

# Lecture 04 : More on Locality

# Announcements

1. Assignment 01 due next Friday
  - draft posted
  - work in small groups — 1 submission per group
  - group assignment survey (see email)
2. Attendance + COVID
  - college masking policies strictly enforced
    - no food/drink in class
  - all sessions recorded, notes posted
  - do not come to class if sick/exposed
  - risk = shared risk

# Last Time

## Port #ing

Introduced Port Ordering (PO) Model

- network = graph w/ port ordering
  - no unique identifiers for nodes
- execution in synchronous rounds
  - receive msgs
  - perform local computations (no restrictions!)
  - send msgs to neighbors

LOCAL = PO + IDs

- Stable Matchings in PO model

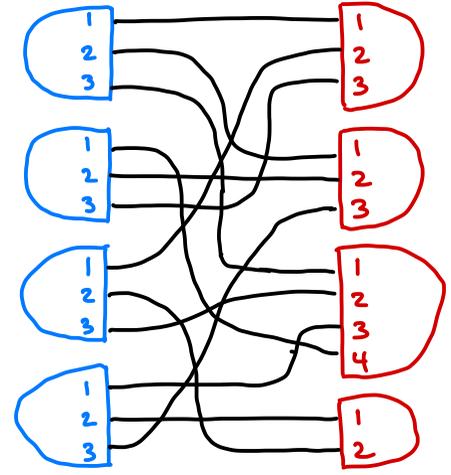
- SM instances → PO networks
- additional input: role (student or internship)
- GS algorithm has natural interpretation as protocol in PO model
- ISSUE: termination detection

## SM Instance

	1	2	3	4		1	2	3	4
Anna:	a	b	c		a:	A	C	B	
Beck:	c	b	a		b:	A	B	D	
Cameron:	a	d	c		c:	A	C	D	B
Daniel:	c	d	b		d:	D	C		



## PO Network



## Gale Shapley as a PO protocol

### Procedure for students:

Initialize:

$cur = 1$  —  $cur$  match

each round  $i$ , do

if  $i = 1$ , send "apply" to  $cur$

→ if received "reject" from  $cur$  in round  $i-1$

if  $cur = \text{my degree}$ , return  $\perp$  and halt

→ else,  $cur \leftarrow cur + 1$ , send "apply" to  $cur$

return  $cur$  and halt

### Procedure for internships

Initialize:

