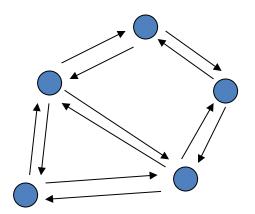
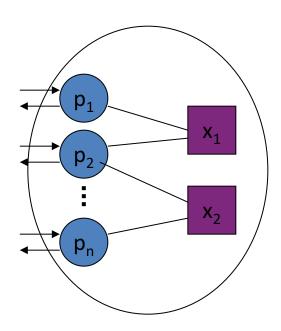
Distributed Algorithms 6.046J, Spring, 2015

Nancy Lynch

What are Distributed Algorithms?

- Algorithms that run on networked processors, or on multiprocessors that share memory.
- They solve many kinds of problems:
 - Communication
 - Data management
 - Resource management
 - Synchronization
 - Reaching consensus
 - **—** ...
- They work in difficult settings:
 - Concurrent activity at many processing locations
 - Uncertainty of timing, order of events, inputs
 - Failure and recovery of processors, of channels.
- So they can be complicated:
 - Hard to design, prove correct, and analyze.



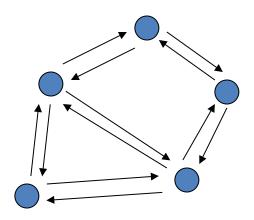


Distributed Algorithms

- Studied since 1967, starting with Dijkstra and Lamport.
 - Edsger Dijkstra: 1972 Turing award winner
 - Leslie Lamport: 2013 Turing award winner
- Some sources:
 - Lynch, Distributed Algorithms
 - Attiya and Welch, Distributed Computing:
 Fundamentals, Simulations, and Advanced
 Topics
 - Morgan Claypool series of monographs on Distributed Computing Theory
 - Conferences:
 - Principles of Distributed Computing (PODC)
 - Distributed Computing (DISC)

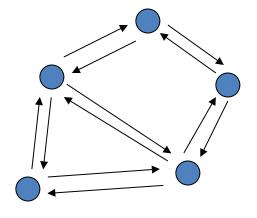
This Week

- A quick introduction:
 - Two common distributed computing models,
 - A few fundamental algorithms, and
 - Techniques for analyzing algorithms.
- Synchronous distributed algorithms:
 - Leader Election
 - Maximal Independent Set
 - Breadth-First Spanning Trees
 - Shortest Paths Trees
- Asynchronous distributed algorithms:
 - Breadth-First Spanning Trees
 - Shortest Paths Trees



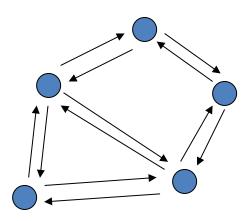
Modeling, Proofs, Analysis

- Important for distributed algorithms, because they are complicated and error-prone.
- Infinite-state interacting state machines.
- Proofs use invariants, proved by induction.
- Abstraction relations.
- New complexity measures:
 - Time complexity: Rounds, or real time.
 - Communication complexity:
 Messages, or bits.

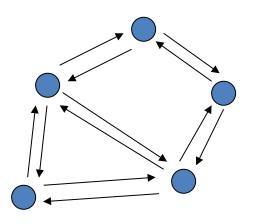


Distributed Networks

- Based on an undirected graph G = (V, E).
 - -n = |V|
 - $-\Gamma(u)$, set of neighbors of vertex u.
 - $-\deg(u)=|\Gamma(u)|$, number of neighbors of vertex u.
- Associate a process with each graph vertex.
 - An infinite-state automaton.
 - Sometimes refer to processes, or vertices, as nodes.
- Associate two directed communication channels with each edge.
- Note: We won't consider failures.

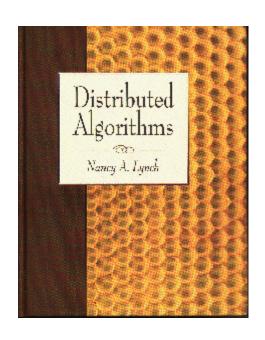


Synchronous Distributed Algorithms



Synchronous Network Model

- Lynch, Distributed Algorithms, Chapter 2.
- Processes at nodes of an undirected graph, communicate using messages.
- Each process has output ports, input ports that connect to communication channels.
 - Process at vertex u doesn't know who its ports' channels connect to.
 - Knows the ports only by local names, such as 1, 2, ..., k, where $k = \deg(u)$.
- Processes need not be distinguishable.
 - E.g., they needn't have Unique Identifiers (UIDs).
 - They just know how many ports they have, and their local names.

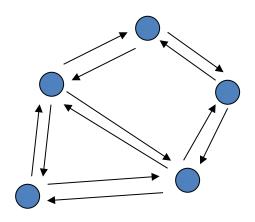


Execution

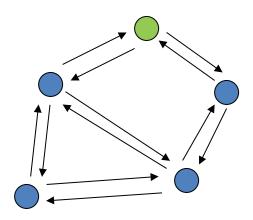
- An algorithm executes in synchronous rounds.
- In each round, each process determines, based on its state, the messages to send on all its ports.
 - At most one message per port per round.
- Each message gets put into its channel, and delivered to the process at the other end.
- Then each process computes a new state based on its old state and the arriving messages.

Note on costs:

- Generally ignore costs of local computation (time and space).
- Focus on time (number of rounds), and communication (number of messages or total number of bits).

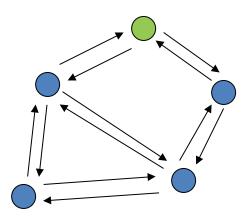


Leader Election



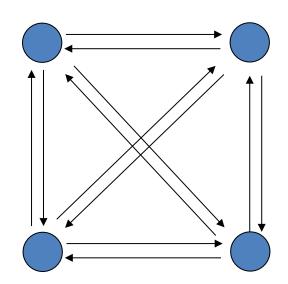
Leader Election

- G = (V, E) is an arbitrary connected (undirected) graph.
- Goal is for exactly one process to output a special leader signal.
- Motivation: Leader can take charge of:
 - Communication
 - Coordinating data processing
 - Allocating resources
 - Scheduling tasks
 - Reaching consensus
 - **–** ...



Simple case: Clique Network

- All vertices are directly connected to all others.
- Theorem 1: Let G = (V, E) be an nvertex clique. Then there is no algorithm
 consisting of deterministic,
 indistinguishable processes that is
 guaranteed to elect a leader in G.



Proof:

- For example, consider n = 2.
- Two identical, deterministic processes.
- Show by induction that the processes remain in the same state forever...

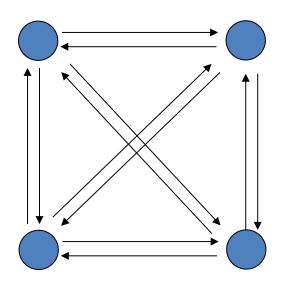


Proof

- By contradiction. Assume an algorithm A that solves the problem.
- Both processes begin in the same start state.
- Prove by induction on the number r of completed rounds that both processes are in identical states after r rounds.
 - Generate the same message.
 - Receive the same message.
 - Make the same state change.
- Since the algorithm solves the leader election problem, eventually one of them gets elected.
- Then they both get elected, contradiction.

Clique Network

• Theorem 1: Let G = (V, E) be an n-vertex clique. Then there is no algorithm consisting of deterministic, indistinguishable processes that is guaranteed to elect a leader in G.



Proof:

- Now consider n > 2.
- Processes have n-1 output ports and n-1 input ports, each set numbered $1,2,\ldots,n-1$.
- Assume ports are connected "consistently": output port k at a process is connected to input port k at the other end.
- Show by induction that all the processes remain in the same state forever...

Proof

- Assume an algorithm A that solves the leader election problem.
- All processes start in the same start state.
- Prove by induction on the number r of completed rounds that all processes are in identical states after r rounds.
 - For each k, they generate the same message on port k.
 - For each k, they receive the same message on port k.
 - Make the same state changes.
- Since the algorithm solves the leader election problem, eventually some process gets elected.
- Then everyone gets elected, contradiction.

- A basic problem for distributed algorithms: Breaking symmetry
- Deterministic, indistinguishable processes can't do it.
- So we need something more...

So we need something more...

Unique Identifiers (UIDs)

- Assume processes have unique identifiers (UIDs), which they "know", e.g., each process starts with its own UID in a special state variable.
- UIDs are elements of some totally-ordered set, e.g., the natural numbers.
- Different UIDs can appear anywhere in the graph, but each can appear only once.

