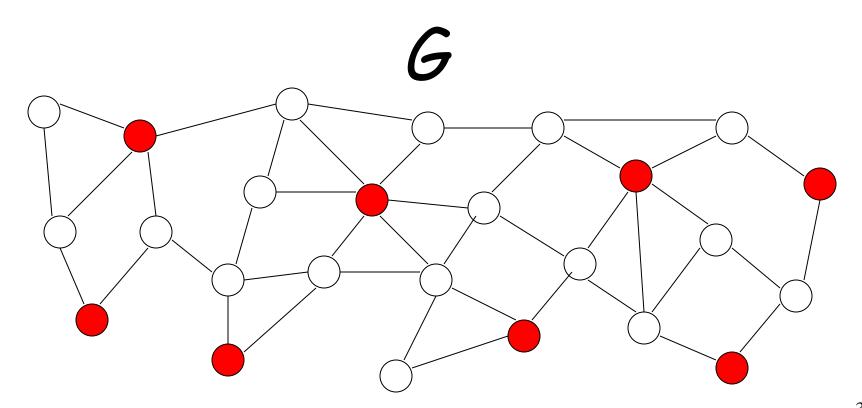


remove  $I_3$  and neighbors  $N(I_3)$ 

 $G_4$ 

No nodes are left

# Final MIS I



#### Observation:

The number of phases depends on the choice of independent set in each phase:

The larger the independent set at each phase the faster the algorithm

Example: If  $I_1$  is MIS, 1 phase is needed

Example: If each  $I_k$  contains one node, O(n) phases are needed

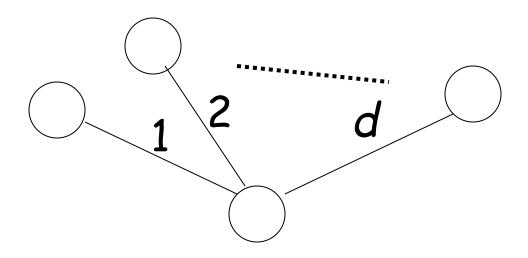
(sequential greedy algorithm)

## A Simple Distributed Algorithm

Same as the general MIS algorithm

At each phase the independent set is chosen randomly so that it includes many nodes of the remaining graph

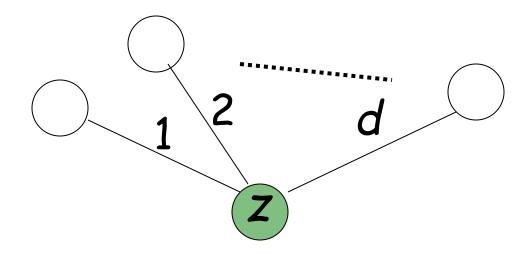
# Let d be the maximum node degree in the whole graph



Suppose that d is known to all the nodes

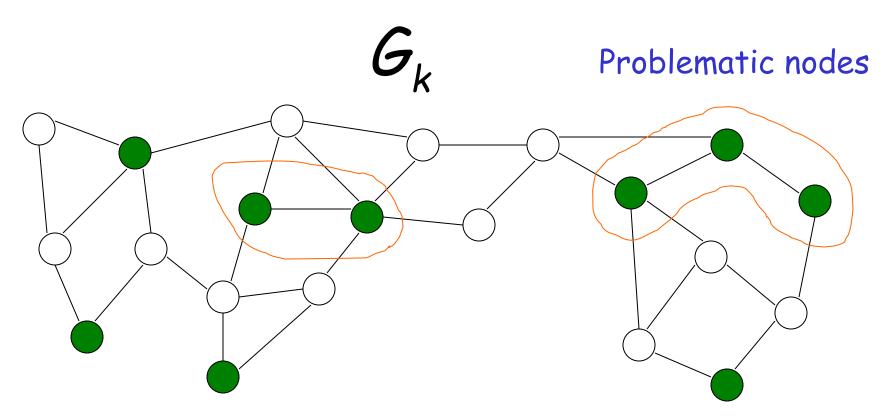
## At each phase k:

