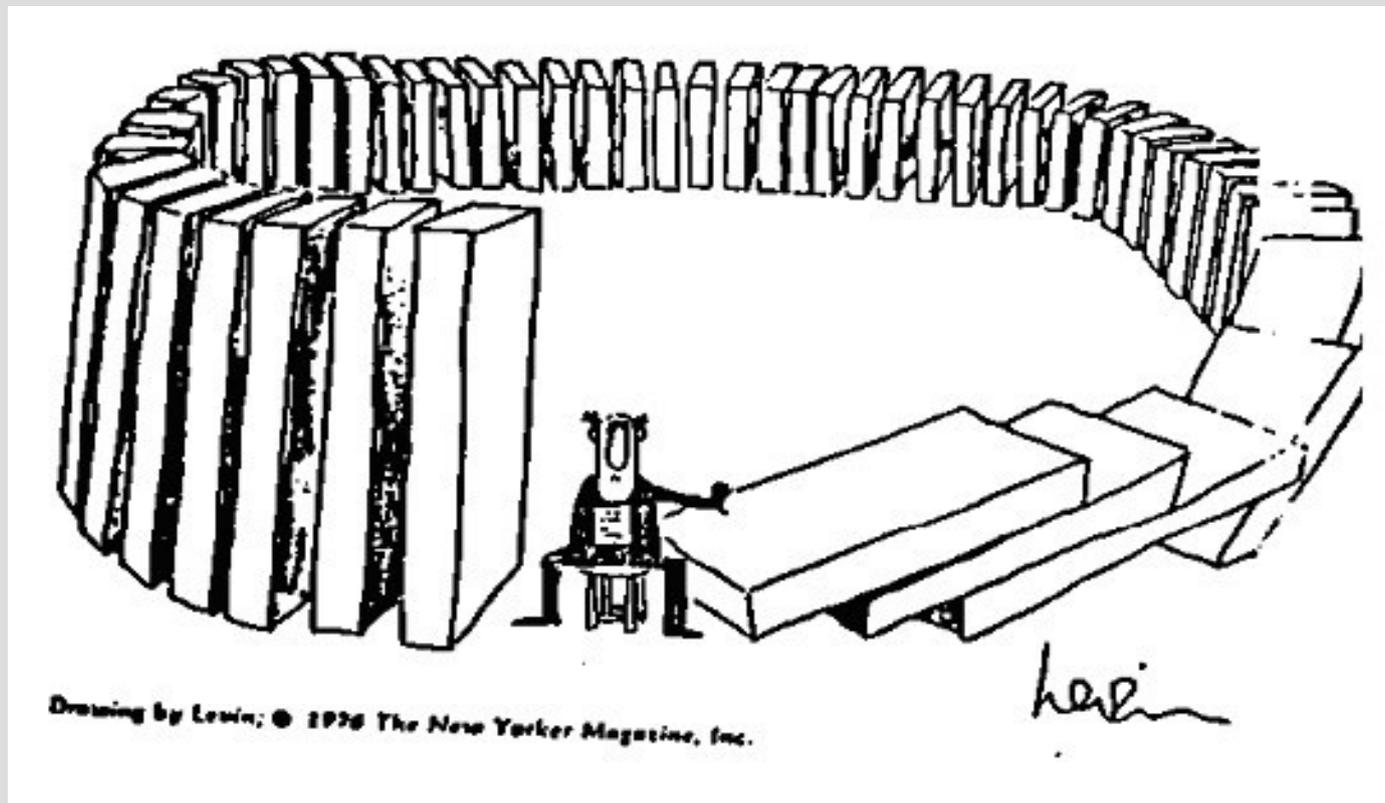


# Unweighted directed graphs



# Announcements

## Midterm & gradescope

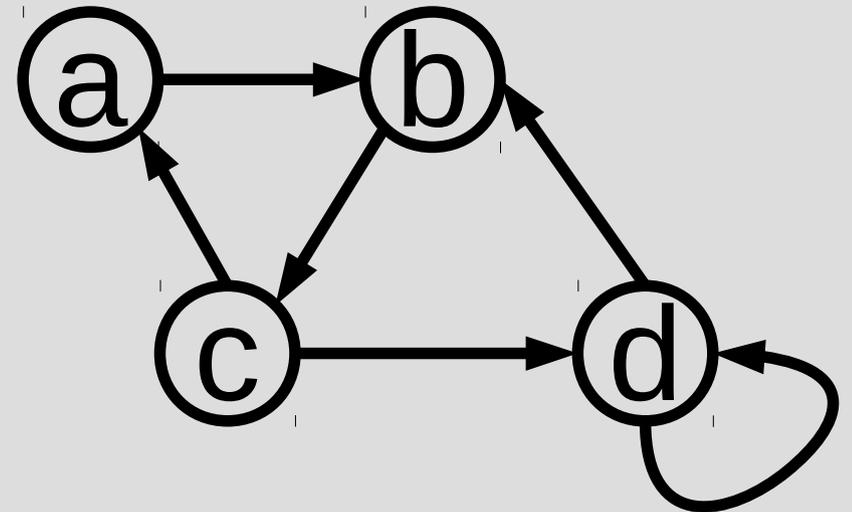
- will get an email today to register  
(username name is your email)
- tests should appear by next Monday  
(nothing there now)

# Graph

A directed graph  $G$  is a set of edges and vertices:  $G = (V, E)$

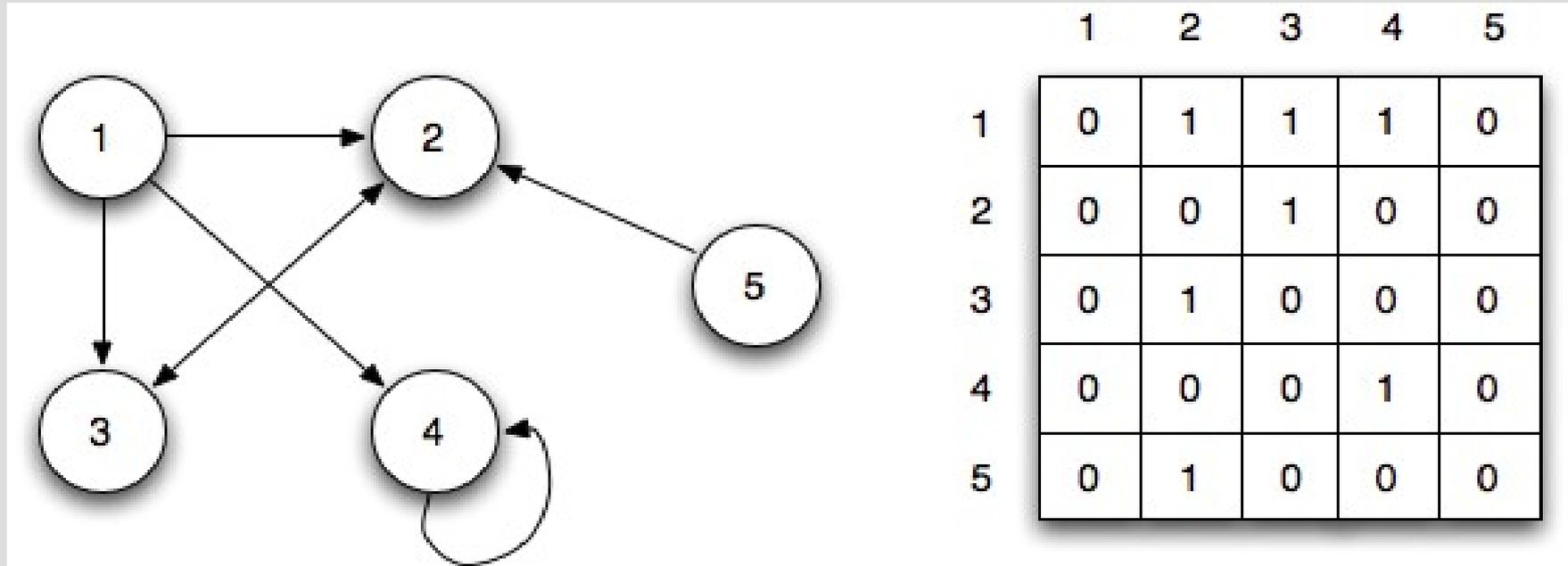
Two common ways to represent a graph:

- Adjacency matrix
- Adjacency list



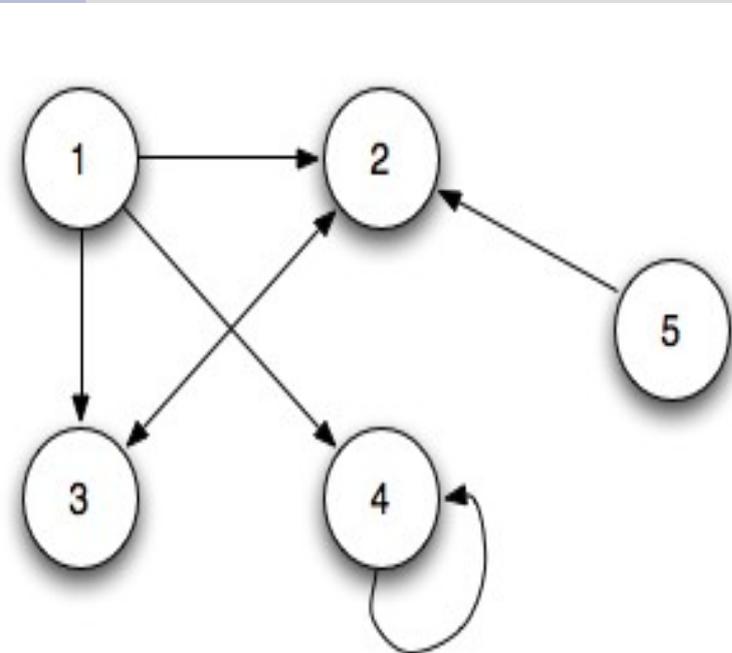
# Graph

An adjacency matrix has a 1 in row  $i$  and column  $j$  if you can go from node  $i$  to node  $j$

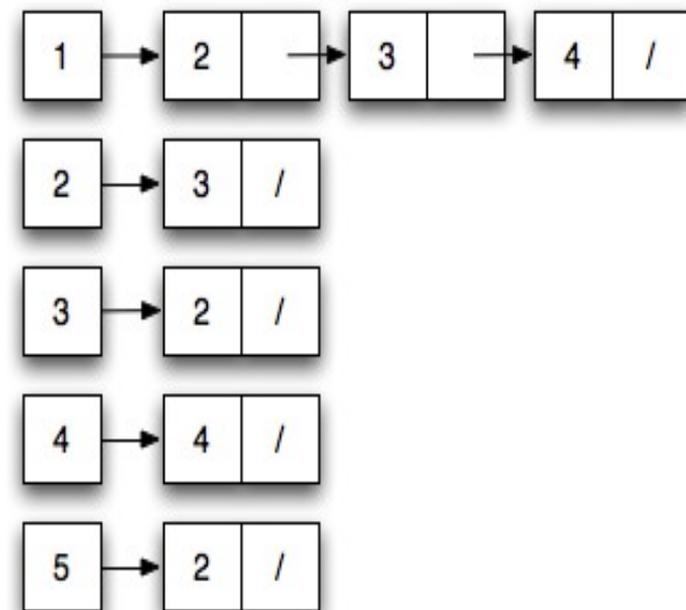


# Graph

An adjacency list just makes lists out of each row (list of edges out from every vertex)



	1	2	3	4	5
1	0	1	1	1	0
2	0	0	1	0	0
3	0	1	0	0	0
4	0	0	0	1	0
5	0	1	0	0	0



# Graph

Difference between adjacency matrix and adjacency list?

# Graph

Difference between adjacency matrix and adjacency list?

Matrix is more memory  $O(|V|^2)$ ,  
less computation:  $O(1)$  lookup

List is less memory  $O(E+V)$  if sparse,  
more computation:  $O(\text{branch factor})$

# Graph

Adjacency matrix,  $A=A^1$ , represents the number of paths from row node to column node in 1 step

Prove:  $A^n$  is the number of paths from row node to column node in  $n$  steps

# Graph

Proof: Induction

Base:  $A^0 = I$ , 0 steps from  $i$  is  $i$

Induction: (Assume  $A^n$ , show  $A^{n+1}$ )

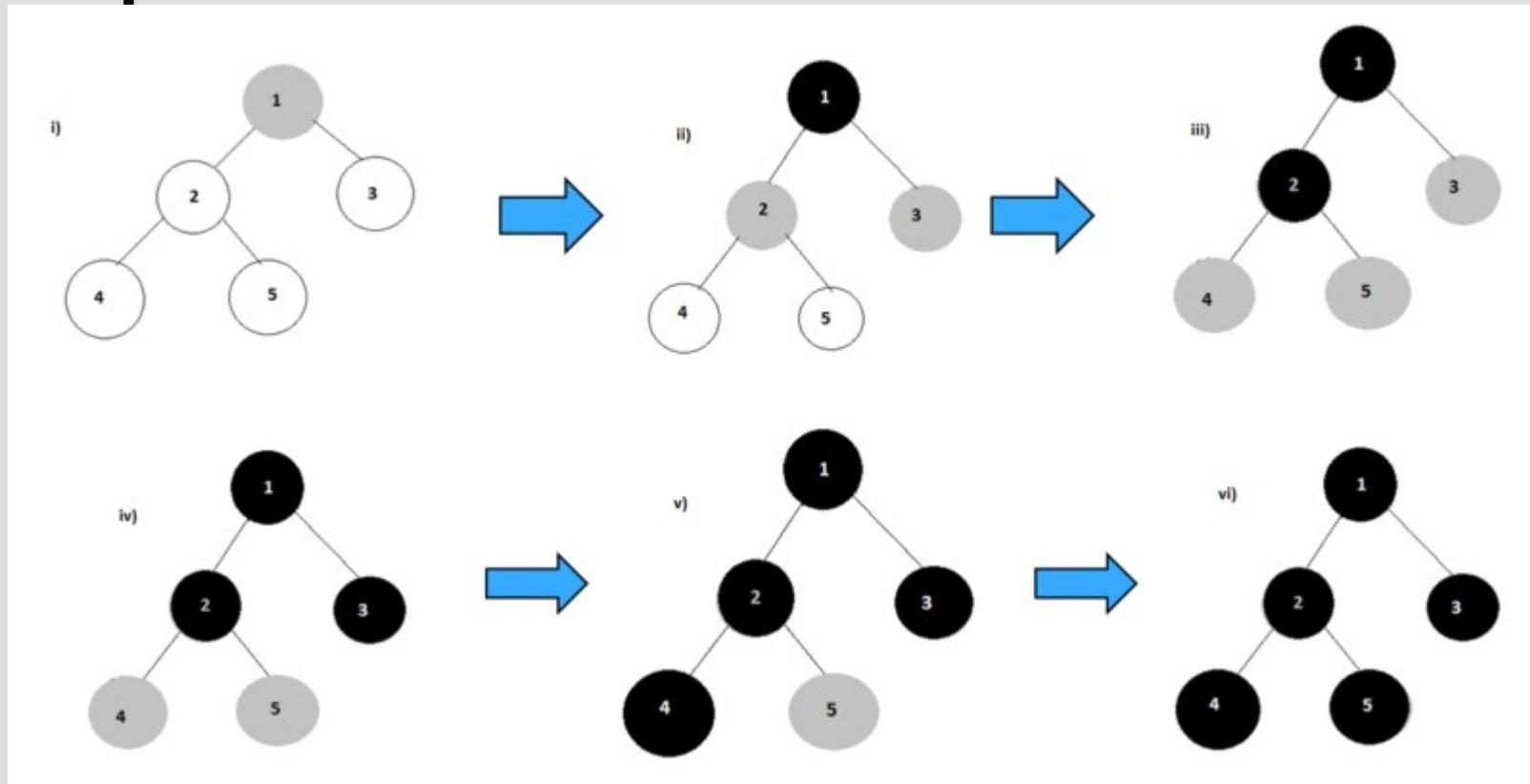
Let  $a_{i,j}^n = i^{\text{th}}$  row,  $j^{\text{th}}$  column of  $A^n$