Randomness

Algorithm Using UIDs

- Theorem 2: Let G = (V, E) be an n-vertex clique. Then there is an algorithm consisting of deterministic processes with UIDs that is guaranteed to elect a leader in G.
- The algorithm takes only 1 round and uses only n^2 point-to-point messages.

Algorithm:

- Everyone sends its UID on all its output ports, and collects
 UIDs received on all its input ports.
- The process with the maximum UID elects itself the leader.

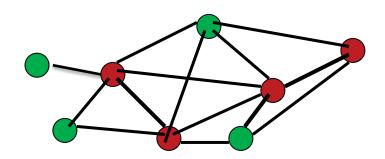
Algorithm Using Randomness

- Idea: Processes choose IDs randomly, from a sufficiently large set so that it is likely that all are different. Then use them like UIDs.
- Q: What is a "sufficiently large" set?
- Lemma 3: Let ϵ be a real, $0 < \epsilon < 1$. Suppose that n processes choose ids uniformly at random, independently, from $\{1, ..., r\}$, where $r = \lceil n^2 / 2 \epsilon \rceil$. Then with probability at least 1ϵ , all the chosen numbers are different.
- Proof: The probability of any two particular processes choosing the same number is $\frac{1}{r}$. Taking a union bound for all $\frac{n^2}{2}$ pairs, the probability is still $\leq \epsilon$.

Algorithm Using Randomness

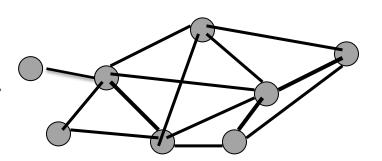
- Theorem 4: Let G = (V, E) be an n-vertex clique. Then there is an algorithm consisting of randomized, indistinguishable processes that eventually elects a leader in G, with probability 1.
- The algorithm takes expected time $\leq \frac{1}{1-\epsilon}$.
- Also, with probability $\geq 1 \epsilon$, the algorithm finishes in only one round.
- Algorithm:
 - Processes choose random ids from a sufficiently large space.
 - Exchange ids; if the maximum is unique, the maximum wins.
 - Otherwise repeat, as many times as necessary.

Maximal Independent Set

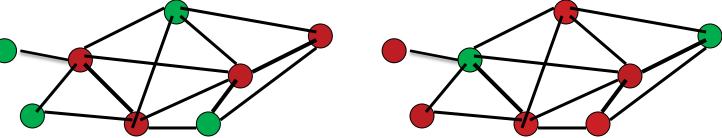


Maximal Independent Set (MIS)

- General undirected graph network:
- Problem: Select a subset S of the nodes, so that they form a Maximal Independent Set.



- Independent: No two neighbors are both in the set.
- Maximal: We can't add any more nodes without violating independence.



• Every node is either in S or has a neighbor in S.

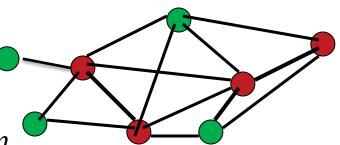
Distributed MIS

Assume:

- No UIDs
- Processes know a good upper bound on n.

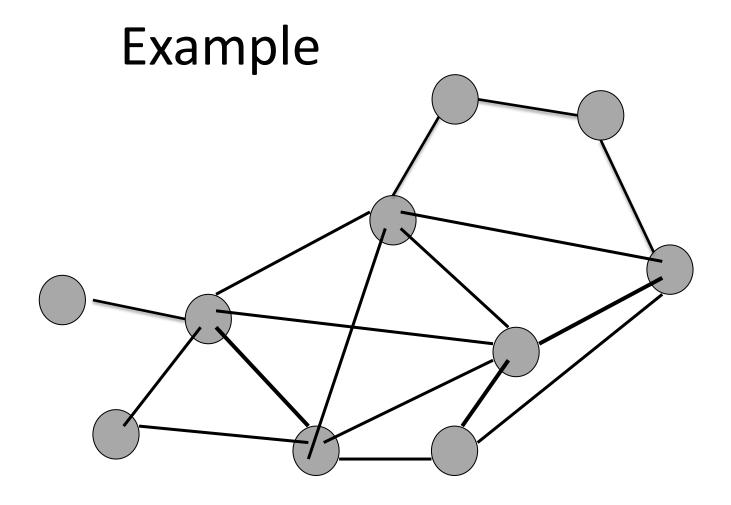


- Compute an MIS S of the entire network graph.
- Each process in S should output in, others output out.
- Unsolvable by deterministic algorithms, in some graphs.
- So consider randomized algorithms.
- Applications of distributed MIS:
 - Communication networks: Selected processes can take charge of communication, convey information to their neighbor processes.
 - Developmental biology: Distinguish cells in fruit fly's nervous system to become "Sensory Organ Precursor" cells [Afek, Alon, et al., Science].

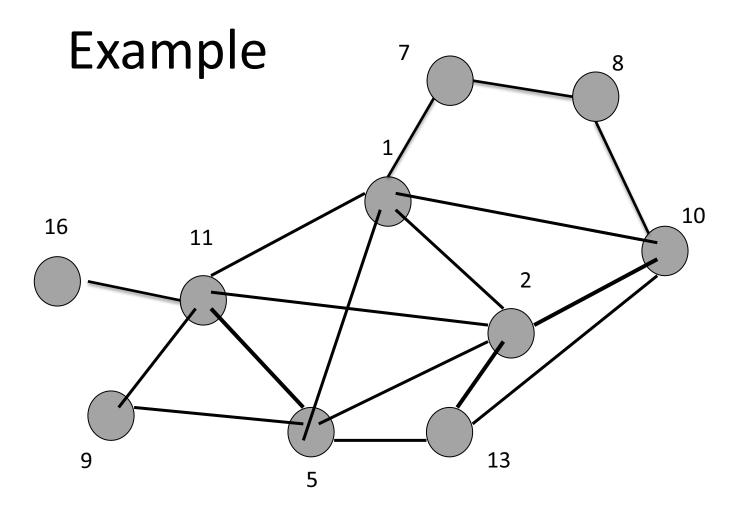


Luby's MIS Algorithm

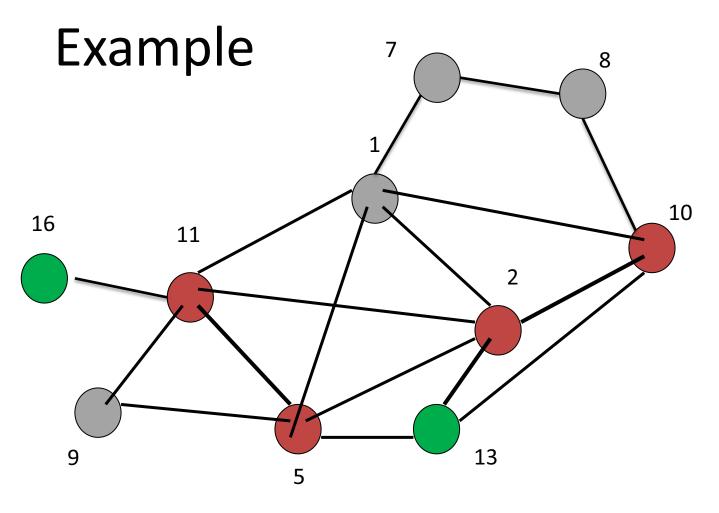
- Executes in 2-round phases.
- Initially all nodes are active.
- At each phase, some active nodes decide to be in, others decide to be out, algorithm continues to the next phase with a smaller graph.
- Repeat until all nodes have decided.
- Behavior of active node u at phase ph:
- Round 1:
 - Choose a random value r in $\{1,2,\ldots,n^5\}$, send it to all neighbors.
 - Receive values from all active neighbors.
 - If r is strictly greater than all received values, then join the MIS, output in.
- Round 2:
 - If you joined the MIS, announce it in messages to all (active) neighbors.
 - If you receive such an announcement, decide not to join the MIS, output out.
 - If you decided one way or the other at this phase, become inactive.