$cur = \infty$

each round  $i$ , do

set reject  $\leftarrow \emptyset$

for each  $j$  from which received "apply" in round  $i-1$

if  $j < cur$  ←

add  $cur$  to reject

set  $cur \leftarrow j$

else

add  $j$  to reject ←

for each  $j$  in reject

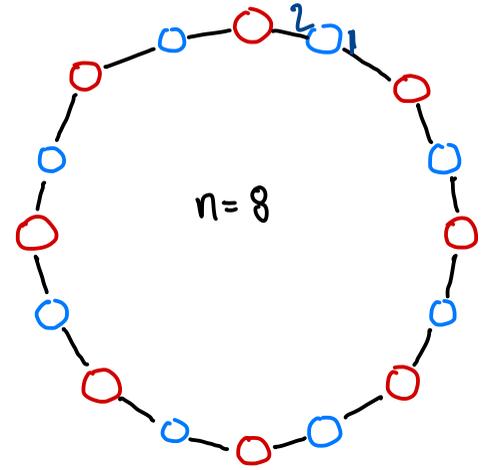
send "reject" to  $j$  }

return  $cur$

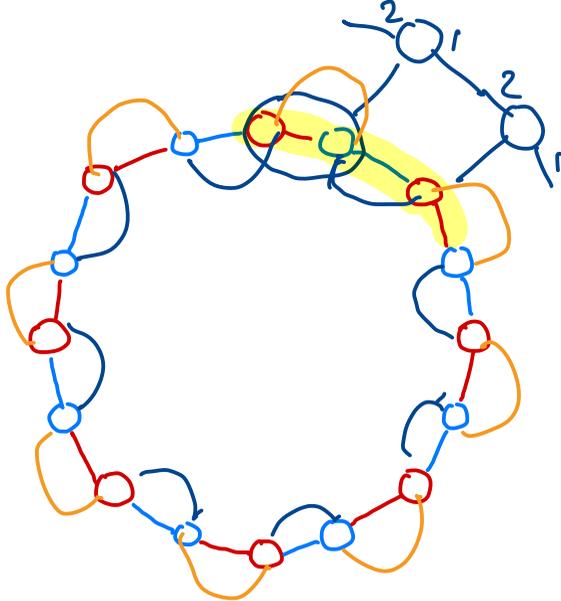
applicant  
to reject

## An Illustrative Family of SM Instances

1.  $n$  students,  $n$  internships
2. each agent has only 2 acceptable partners
3. the graph of acceptable partners forms a cycle



## Some preferences



- students prefer blue neighbor to red neighbor
- internships prefer red neighbor to blue neighbor

find matching of blue edges  
↖

## Questions.

1. Which SM does GS algorithm find?
  2. What other SMs are there?
- Internships  
← apply to students  
red matching

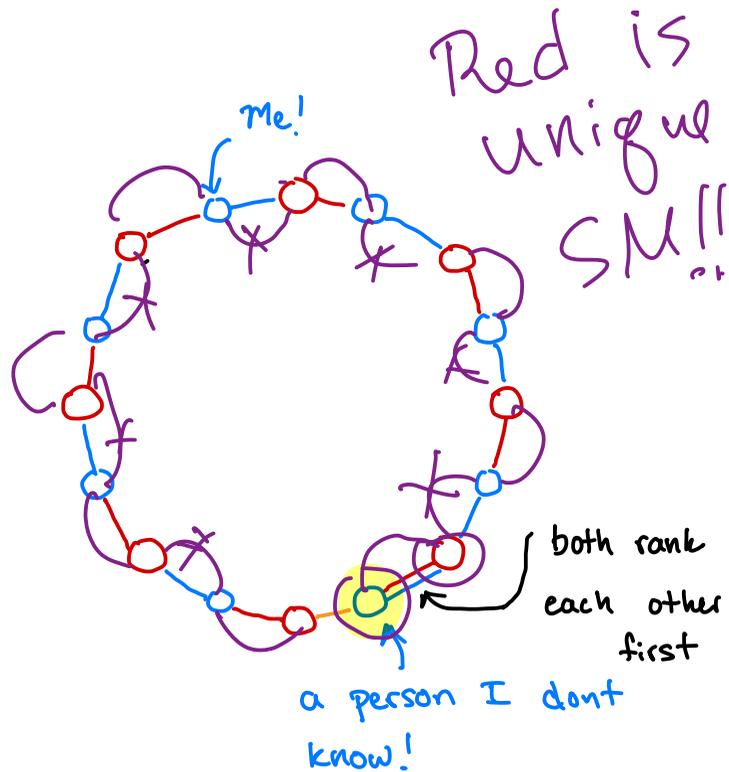
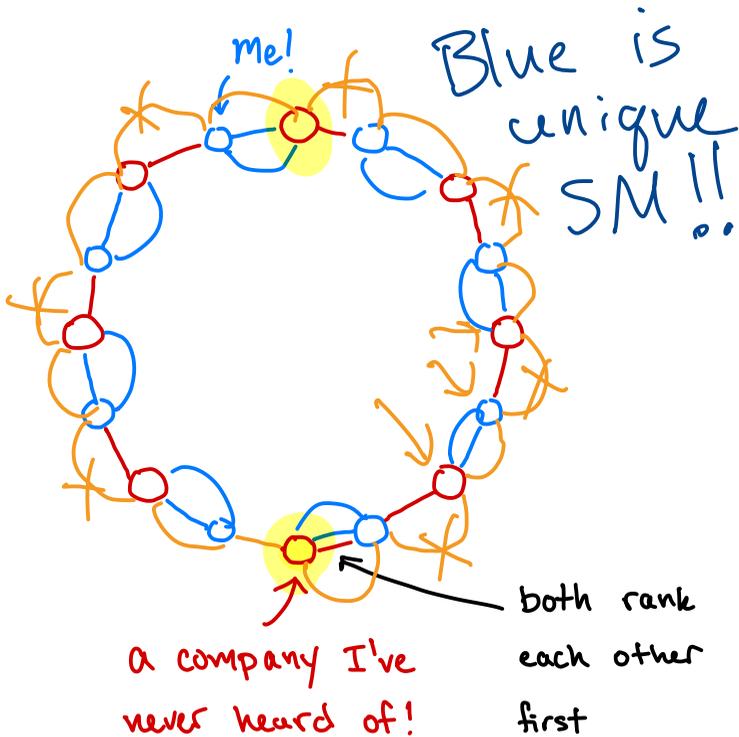
## Recall