Then  $a_{i,j}^{n+1} = \sum_k a_{i,k}^n a_{k,j}^1$

This is just matrix multiplication

# Breadth First Search Overview

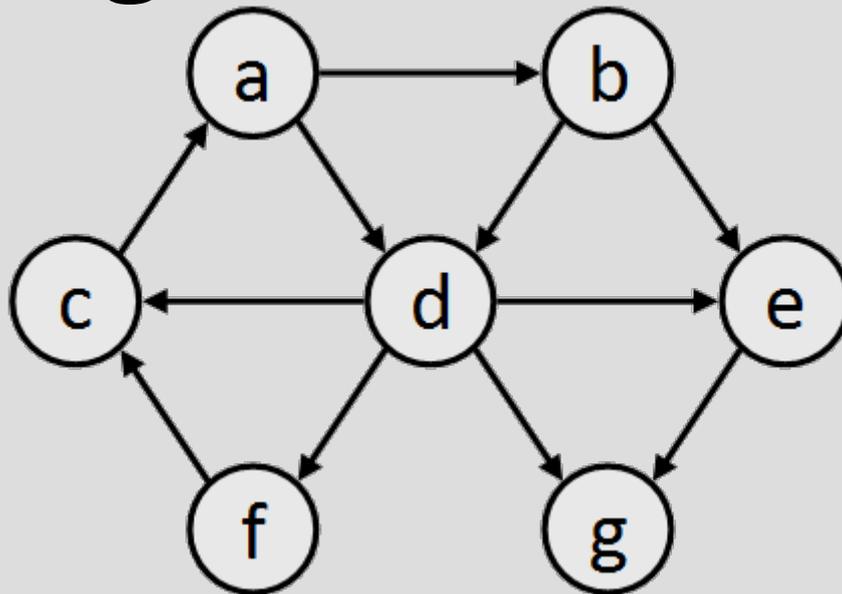
Create first-in-first-out (FIFO) queue to explore unvisited nodes



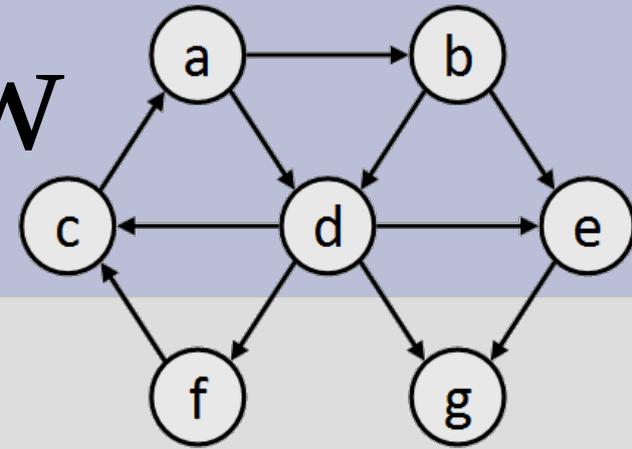
# Breadth First Search Overview

Consider the graph below

Suppose we wanted to get from “a” to “c” using breadth first search



# BFS Overview



To keep track of which nodes we have seen, we will do:

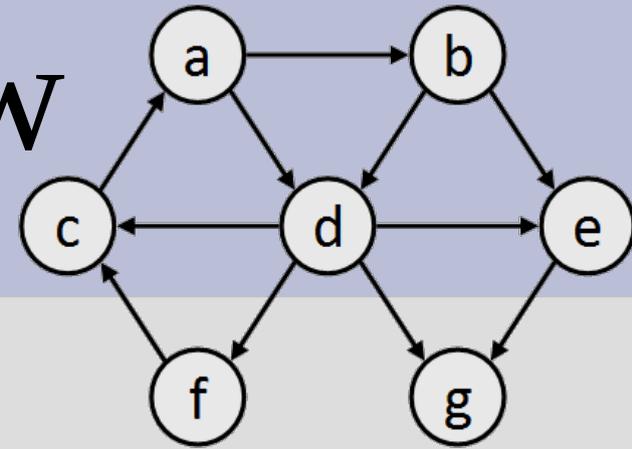
White nodes = never seen before

Grey nodes = nodes in  $Q$

Black nodes = nodes that are done

To keep track of who first saw nodes  
I will make red arrows ( $\pi$  in book)

# BFS Overview



First, we add the start to the queue, so  $Q = \{a\}$

Then we will repeatedly take the left-most item in  $Q$  and add all of its neighbors (that we haven't seen yet) to the  $Q$  on the right