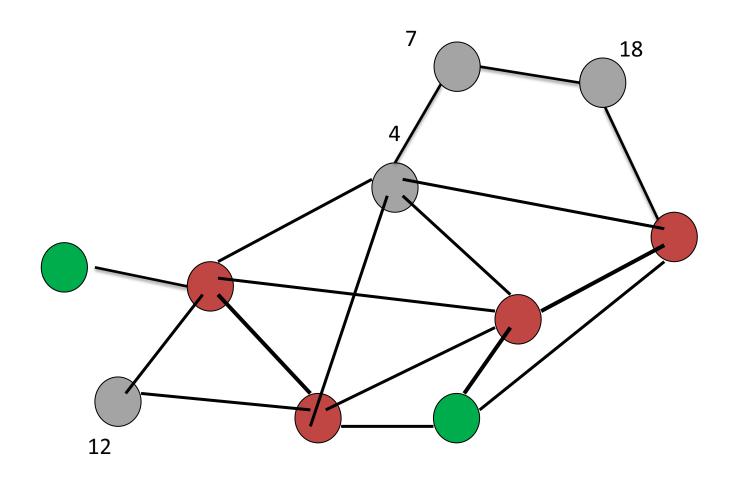
• All nodes start out identical.



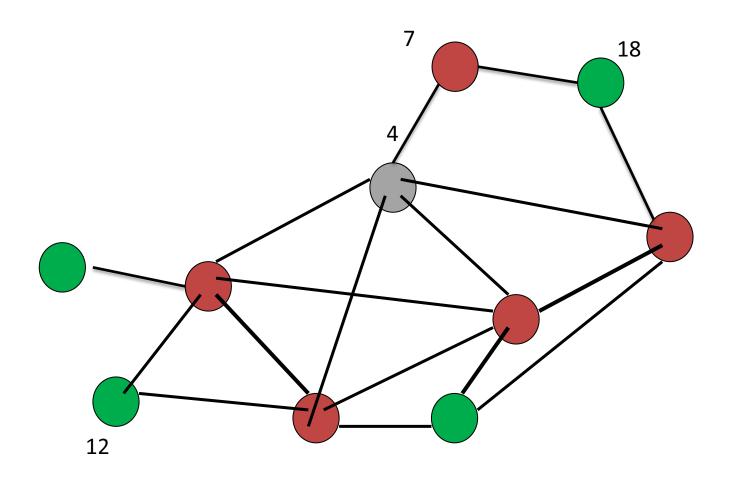
• Everyone chooses an ID.



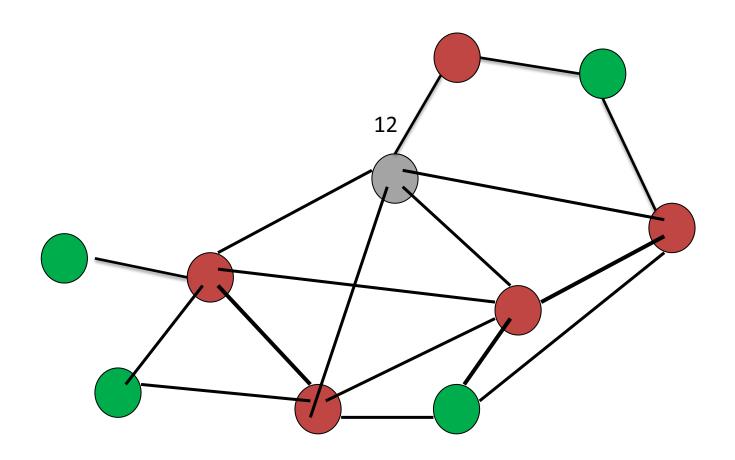
- Processes that chose 16 and 13 are in.
- Processes that chose 11, 5, 2, and 10 are out.



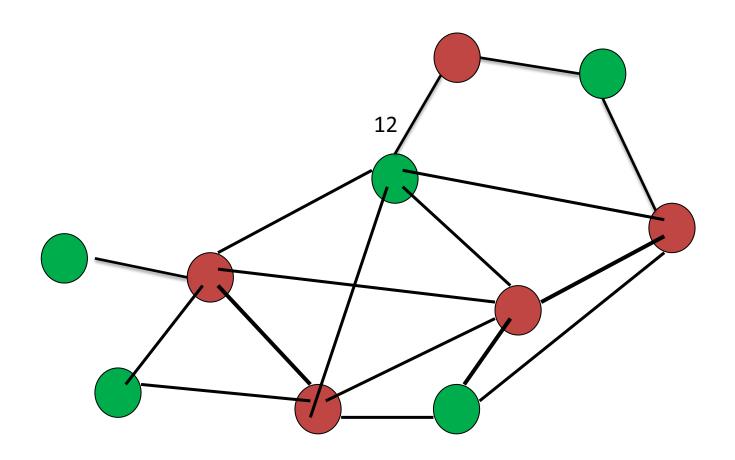
• Undecided (gray) processes choose new IDs.



- Processes that chose 12 and 18 are in.
- Process that chose 7 is out.



• Undecided (gray) process chooses a new ID.



• It's in.

Independence

• Theorem 5: If Luby's algorithm ever terminates, then the final set *S* satisfies the independence property.

Proof:

- Each node joins S only if it has the unique maximum value in its neighborhood, at some phase.
- When it does, all its neighbors become inactive.

Maximality

• Theorem 6: If Luby's algorithm ever terminates, then the final set *S* satisfies the maximality property.

Proof:

- A node becomes inactive only if it joins S or a neighbor joins S.
- We continue until all nodes are inactive.

Termination

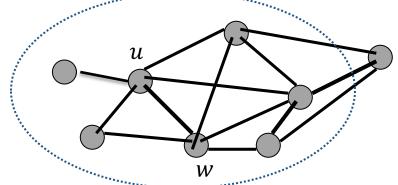
- With probability 1, Luby's MIS algorithm eventually terminates.
- Theorem 7: With probability at least $1 \frac{1}{n}$, all nodes decide within $4 \log n$ phases.
- Proof uses a lemma similar to before:
- Lemma 8: With probability at least $1 \frac{1}{n^2}$, in each phase 1, ..., $4 \log n$, all nodes choose different random values.
- So we can essentially pretend that, in each phase, all the random numbers chosen are different.
- Key idea: Show the graph gets sufficiently "smaller" in each phase.
- Lemma 9: For each phase ph, the expected number of edges that are live (connect two active nodes) at the end of the phase is at most half the number that were live at the beginning of the phase.

Termination

• Lemma 9: For each phase ph, the expected number of edges that are live (connect two active nodes) after the phase is at most half the number that were live before the phase.

• Proof:

- If node u has some neighbor w whose chosen value is greater than all of w's neighbors and all of u's other neighbors, then u must decide out in phase ph.
- Probability that w chooses such a value is at $\frac{1}{\deg(u) + \deg(w)}$.
- Then the probability node u is "killed" by some neighbor in this way is at least $\sum_{w \in \Gamma(u)} \frac{1}{\deg(u) + \deg(w)}$.
- Now consider an edge $\{u, v\}$.



Termination, cont'd

Proof:

- Probability u killed $\geq \sum_{w \in \Gamma(u)} \frac{1}{\deg(u) + \deg(w)}$.
- Probability that edge $\{u, v\}$ "dies" $\geq \frac{1}{2}$ (probability u killed + probability v killed).
- So the expected number of edges that die $\geq \frac{1}{2} \, \Sigma_{\{u,v\}} \ (\text{probability } u \ \text{ killed} + \text{probability } v \ \text{killed}).$
- The sum includes the "kill probability" for each node u exactly deg(u) times.
- So rewrite the sum as:

$$\frac{1}{2} \Sigma_u \deg(u)$$
 (probability u killed).

– Plug in the probability lower bound:

$$\geq \frac{1}{2} \sum_{u} \deg(u) \sum_{w \in \Gamma(u)} \frac{1}{\deg(u) + \deg(w)}.$$

$$= \frac{1}{2} \sum_{u} \sum_{w \in \Gamma(u)} \frac{\deg(u)}{\deg(u) + \deg(w)}.$$

Termination, cont'd

Proof:

- Expected number of edges that die $\geq \frac{1}{2} \sum_{u} \sum_{w \in \Gamma(u)} \frac{\deg(u)}{\deg(u) + \deg(w)}$.
- Write this expression equivalently as a sum over directed edges (u, v):

$$\frac{1}{2} \sum_{(u,v)} \frac{\deg(u)}{\deg(u) + \deg(v)}.$$

- Here each undirected edge is counted twice, once for each direction, so this is the same as the following sum over undirected edges $\{u, v\}$.