Each node  $z \in G_k$  elects itself with probability  $p = \frac{1}{d}$ 

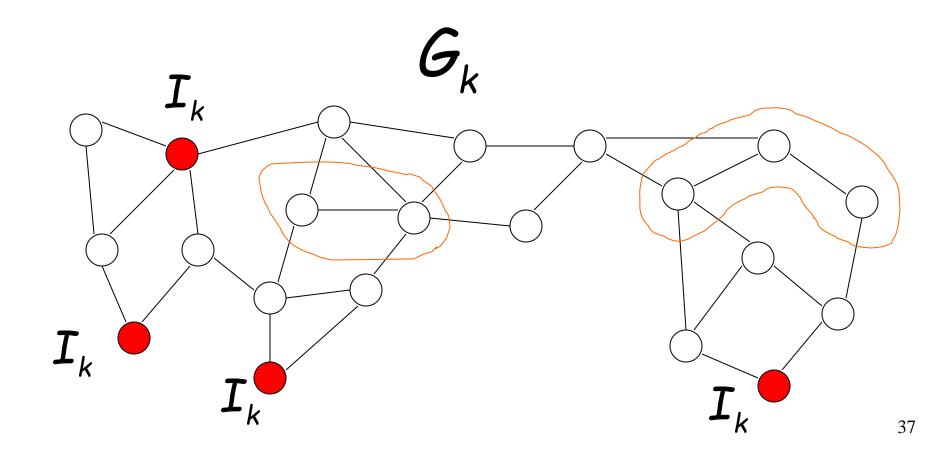


Elected nodes are candidates for independent set  $\, T_k \,$ 

# However, it is possible that neighbor nodes may be elected simultaneously



All the problematic nodes must be un-elected. The remaining elected nodes form independent set  $\mathcal{I}_k$ 

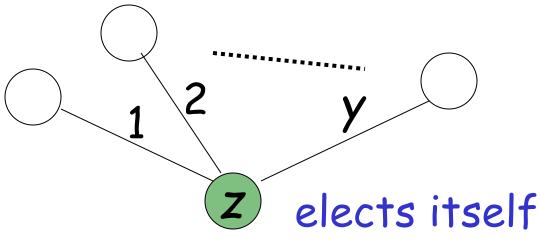


### Analysis:

Success for a node  $Z \in G_k$  in phase k: Z disappears at end of phase k (enters  $I_k$  or  $N(I_k)$ )

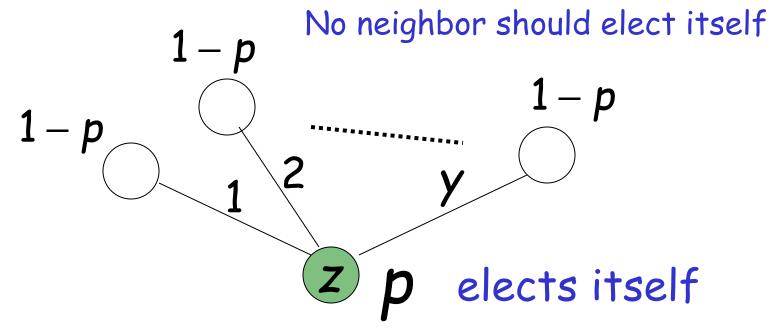
A good scenario that guarantees success

No neighbor elects itself



#### Probability of success in phase:

At least 
$$p(1-p)^y \ge p(1-p)^d$$



#### Fundamental inequalities

$$\frac{1}{e} \cong \left(1 - \frac{1}{d}\right)^d$$

#### Probability of success in phase:

#### At least

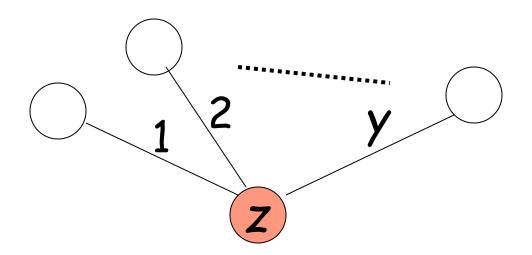
$$p(1-p)^{y} \ge p(1-p)^{d}$$

$$\ge \frac{1}{d} \left(1 - \frac{1}{d}\right)^{d}$$

$$\le \frac{1}{ed}$$

For  $d \geq 2$ 

Therefore, node Z will enter  $I_k$  and disappear in phase k with probability at least  $\frac{1}{ed}$ 