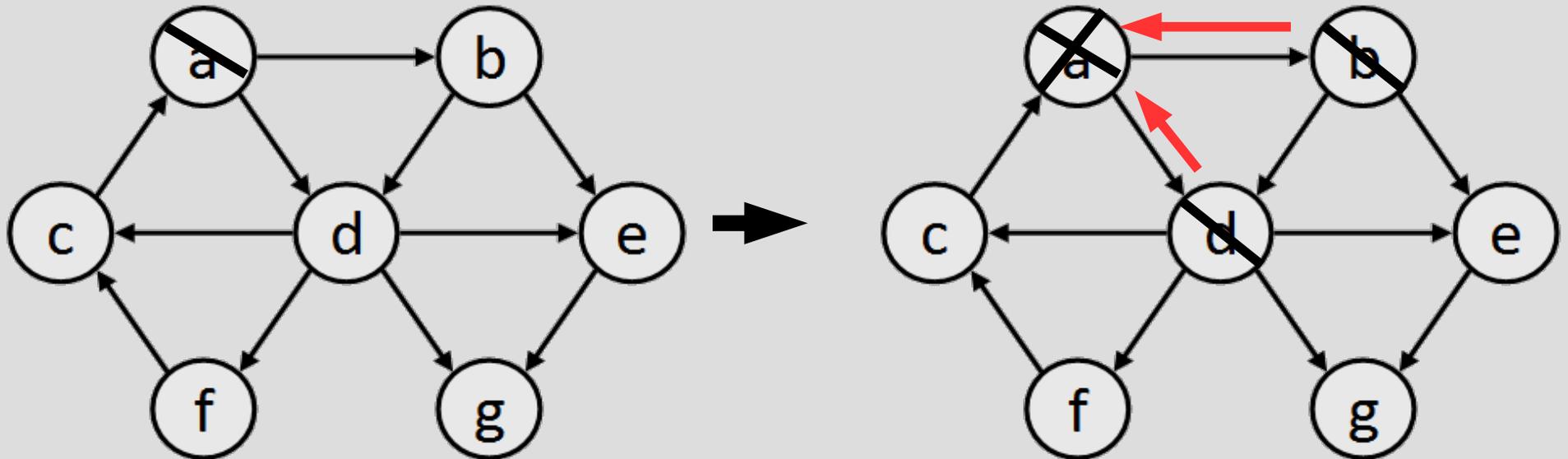
# BFS Overview

$Q = \{a\}$

Left-most = a

White neighbors = b & d

New  $Q = \{b, d\}$



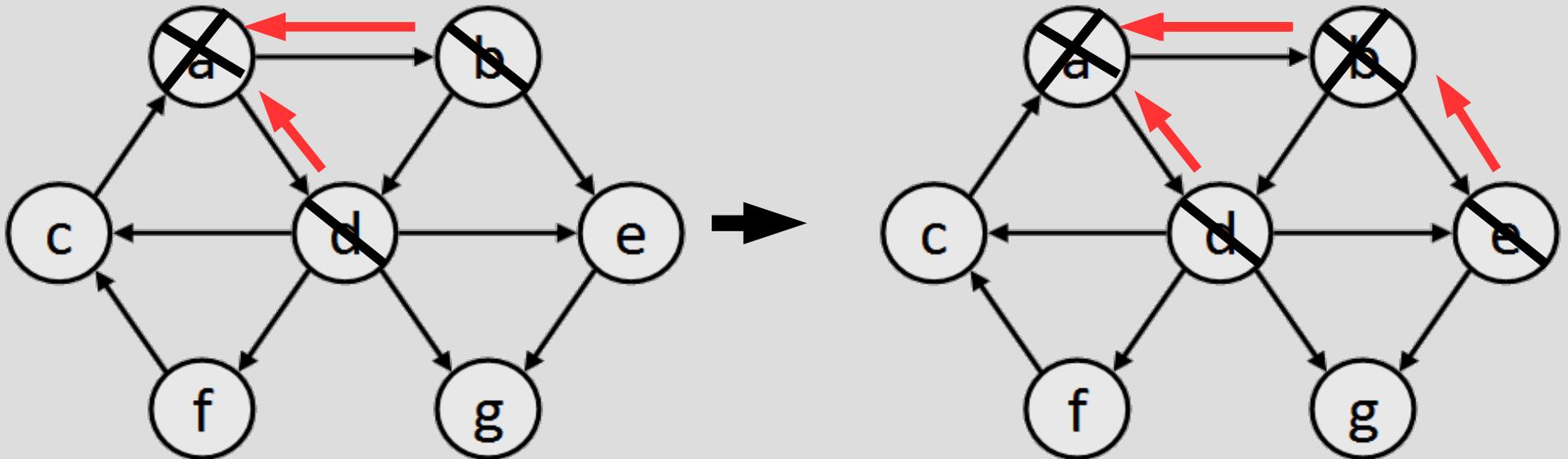
# BFS Overview

$Q = \{b, d\}$

Left-most = b

White neighbors = e

New  $Q = \{d, e\}$



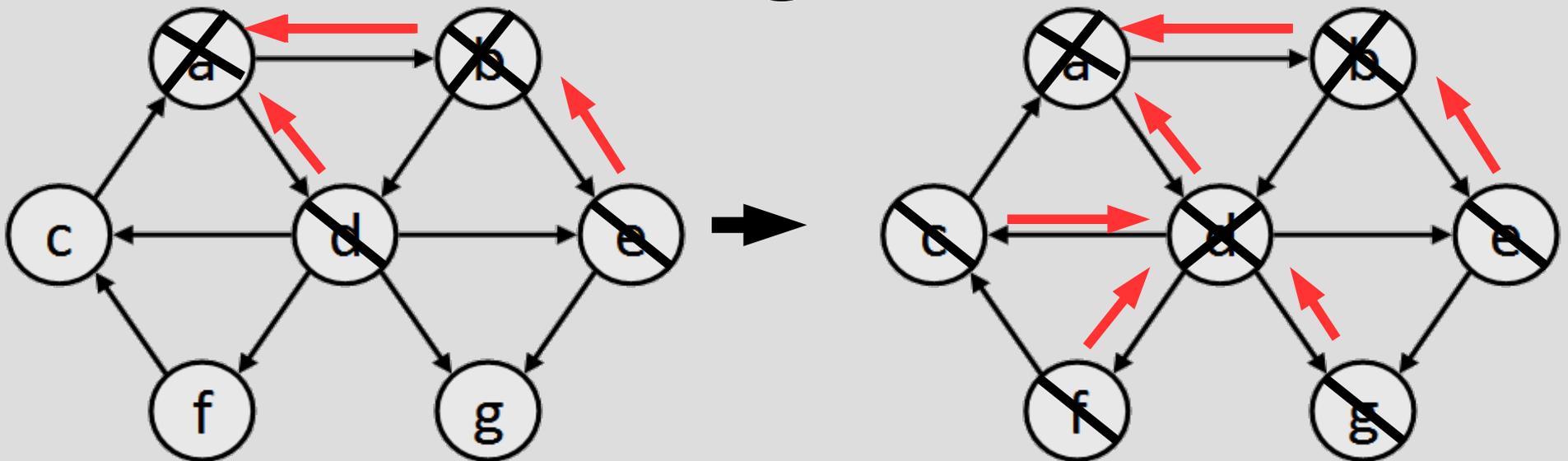
# BFS Overview

$Q = \{d, e\}$

Left-most = d

White neighbors = c & f & g

New  $Q = \{e, c, f, g\}$



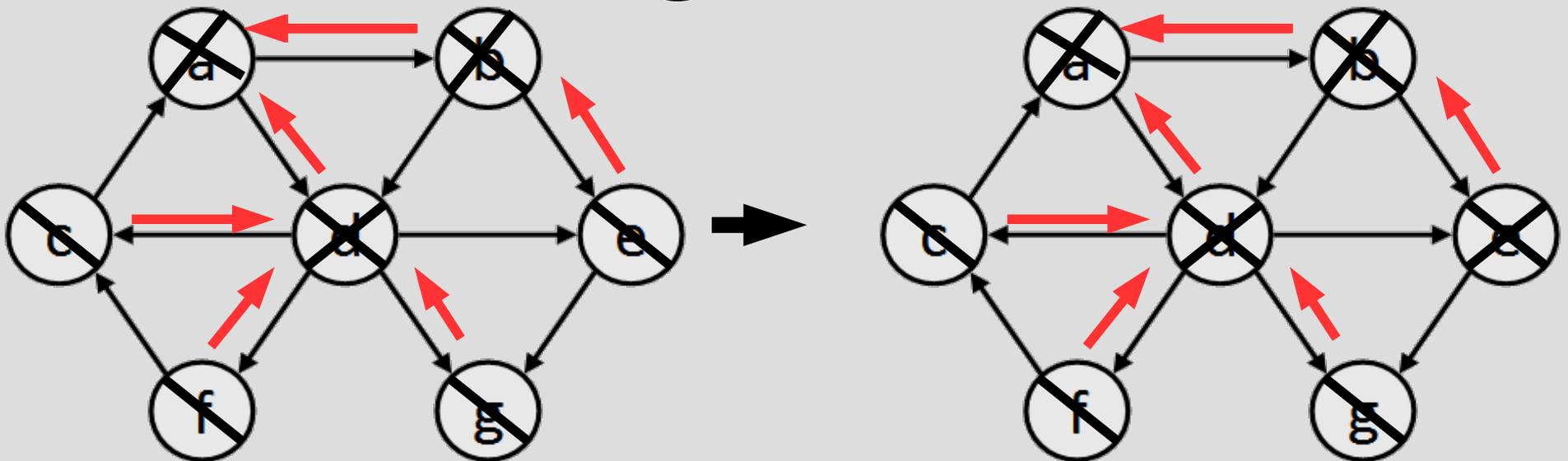