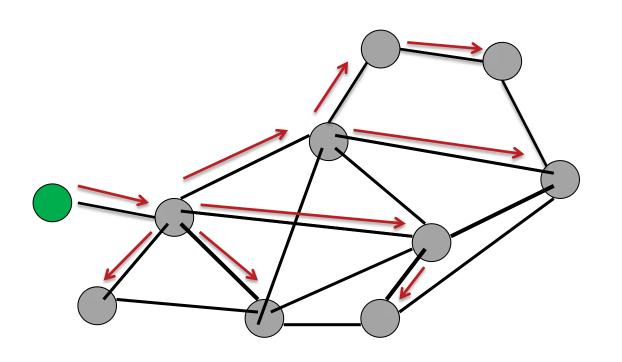
$$\frac{1}{2}\sum_{\{u,v\}}\frac{\deg(u)+\deg(v)}{\deg(u)+\deg(v)}.$$

- This is half the total number of undirected edges, as needed!
- Lemma 9: For each phase ph, the expected number of edges that
 are live (connect two active nodes) at the end of the phase is at
 most half the number that were live at the beginning of the phase.

Termination, cont'd

- Lemma 9: For each phase ph, the expected number of edges that are live (connect two active nodes) at the end of the phase is at most half the number that were live before the phase.
- Theorem 7: With probability at least $1 \frac{1}{n}$, all nodes decide within $4 \log n$ phases.
- Proof sketch:
 - Lemma 9 implies that the expected number of edges still live after $4 \log n$ phases is at most $\frac{n^2}{2} \div 2^{4 \log n} = \frac{1}{2n^2}$.
 - Then the probability that *any* edges remain live is $\leq \frac{1}{2n^2}$ (by Markov).
 - The probability that the algorithm doesn't terminate within $4 \log n$ phases $\leq \frac{1}{2n^2} + \frac{1}{n^2} < \frac{1}{n}$.

Breadth-First Spanning Trees

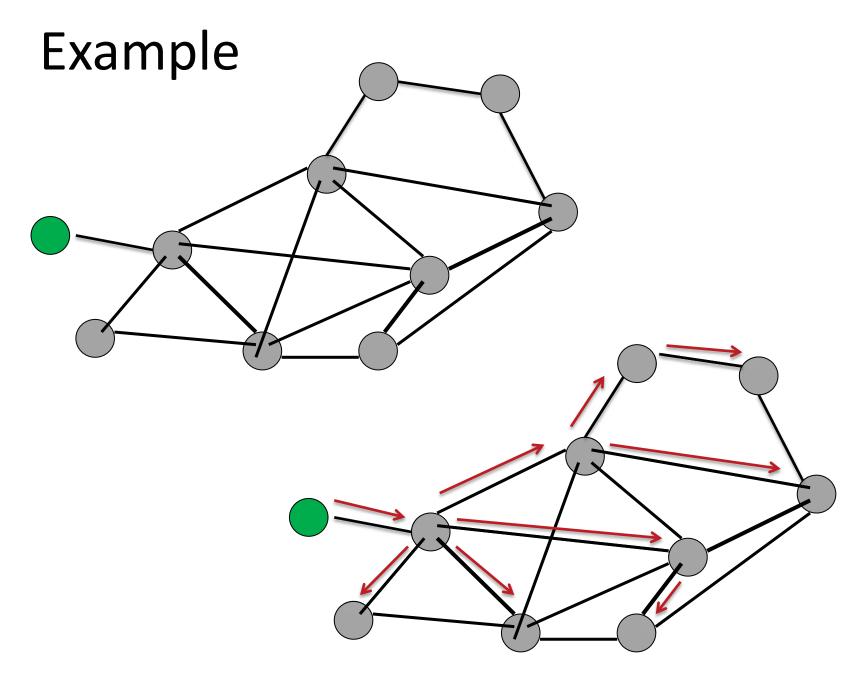


Breadth-First Spanning Trees

- New problem, new setting.
- Assume graph G = (V, E) is connected.
- V includes a distinguished vertex v_0 , which will be the origin (root) of the BFS tree.
- Generally, processes have no knowledge about the graph.
- Processes have UIDs.
 - Each process knows its own UID.
 - i_0 is the UID of the root v_0 .
 - Process with UID i_0 knows it is located at the root.
- We may assume (WLOG) that processes know the UIDs of their neighbors, and know which input and output ports are connected to each neighbor.
- Algorithms will be deterministic (or nondeterministic), but not randomized.

Breadth-First Spanning Trees

- Processes must produce a Breadth-First Spanning Tree rooted at vertex v_0 .
- Branches are directed paths from v_0 .
 - Spanning: Branches reach all vertices.
 - Breadth-first: Vertex at distance d from v_0 appears at depth exactly d in the tree.
- Output: Each process $i \neq i_0$ should output parent(j), meaning that j's vertex is the parent of i's vertex in the BFS tree.



Simple BFS Algorithm

- Processes mark themselves as they get incorporated into the tree.
- Initially, only i_0 is marked.
- Algorithm for process i:
 - Round 1:
 - If $i = i_0$ then process i sends a *search* message to its neighbors.
 - If process *i* receives a message, then it:
 - Marks itself.
 - Selects i_0 as its parent, outputs $parent(i_0)$.
 - Plans to send at the next round.
 - Round r > 1:
 - If process i planned to send, then it sends a search message to its neighbors.
 - If process i is not marked and receives a message, then it:
 - Marks itself.
 - Selects one sending neighbor, j, as its parent, outputs parent(j).
 - Plans to send at the next round.

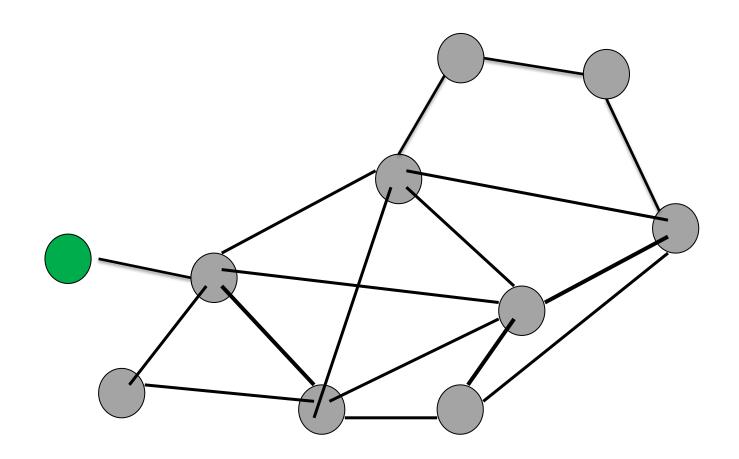
Nondeterminism

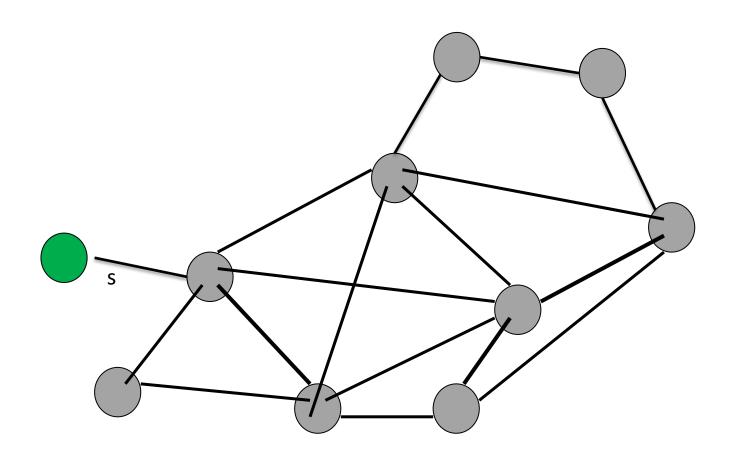
- The algorithm is slightly nondeterministic, in that a process can choose arbitrarily among several possible parents.
- We could make this deterministic, by using a default, like "always choose the sender with the smallest UID".
- But it's also OK to leave it nondeterministic.
- For distributed algorithms, nondeterminism is regarded differently from the way it is for sequential algorithms.
- A distributed algorithm should work correctly for all ways of resolving the nondeterministic choices.