- can sometimes have multiple possible stable matchings
- here's how to find 2 (if possible):
  - run GS alg w/ students applying to internships
  - run GS alg. w/ internships applying to students

Fact. If the SM instance has multiple stable matchings then procedures above will give different stable matchings

⇒ if procedures above give same SM, then it is the unique SM for the instance.

# Slight Modifications



What happens now?

## The Moral.

My (correct) output depends on input (prefs) of some company I've never heard of and some person I don't know!

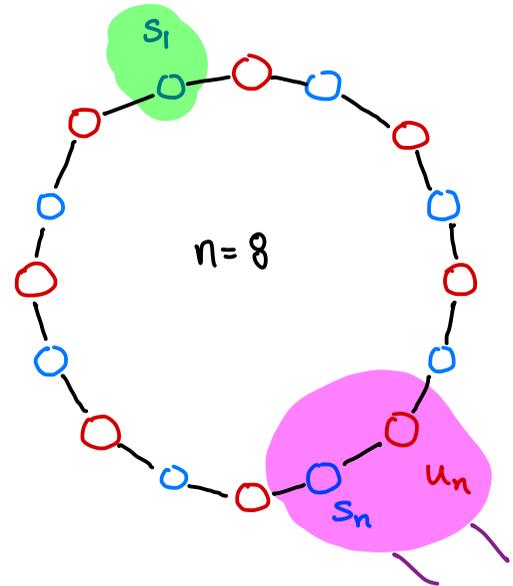
- correct<sub>local</sub> output requires knowledge about nodes that are far away in the network

Obs. Correct output of  $S_i$  depends on input of  $S_n$  and  $u_n$ , regardless of what algorithm is used to find SM

Next. We will use this observation to show that no distributed algorithm (in PO or LOCAL model) can find SMs "much faster" than GS on some instances

"A distributed system is one in which the failure of a computer you didn't even know existed can render your own computer unusable."

- Leslie Lamport



## A Bit More Graph Theory

- $G = (V, E)$  a graph
- edges labelled w/ port ordering
- vertices have local inputs  
(possibly unique identifiers)

Def. Given vertex  $v$ , distance  $d$  the  $d$ -neighborhood of  $v$  is the subgraph of  $G$  consisting of all vertices within distance  $d$  of  $v$  and edges w/ one endpoint within distance  $d-1$  of  $v$