# BFS Overview

$Q = \{e, c, f, g\}$

Left-most = e

White neighbors = (none)

New  $Q = \{c, f, g\}$

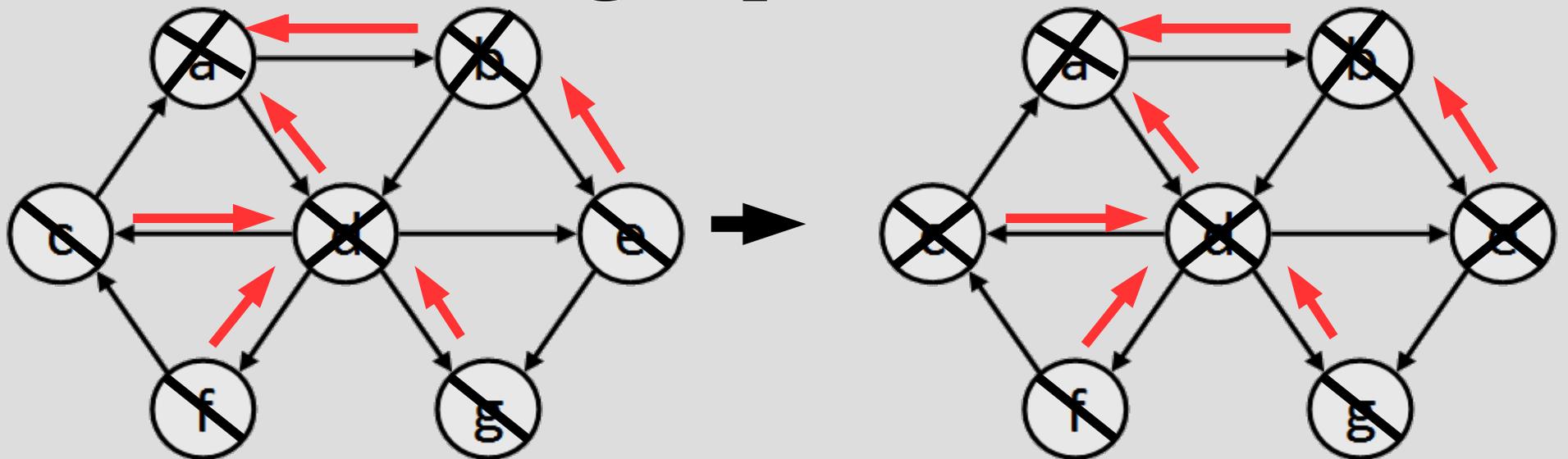


# BFS Overview

$Q = \{c, f, g\}$

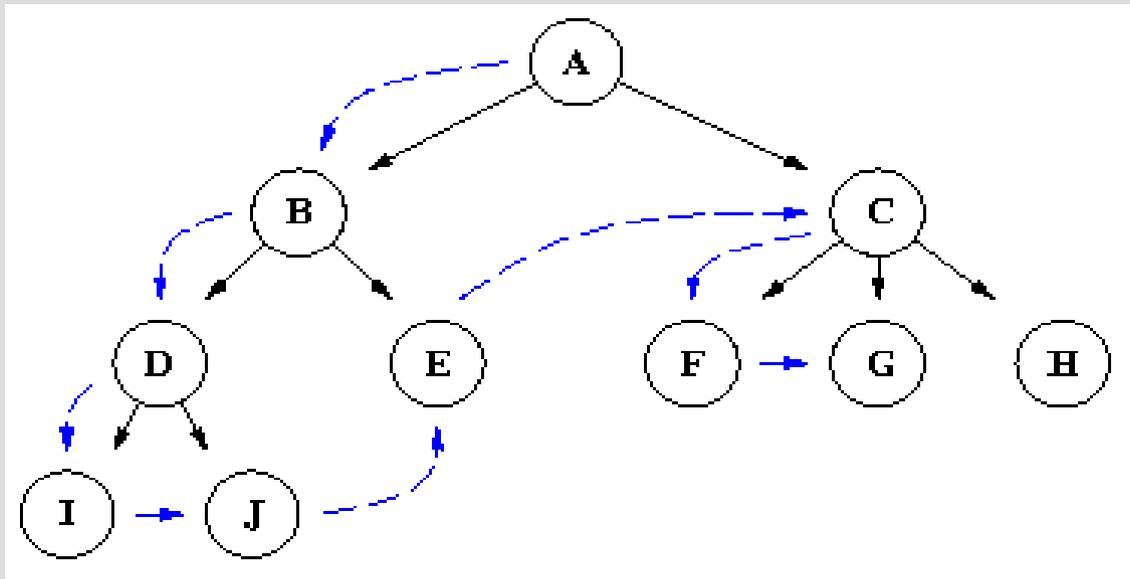
Left-most = c

Done! We found c, backtrack on red arrows to get path from “a”



# Depth First Search Overview

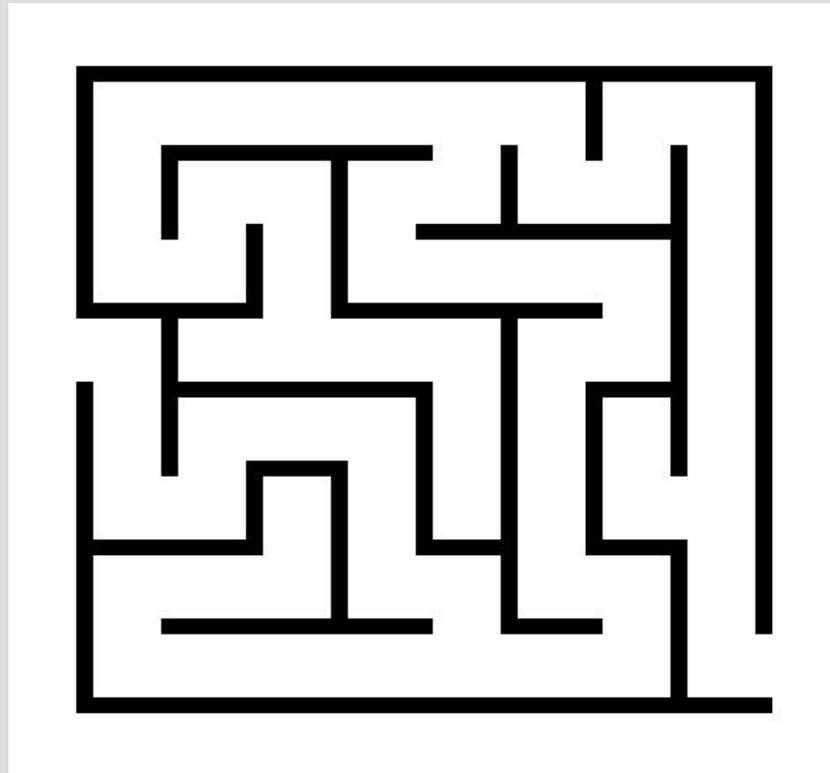
Create first-in-last-out (FILO) queue to explore unvisited nodes