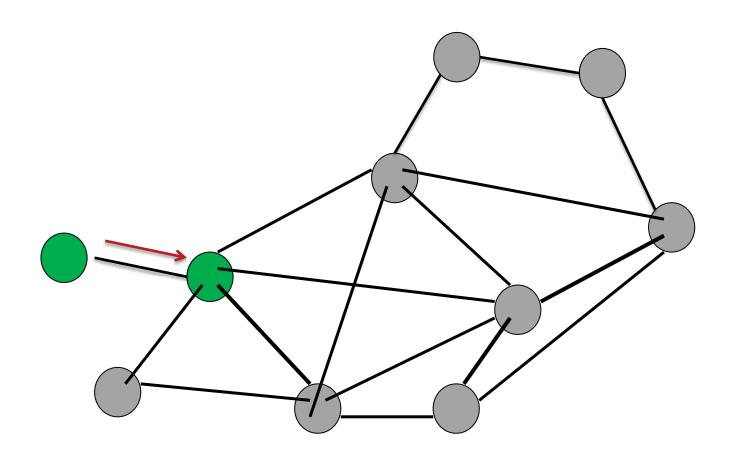
Simple BFS Algorithm

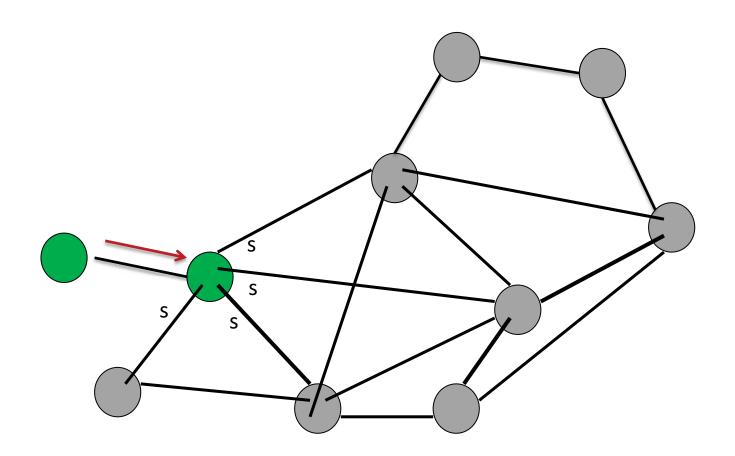
- Q: Why does this algorithm yield a BFS tree?
- Because all branches are created synchronously, growing one hop at each round.
- Q: Why does it eventually span all the nodes?
- Because the graph is connected, and any marked node sends messages to its neighbors.

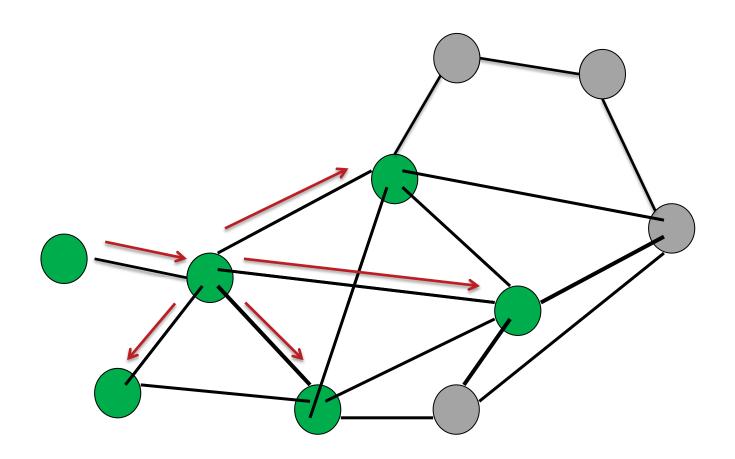
Example: Simple BFS

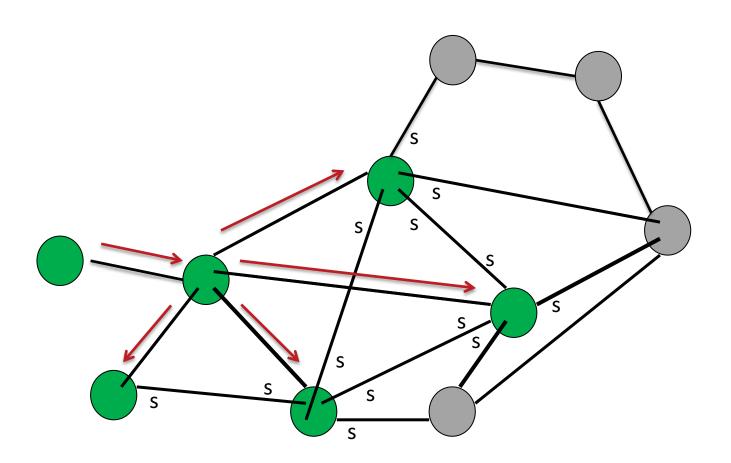


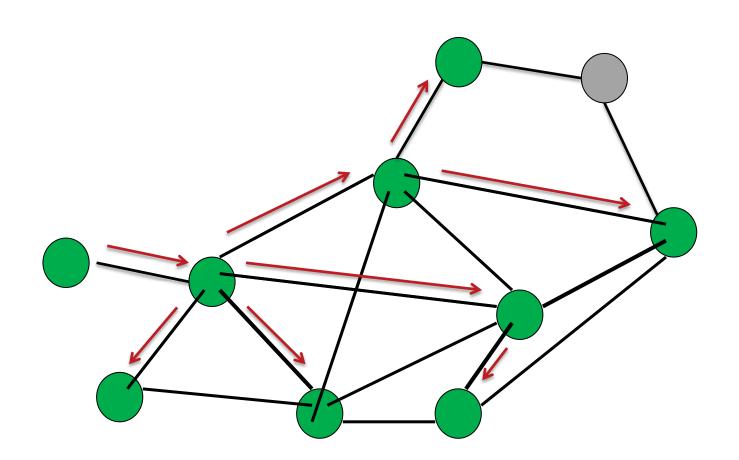


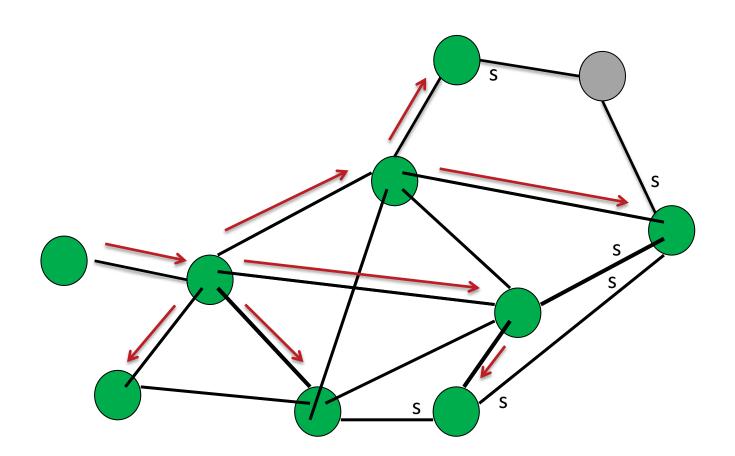


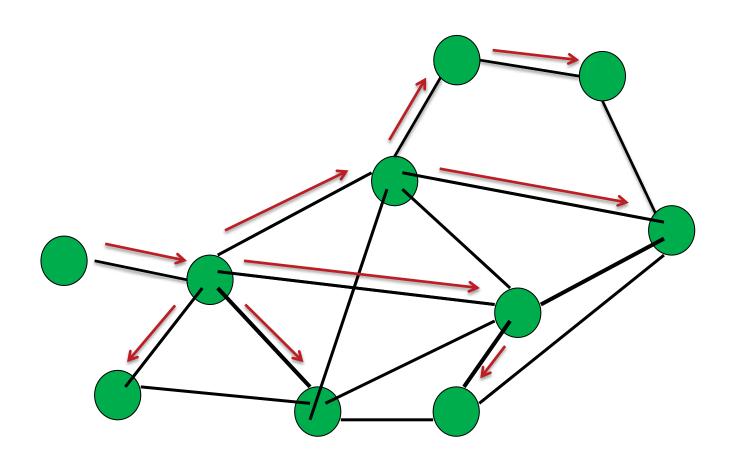












Correctness

- Q: Key invariants describing the state after r rounds?
- State variables, per process:
 - marked, a Boolean, initially true for i_0 , false for others
 - parent, a UID or undefined
 - send, a Boolean, initially true for i_0 , false for others
 - uid

Invariants:

- At the end of r rounds, exactly the processes at distance $\leq r$ from v_0 are marked.
- A process $\neq i_0$ has its *parent* defined iff it is marked.
- For any process at distance d from v_0 , if its parent is defined, then it is the UID of a process at distance d-1 from v_0 .
- Could use these in a formal correctness proof.

Complexity

Time complexity:

- Number of rounds until all nodes outputs their parent information.
- Maximum distance of any node from v_0 , which is $\leq diam$

Message complexity:

- Number of messages sent by all processes during the entire execution.
- O(|E|)

Bells and Whistles

Child pointers:

- So far, each process learns its parent, but not its children.
- To add child pointers, everyone who receives a search message sends back a parent or nonparent response, at the next round.