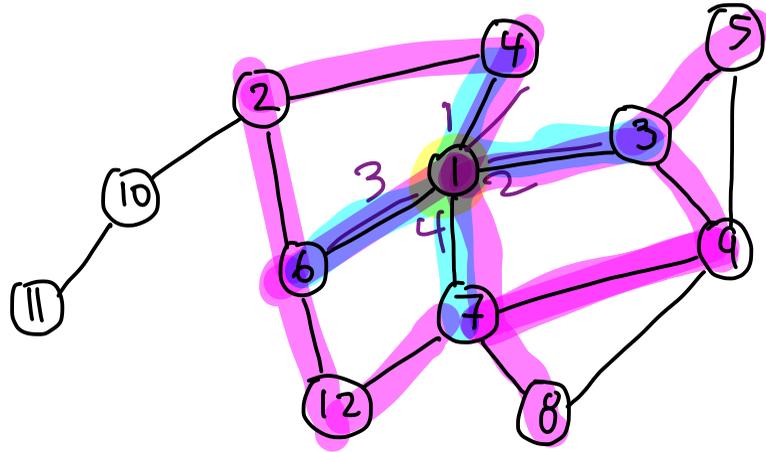
$$\Gamma_d(v) = (V', E')$$

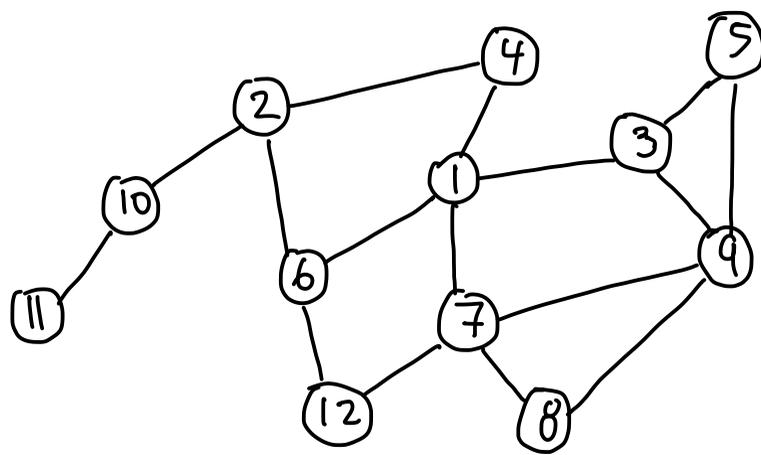
$V'$  = vertices within distance  $d$  of  $v$

$E'$  = edges w/ one or more endpoints within dist  $d-1$  of  $v$

$$E = \{(1,3), (1,4), (1,6), (1,7), \dots\}$$

$$\text{vertices} = \{1, 2, \dots, 12\}$$

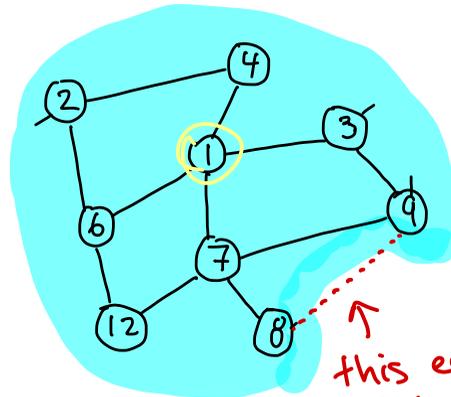
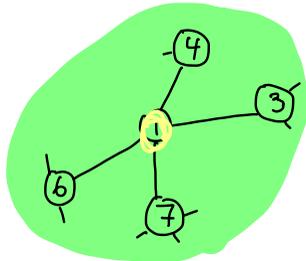




$\Gamma_0(1)$



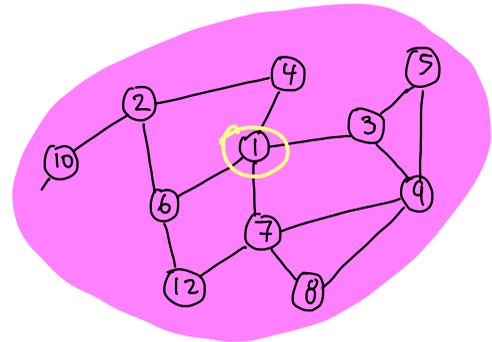
$\Gamma_1(1)$



$\Gamma_2(1)$

this edge  
not included!

$\Gamma_3(1)$



## Connecting Neighborhoods and Protocols.

Goal. In any distributed protocol, messages sent in round  $r$  are determined by distance  $r-1$  neighborhoods.