# Depth First Search Overview

You can solve mazes by putting your left-hand on the wall and following it

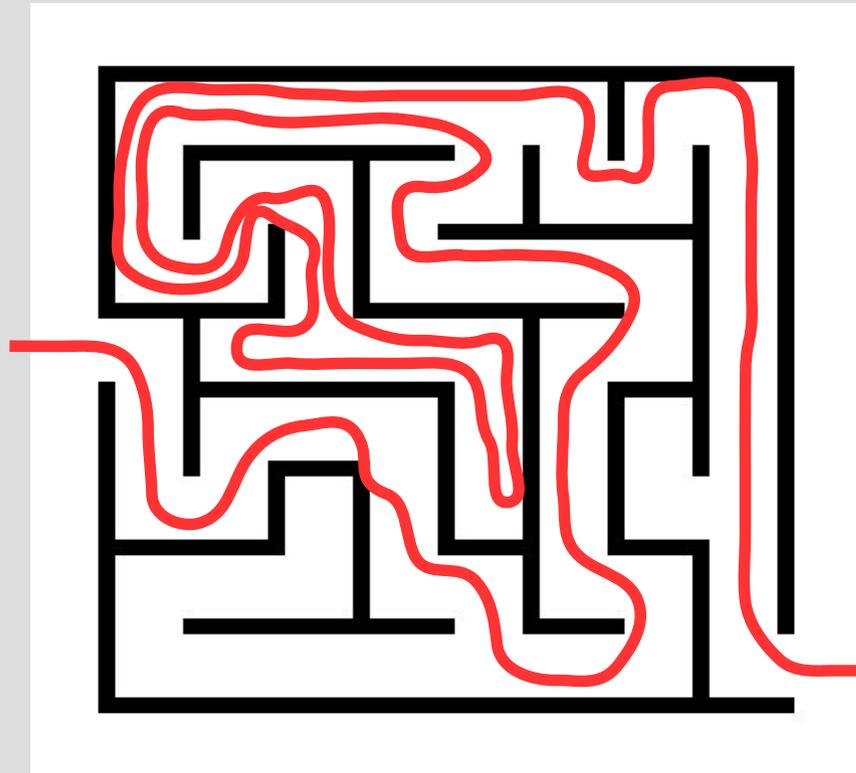
(i.e. left turns at every intersection)



# Depth First Search Overview

You can solve mazes by putting your left-hand on the wall and following it

(i.e. left turns at every intersection)





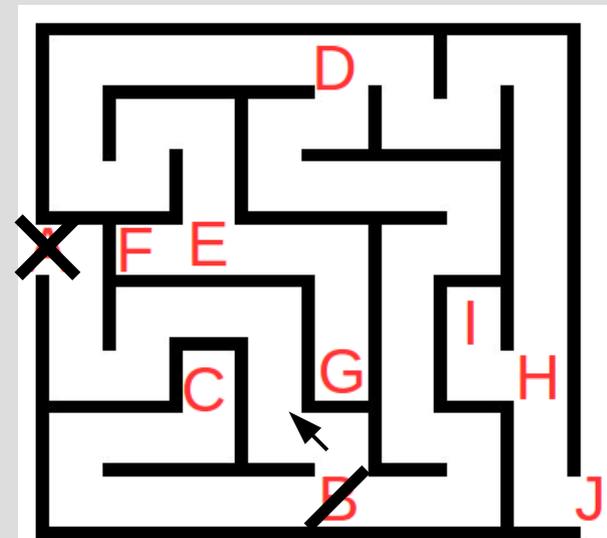
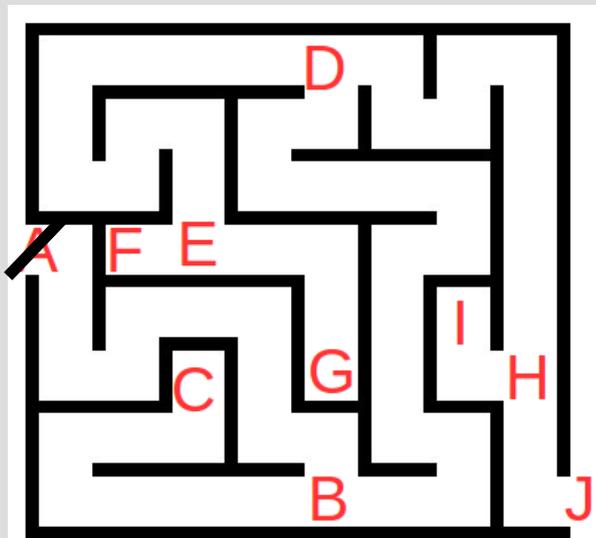
# Depth First Search Overview

$Q = \{A\}$

Right most = A

White neighbors =  $\{B\}$

New  $Q = \{B\}$



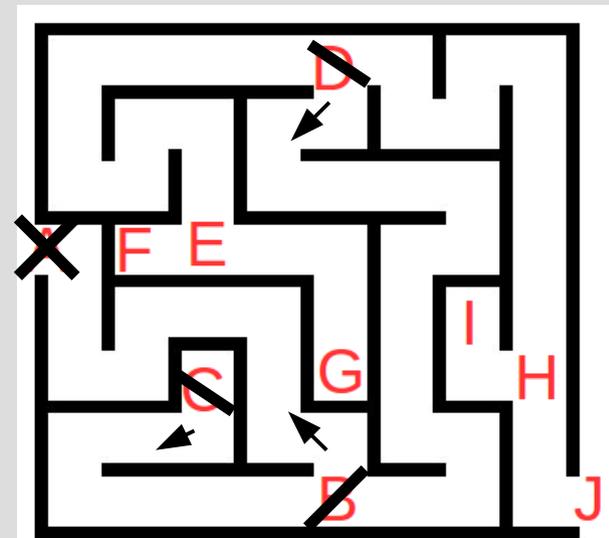
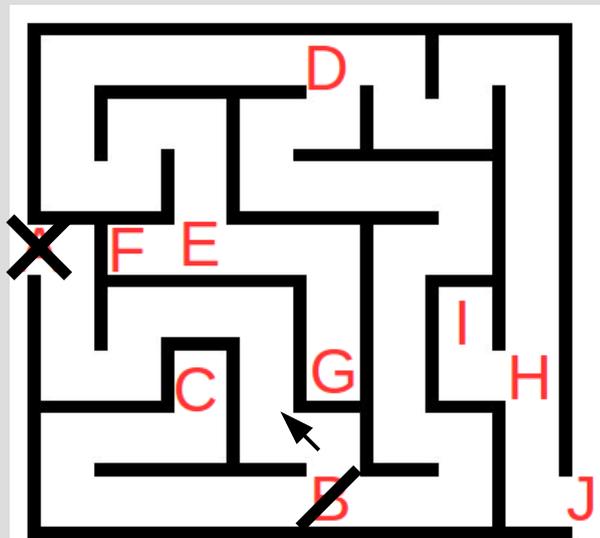
# Depth First Search Overview