### Expected number of phases until node Z disappears:

at most 
$$\frac{1}{\text{probability of sucess in phase}} = ed$$
 phases

#### Bad event for node Z:

after  $2ed \ln n$  phases node Z did not disappear

#### Probability:

$$\left(1 - \frac{1}{ed}\right)^{2ed\ln n} \le \frac{1}{e^{2\ln n}} = \frac{1}{n^2}$$

Bad event for any node in G:

after  $2ed \ln n$  phases

at least one node did not disappear

#### Probability:

$$\sum_{x \in G} (\text{probability of bad event for } x) \le n \frac{1}{n^2} = \frac{1}{n}$$

Good event for all nodes in G:
within  $2ed \ln n$  phases
all nodes disappear

### Probability:

$$1-[probability of bad event] \ge 1-\frac{1}{n}$$

(high probability)

#### Total number of phases:

```
2ed \ln n = O(d \log n)
with high probability
```

Time duration of each phase: O(1)

Total time: O(d logn)

Still could be very large if the max degree is O(n)

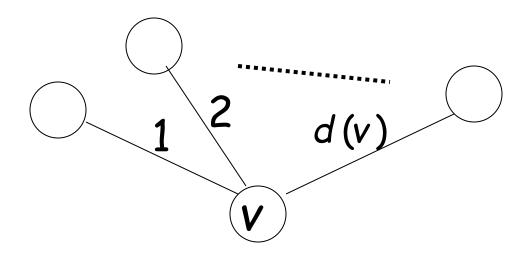
#### Luby's MIS Distributed Algorithm

Runs in time O(logn) in expected case

O(logd · logn) with high probability

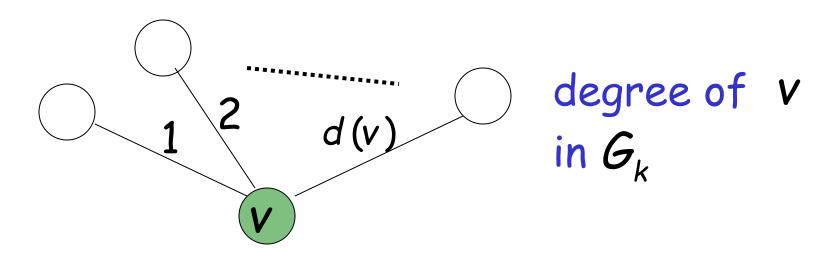
this algorithm is asymptotically better than the previous

### Let d(v) be the degree of node v



#### At each phase k:

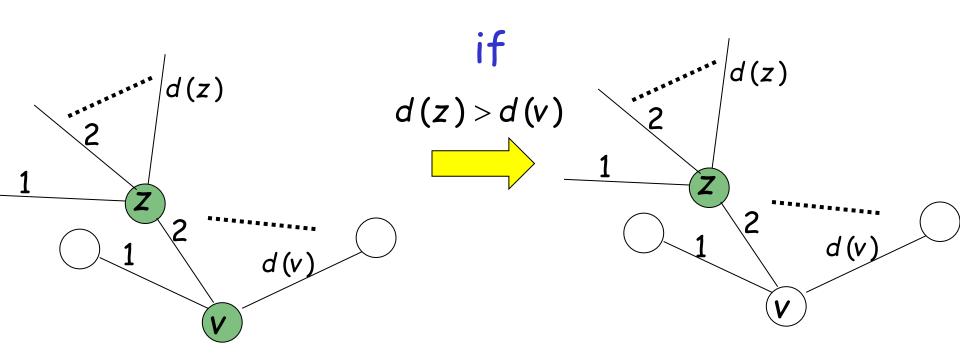
Each node  $v \in G_k$  elects itself with probability  $p(v) = \frac{1}{2d(v)}$ 