- Intuitively: data can only travel one "hop" per round
  - $\Rightarrow$  takes  $r-1$  rounds to travel  $r-1$  hops
  - $\Rightarrow$  by round  $r$ , can only have gotten info. from nodes within dist  $r-1$
- Formally: prove by induction that each  $v$ 's state in round  $r$  is determined by the initial states of nodes in  $\Pi_{r-1}(v)$ 
  - msgs sent in round  $r$  are determined by state in round  $r$ .

## Union of Subgraphs

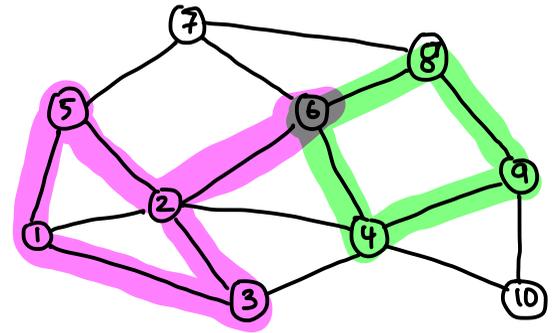
Suppose  $H_1, H_2$  are sub-graphs of a graph  $G = (V, E)$ . Then the union of  $H_1$  and  $H_2$ , denoted  $H_1 \cup H_2$ , consists of

- (1) all vertices in  $H_1$  or  $H_2$
- (2) all edges in  $H_1$  or  $H_2$

We can also form unions of many subgraphs:

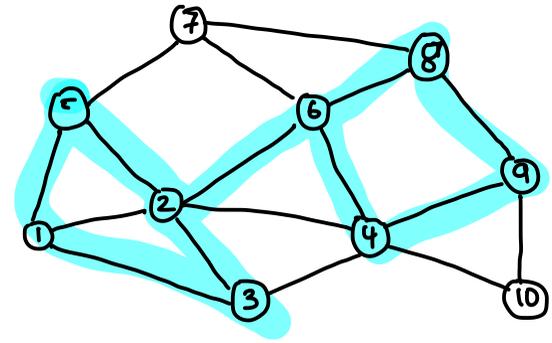
$$H_1 \cup H_2 \cup H_3 \cup \dots \cup H_k = \bigcup_{i=1}^k H_i$$

= vertices and edges in any of the subgraphs



$H_1$

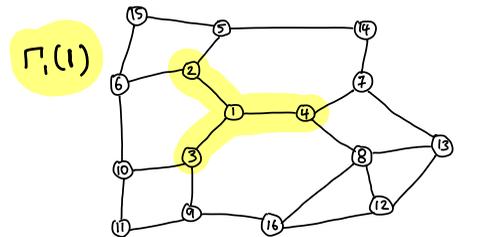
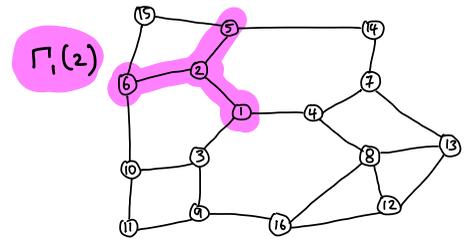
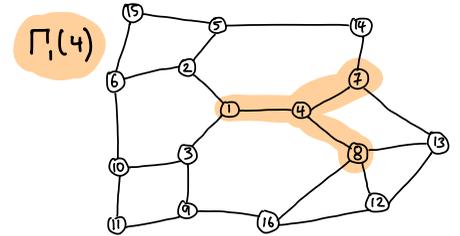
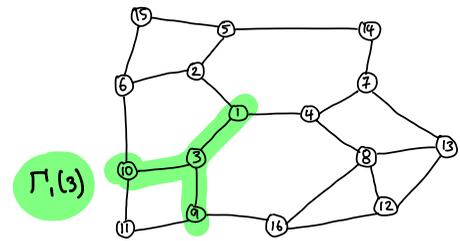
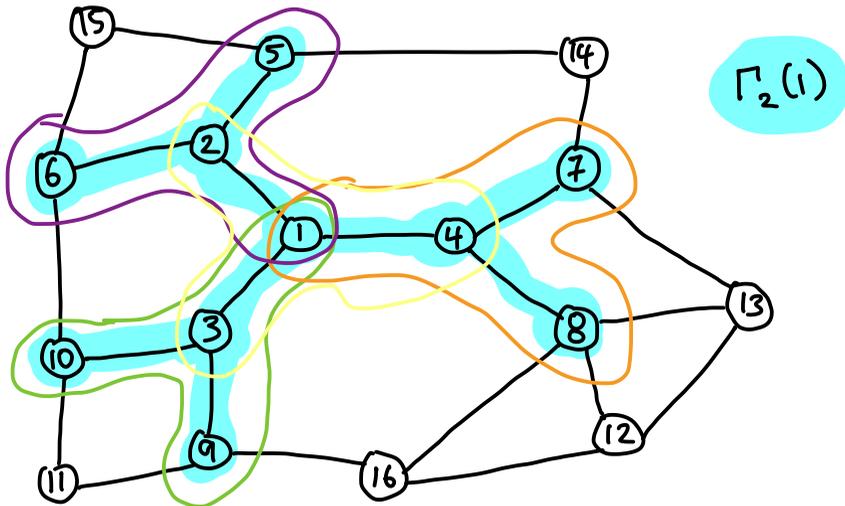
$H_2$