Distances:

- Augment the algorithm so everyone also learns and outputs its distance from v_0 .
- Each process records its distance from v_0 in a new dist variable, initially 0 for process i_0 , and ∞ for everyone else.
- Include the dist value in every search message; when an unmarked process receives a search(d) message, it sets its own dist to d+1.

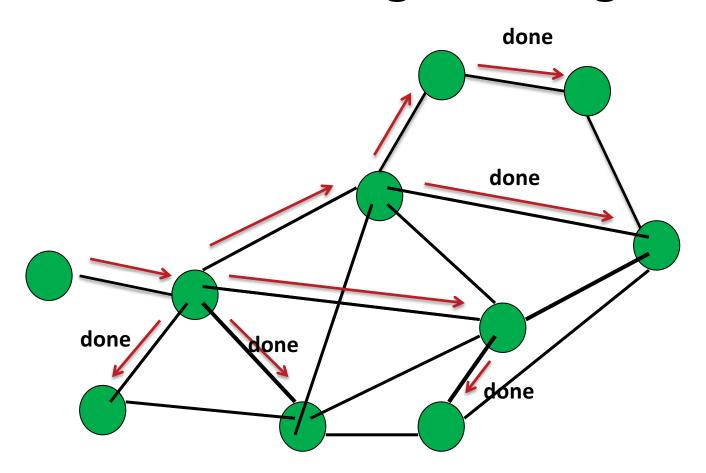
Termination

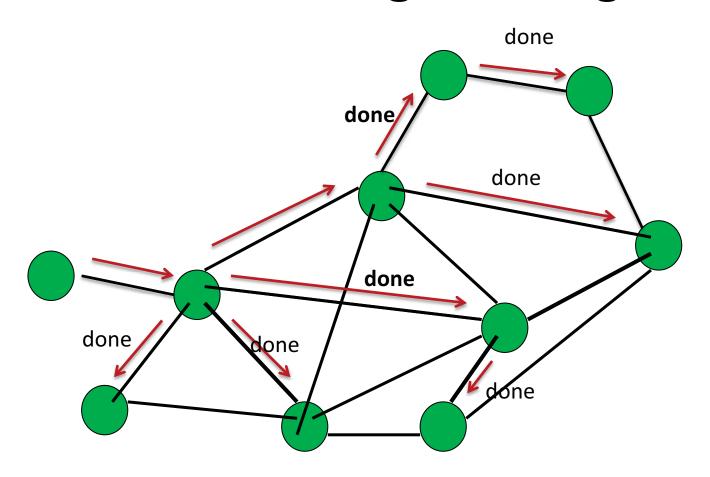
- Q: How can processes learn when the BFS tree is completed?
- If they knew an upper bound on diam, then they could simply wait until that number of rounds have passed.
- Q: What if they don't know anything about the graph?
- Slightly simpler problem: Process i_0 should learn when the tree is done.
- Termination algorithm:
 - Assume each search message receives a response, parent or nonparent.
 - After a node has received responses to all its outgoing search messages, it knows who its children are, and knows they are all marked.
 - The leaves of the tree learn who they are (they receive only nonparent responses).
 - Now use a convergecast strategy:

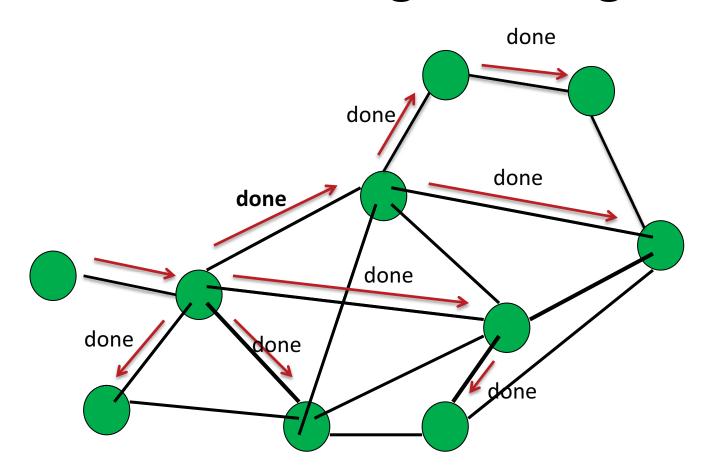
Termination

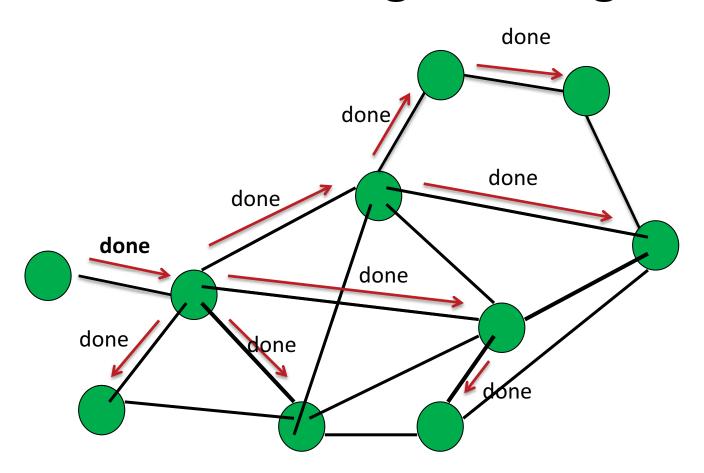
Termination algorithm:

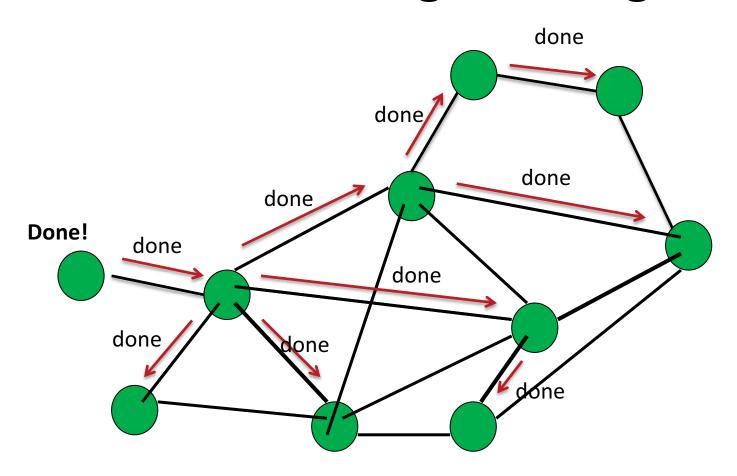
- After a node has received responses to all its
 search messages, it knows who its children are, and knows
 they are marked.
- The leaves of the tree learn who they are.
- Use a convergecast strategy:
 - Starting from the leaves, the processes fan in done messages up the BFS tree, toward i_0 .
 - A process $\neq i_0$ can send a *done* message to its parent after:
 - It has received responses to all its search messages (so it knows who its children are), and
 - It has received done messages from all its children.
 - i_0 knows that the BFS tree is completed, when:
 - It has received responses to all its search messages (so it knows who its children are), and
 - It has received done messages from all its children.











Complexity for Termination

- After the tree is done, it take $s \leq diam$ rounds and n messages for the done information to reach i_0 .
- Q: How can all the processes learn that the BFS tree is completed?
- Process i_0 can broadcast a message to everyone along the edges of the tree.
- Takes an additional diam rounds and n messages.

Applications of Breadth-First Spanning Trees

Message broadcast:

- Process i_0 can use a BFS tree with child pointers to broadcast a sequence of messages to all processes in the network.
- Each takes diam rounds and n messages.
- Could pipeline them, so k messages take only diam + k rounds, not $diam \times k$.

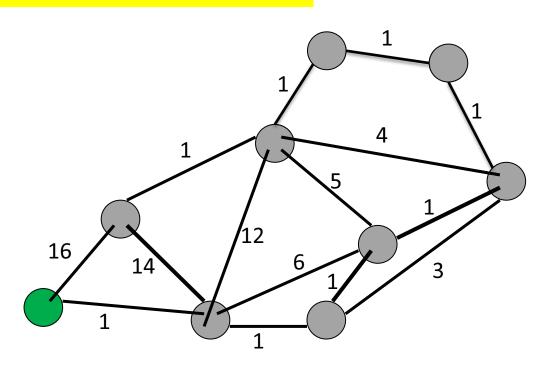
Global computation:

- Suppose every process starts with some initial value, and process i_0 should determine the value of some function of the set of all processes' values.
 - min, max, sum, average,...
- Can use convergecast on a BFS tree.
- Cost is diam rounds and n messages.
- In general, messages may be large, but for many functions, some data aggregation is possible along the way.