$Q = \{B\}$

Right most = B

White neighbors =  $\{C, D\}$

New  $Q = \{C, D\}$



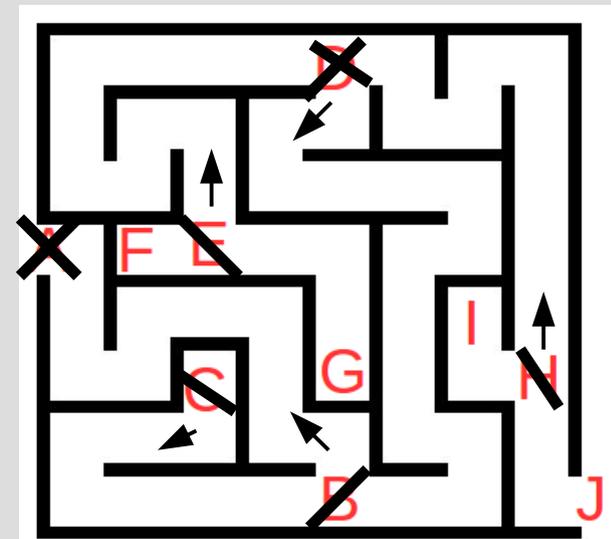
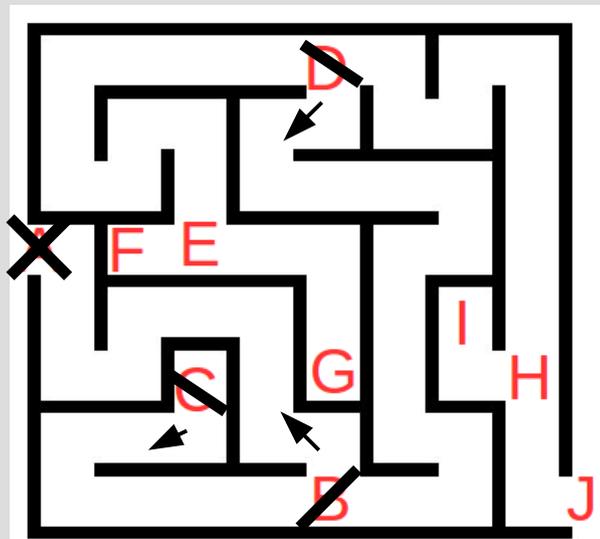
# Depth First Search Overview

$Q = \{C, D\}$

Right most = D

White neighbors =  $\{H, E\}$

New  $Q = \{C, H, E\}$



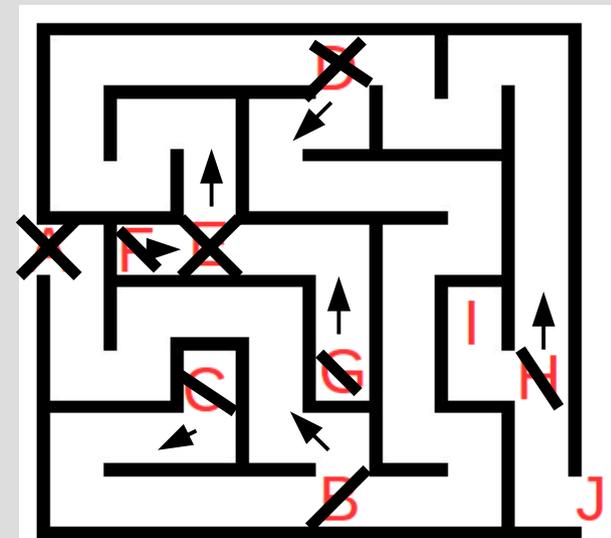
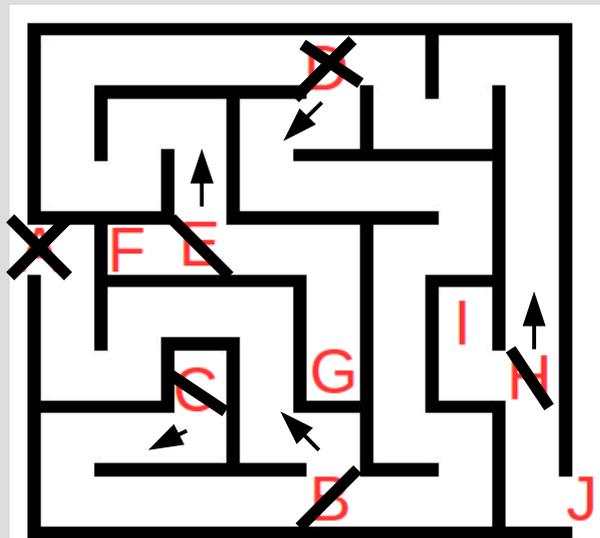
# Depth First Search Overview