$H_1 \cup H_2$

Neighborhood Covering Lemma. Let  $G=(V,E)$  be a graph and  $v \in V$  a vertex in  $G$ . For any distance  $d \geq 2$ , the distance  $d$  neighborhood of  $v$  is the union of  $v$ 's neighbors distance  $d-1$  neighborhoods. Symbolically:

$$\Gamma_d(v) = \bigcup_{w \in \Gamma_1(v)} \Gamma_{d-1}(w)$$

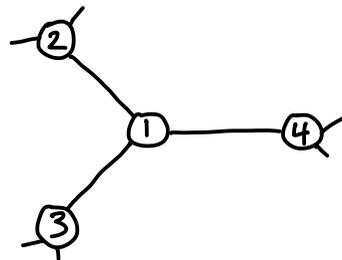
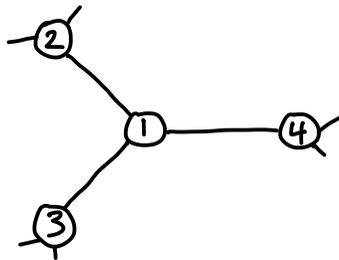


For Now. Take neighborhood covering lemma on faith

→ see notes for a proof

Want to show: state of each vertex  $v$  is determined by the initial state of vertices in  $\Gamma_{r-1}(v)$ . Formally:

Locality Lemma. Suppose  $\Pi$  is a PD protocol,  $G = (V, E)$  is a graph together w/ local inputs for each vertex  $v$ . Then for every round  $r$ , and vertex  $v$ , the state of  $v$  in round  $r$  is determined by the initial state of  $\Gamma_{r-1}(v)$ .



Proof idea. The state of  $v$  in round 1 is determined only by  $v$ 's local input

⇒ msgs sent in round 1 are functions of local inputs

State in round 2 is function of  $v$ 's local input and msgs received in round 1

⇒ msgs sent by  $v$  in round 2 are functions of inputs of distance 1 neighborhoods

...

Proof. Want to show that for all vertices  $v$  and rounds  $r$ ,  $v$ 's state is determined by  $\Gamma_{r-1}(v)$ .

Argue by induction on  $r$ .

Base case:  $r=1$ .

- initial state of  $v$  is determined by  $v$ 's local input: hence  $\Gamma_0(v)$ .