Elected nodes are candidates for the independent set  $I_k$ 

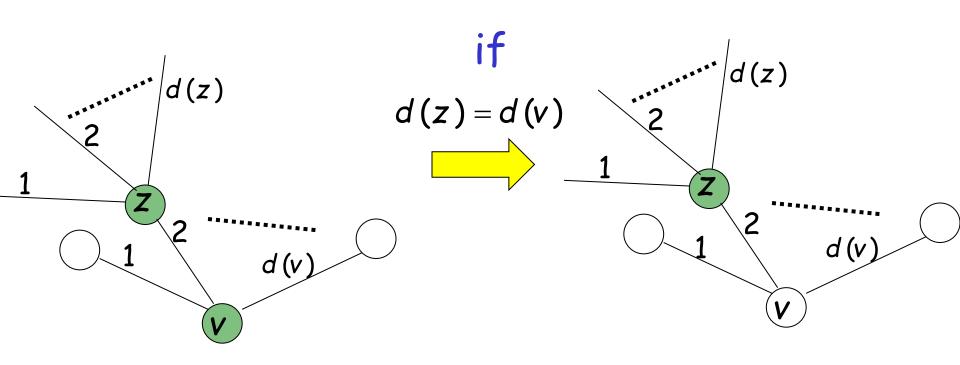
# If two neighbors are elected simultaneously, then the higher degree node wins

#### Example:

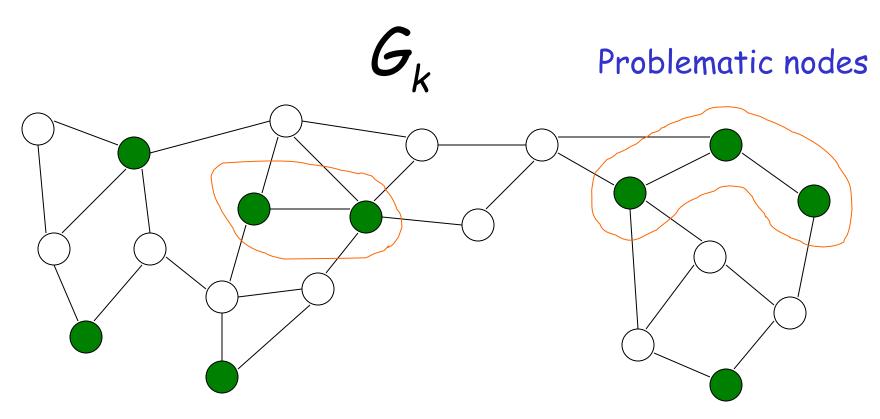


# If both have the same degree, ties are broken arbitrarily

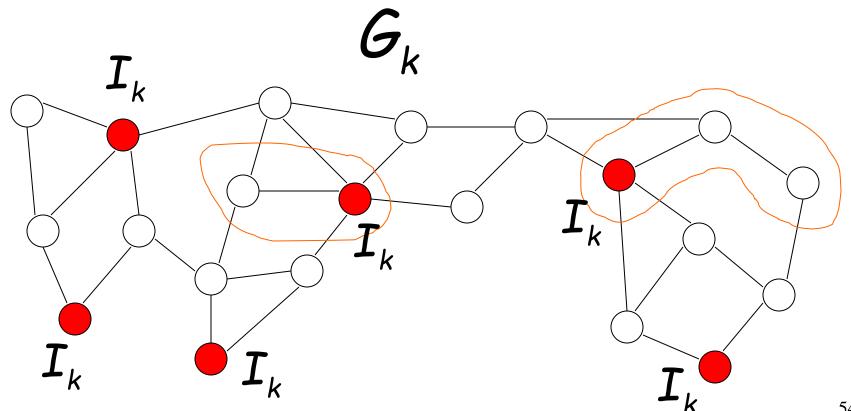
### Example:



### Using previous rules, problematic nodes are removed



# The remaining elected nodes form independent set $I_k$



### Luby's MIS Distributed Algorithm (1982)

this algorithm is asymptotically better than the previous

Runs in time O(logn) in expected case

 $O(\log d \cdot \log n)$  with high probability