Application of the BFS construction

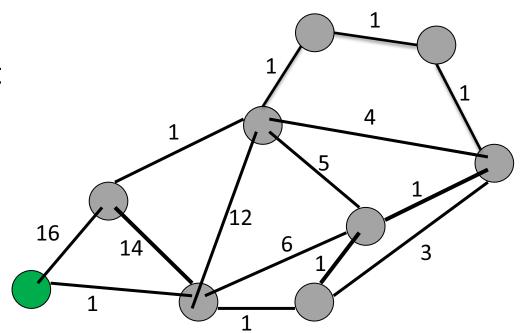
- Leader election in a general graph:
 - No distinguished process i_0 .
 - Processes don't know anything about the graph.
 - Everyone can starts its own BFS, acting as the root.
 - Use this to determine the maximum UID; process with the max UID is the leader.
 - Cost is O(diam) rounds, O(n |E|) messages.

Shortest Paths Trees



Shortest Paths

- Generalize the BFS problem to allow weights on the graph edges, $weight_{\{u,v\}}$ for edge $\{u,v\}$
- Connected graph G = (V, E), root vertex v_0 , process i_0 .
- Processes have UIDs.
- Processes know their neighbors and the weights of their incident edges, but otherwise have no knowledge about the graph.



Shortest Paths

- Processes must produce a Shortest-Paths Spanning Tree rooted at vertex v_0 .
- Branches are directed paths from v_0 .
 - Spanning: Branches reach all vertices.
 - Shortest paths: The total weight of the tree branch to each node is the minimum total weight for any path from v_0 in G.
- Output: Each process $i \neq i_0$ should output parent(j), distance(d), meaning that:
 - j's vertex is the parent of i's vertex on a shortest path from v_0 ,
 - d is the total weight of a shortest path from v_0 to j.

Bellman-Ford Shortest Paths Algorithm

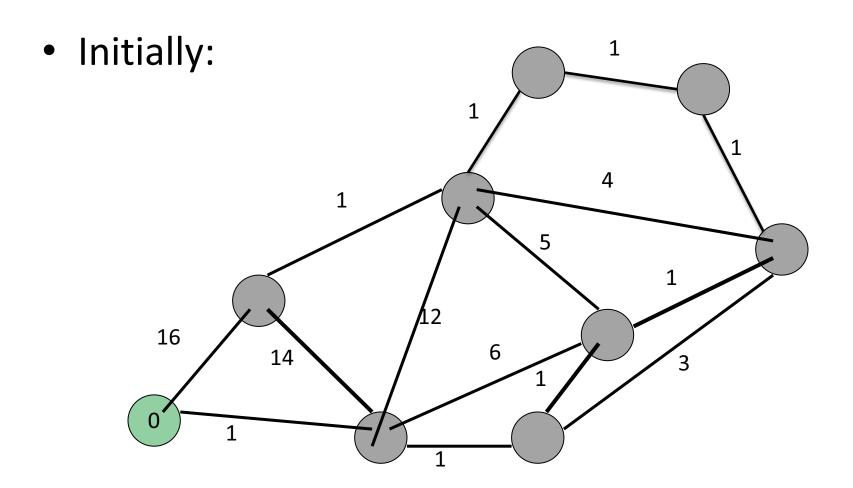
State variables:

- dist, a nonnegative real or ∞ , representing the shortest known distance from v_0 . Initially 0 for process i_0 , ∞ for the others.
- parent, a UID or undefined, initially undefined.
- uid

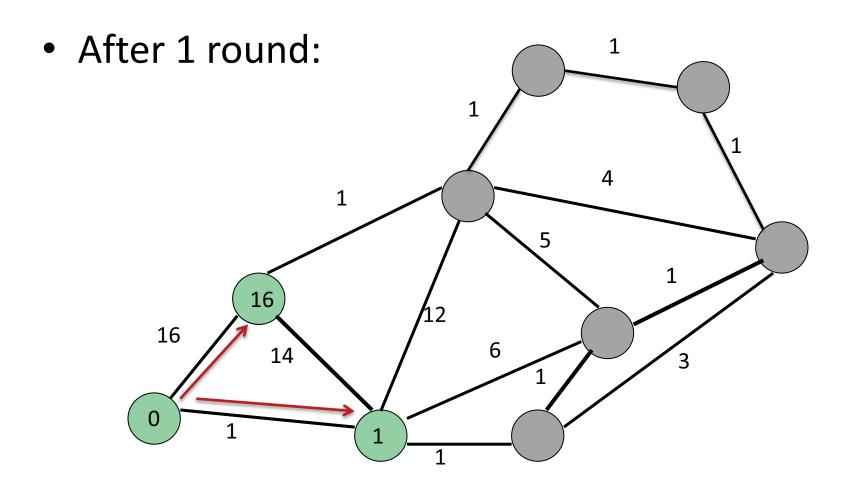
Algorithm for process i:

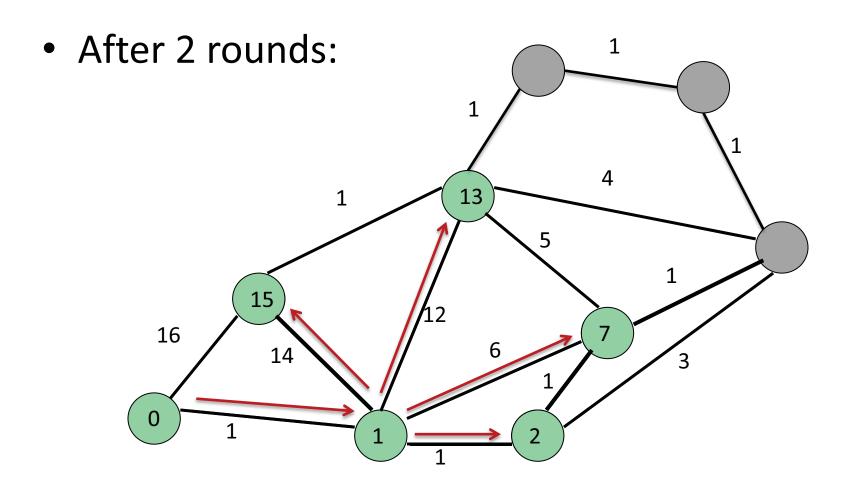
- At each round:
 - Send a distance(dist) message to all neighbors.
 - Receive messages from neighbors; let d_j be the distance received from neighbor j.
 - Perform a relaxation step: $dist := min(dist, min_j(d_j + weight_{\{i,j\}}).$
 - If dist decreases then set parent := j, where j is any neighbor that produced the new dist.