Inductive Step:  $r-1 \Rightarrow r$

Suppose in round  $r-1$ , every  $u$ 's state is determined by  $\Gamma_{r-2}(u)$

Then msgs sent by each  $u$  in rnd  $r-1$  are determined by  $\Gamma_{r-2}(u)$

PO protocol consists of

Sets:

- Inputs = allowable local inputs
- States = allowable local states
- Outputs  $\subseteq$  States halting states w/ outputs
- Messages = allowable messages

Functions ( $d = \text{degree of node}$ )

- init: Inputs  $\rightarrow$  States  
determine initial state from initial input
- send: States  $\rightarrow$  Messages <sup>$d$</sup>   
determine the  $d$  messages to send to neighbors from current state
- receive: States  $\times$  Messages <sup>$d$</sup>   $\rightarrow$  States  
determine how to update state from received messages

Inductive step continued...

Then msgs sent by each  $u$  in  
rnd  $r-1$  are determined by  $\Gamma_{r-2}(u)$

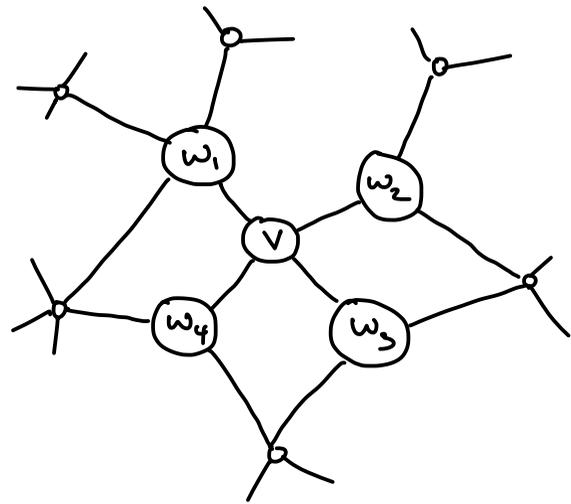
Consider vertex  $v$  w/ neighbors  
 $w_1, w_2, \dots, w_k$  ( $k = \deg(v)$ )

Msg received from  $w_i$  in round  $r-1$   
is determined by  $\Gamma_{r-2}(w_i)$

$\Rightarrow$  all msgs received in round  $r-1$   
are determined from

$$\Gamma_{r-2}(v) \cup \Gamma_{r-2}(w_1) \cup \Gamma_{r-2}(w_2) \cup \dots \cup \Gamma_{r-2}(w_k) = \Gamma_{r-1}(v)$$

$\Rightarrow v$ 's state in round  $r-1$  is determined  
by  $\Gamma_{r-1}(v)$ . //



## A More Algorithmic View of Locality Lemma

In round 1, consider  $\Gamma_{r-1}(v)$

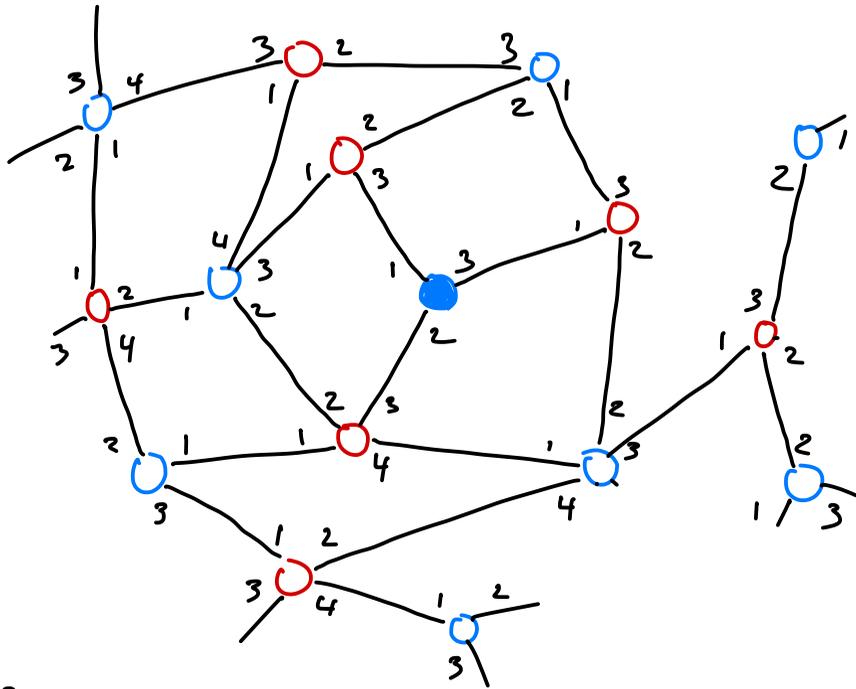
- compute all init. states
- compute all msgs
- determine all received msgs in  $\Gamma_{r-2}(v)$

In round 2, consider  $\Gamma_{r-2}(v)$

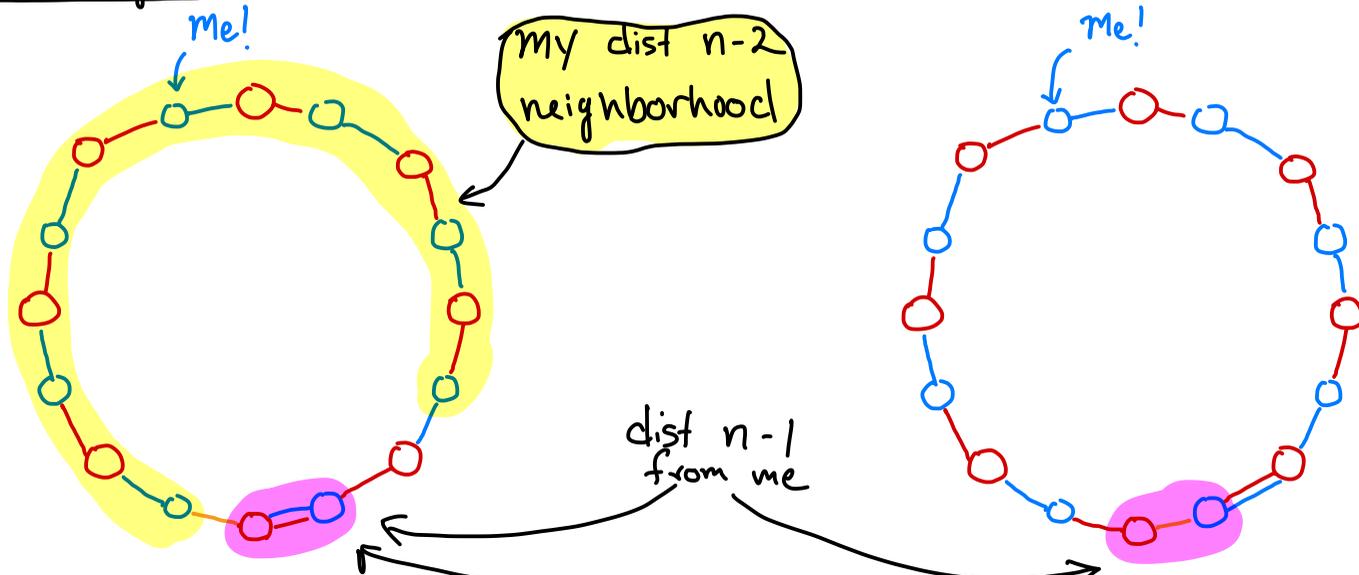
- compute all rnd 2 states
- compute all rnd 2 msgs
- find all received msgs in  $\Gamma_{r-3}(v)$

⋮

In round  $r-1$ , get all states/received msgs in  $\Gamma_0(v) = \{v\}$ .



Consider Again.



•  $n$  students,  $n$  internships

• my correct output depends on input of these

⇒ correct output is not determined by  $\Gamma_{n-2}(me)$

⇒ correct output cannot be determined in  $\leq n-1$  rounds by any algorithm.