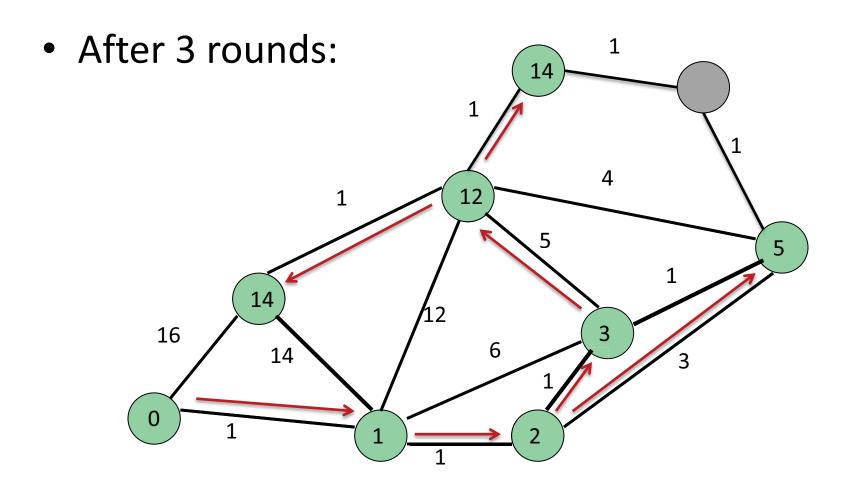
Example: Bellman-Ford

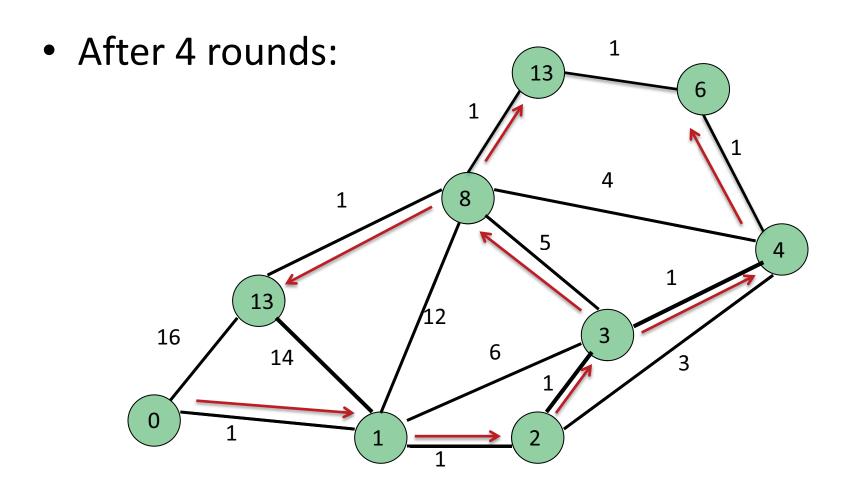


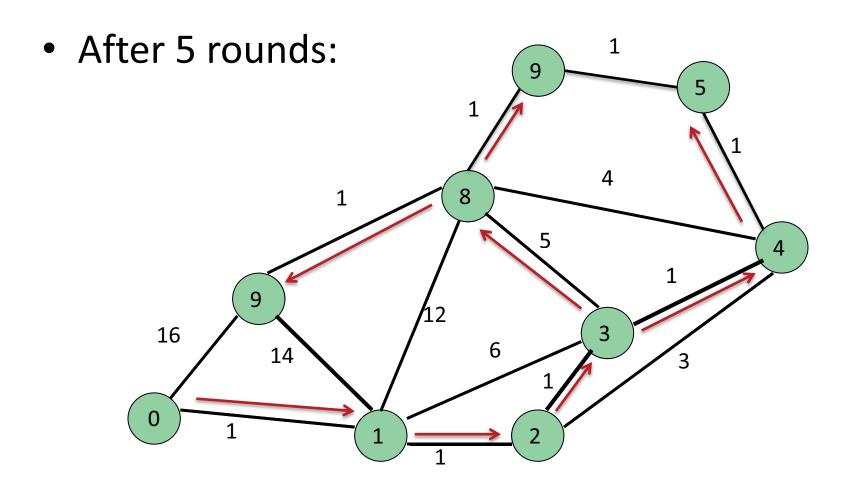
Bellman-Ford

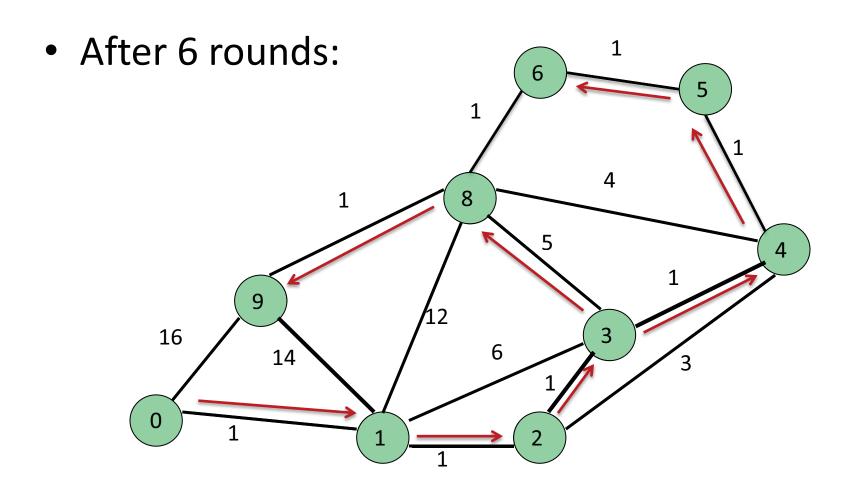


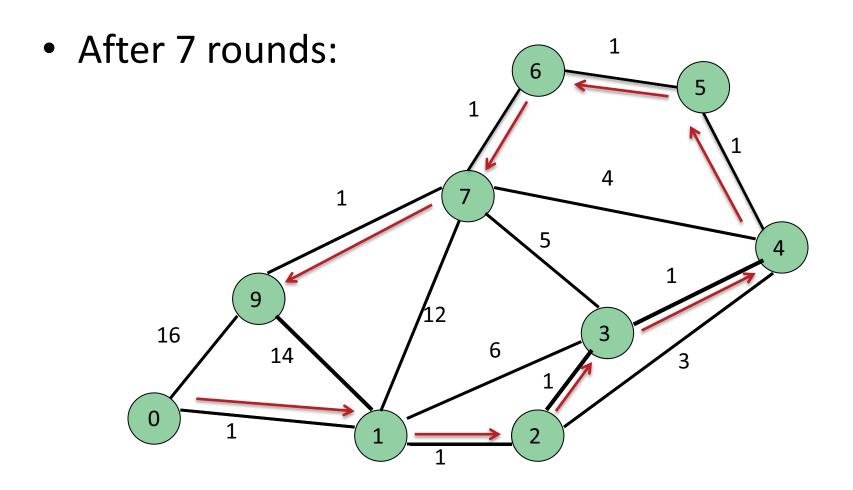


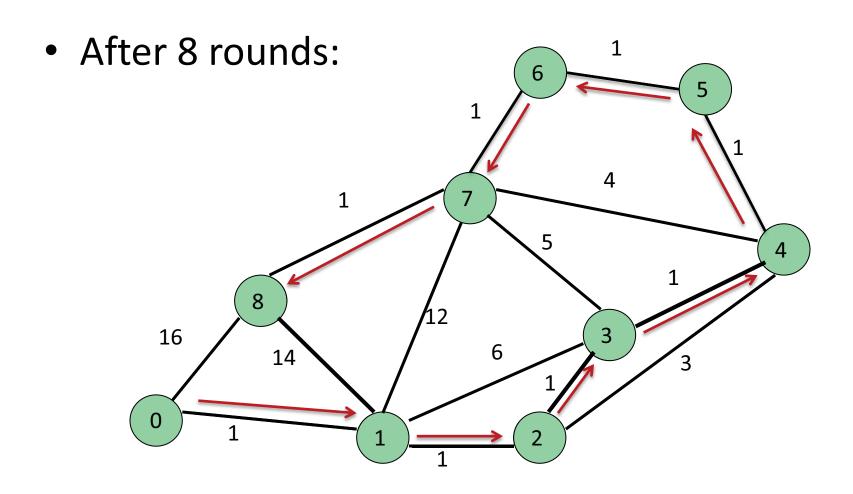












Correctness

- Claim: Eventually, every process i has:
 - dist = minimum weight of a path from i_0 to i, and
 - if $i \neq i_0$, parent = the previous node on some shortest path from i_0 to i.

Key invariant:

— For every r, at the end of r rounds, every process $i \neq i_0$ has its dist and parent corresponding to a shortest path from i_0 to i among those paths that consist of at most r edges; if there is no such path, then $dist = \infty$ and parent is undefined.

Complexity

Time complexity:

- Number of rounds until all the variables stabilize to their final values.
- n-1 rounds
- Message complexity:
 - Number of messages sent by all processes during the entire execution.
 - $O(n \cdot |E|)$
- More expensive than BFS:
 - diam rounds,
 - O(|E|) messages
- Q: Does the time bound really depend on n?

Child Pointers

- Ignore repeated messages.
- When process i receives a message that it does not use to improve dist, it responds with a nonparent message.
- When process i receives a message that it uses to improve dist, it responds with a parent message, and also responds to any previous parent with a nonparent message.
- Process *i* records nodes from which it receives *parent* messages in a set *children*.
- When process *i* receives a *nonparent* message from a current child, it removes the sender from its *children*.
- When process *i* improves *dist*, it empties *children*.

Termination

- Q: How can the processes learn when the shortestpaths tree is completed?
- Q: How can a process know when it can output its own *parent* and *distance*?
- If processes knew an upper bound on n, then they could simply wait until that number of rounds have passed.
- But what if they don't know anything about the graph?
- Recall termination for BFS: Used convergecast.
- Q: Does that work here?

Termination

- Q: How can the processes learn when the shortestpaths tree is completed?
- Q: Does convergecast work here?
- Yes, but it's trickier, since the tree structure changes.

Key ideas:

- A process $\neq i_0$ can send a *done* message to its current parent after:
 - It has received responses to all its *distance* messages, so it believes it knows who its children are, and
 - It has received *done* messages from all of those children.
- The same process may be involved several times in the convergecast, based on improved estimates.

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