$Q = \{C, H, E\}$

Right most = E

White neighbors =  $\{F, G\}$

New  $Q = \{C, H, F, G\}$







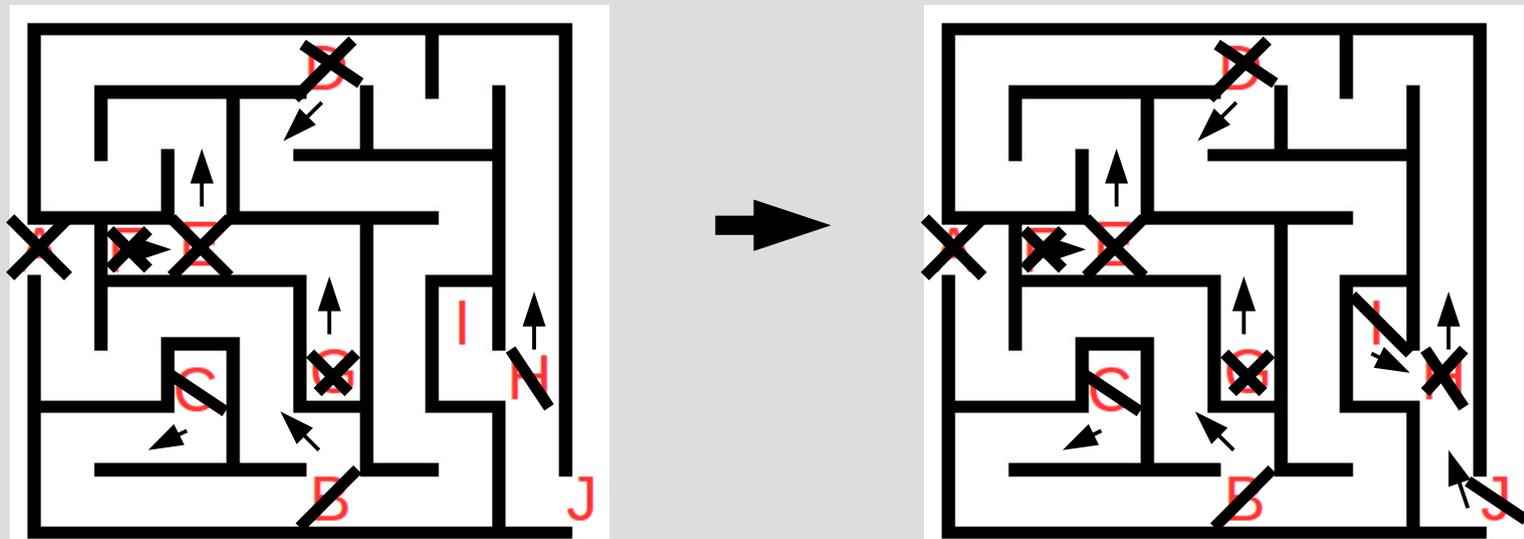
# Depth First Search Overview

$Q = \{C, H\}$

Right most = H

White neighbors =  $\{I, J\}$

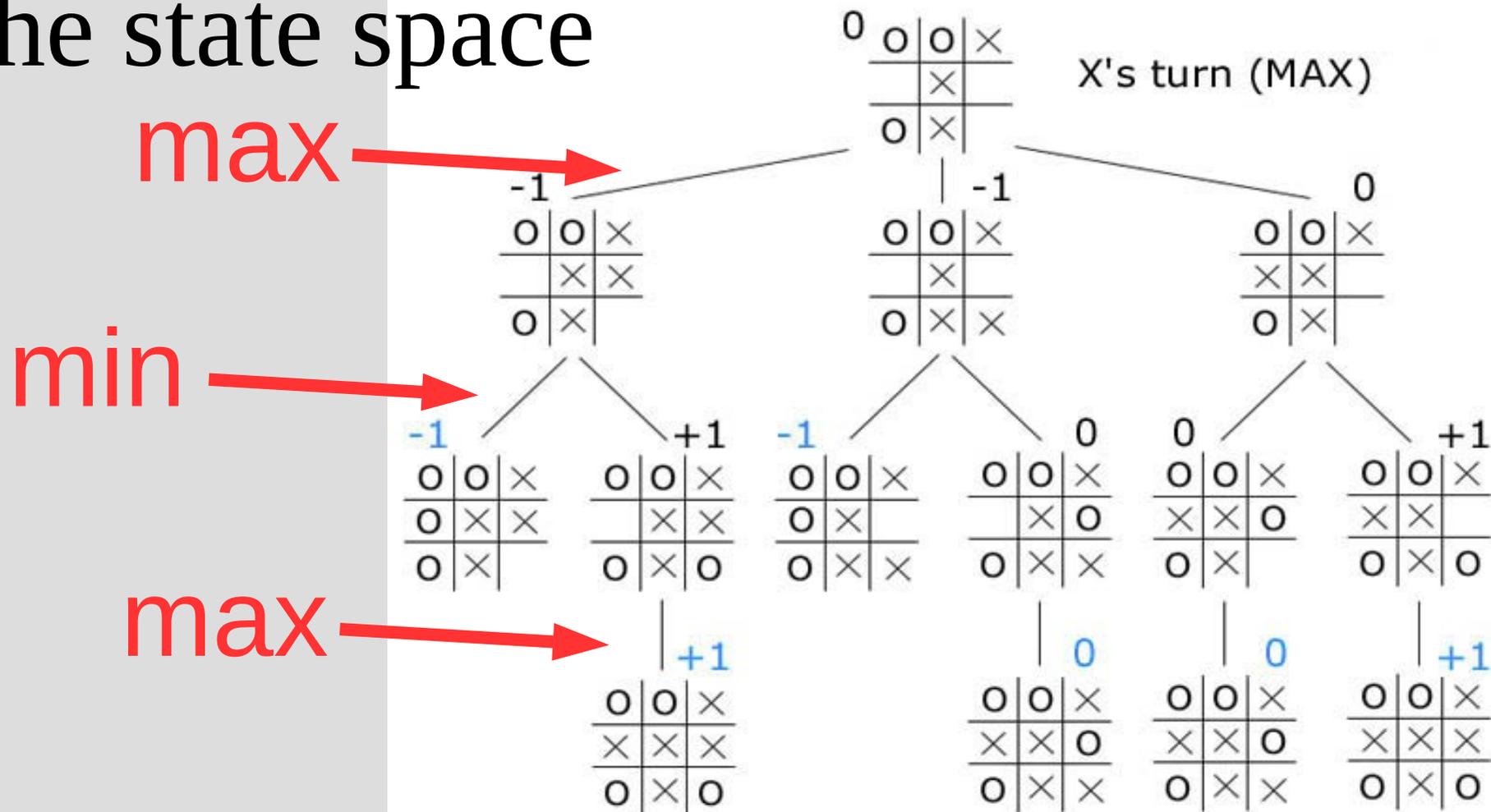
New  $Q = \{C, I, J\}$





# BFS and DFS in trees

Solve problems by making a tree of the state space



# BFS and DFS in trees

Often times, fully exploring the state space is too costly (takes forever)

Chess:  $10^{47}$  states (tree about  $10^{123}$ )

Go:  $10^{171}$  states (tree about  $10^{360}$ )

At 1 million states per second...

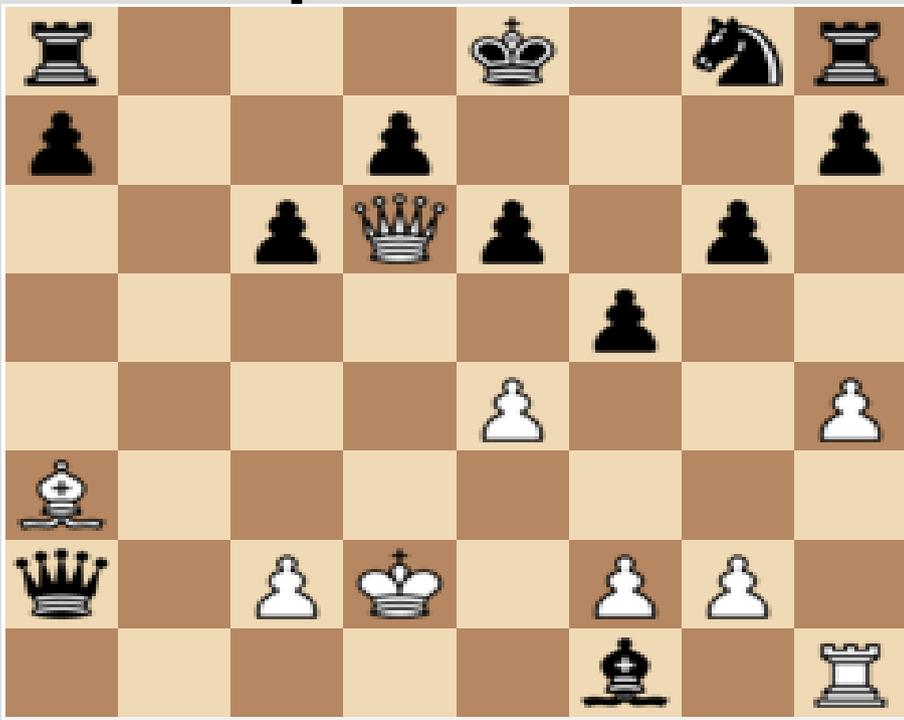
Chess:  $10^{109}$  years (past heat death

Go:  $10^{346}$  years of universe)

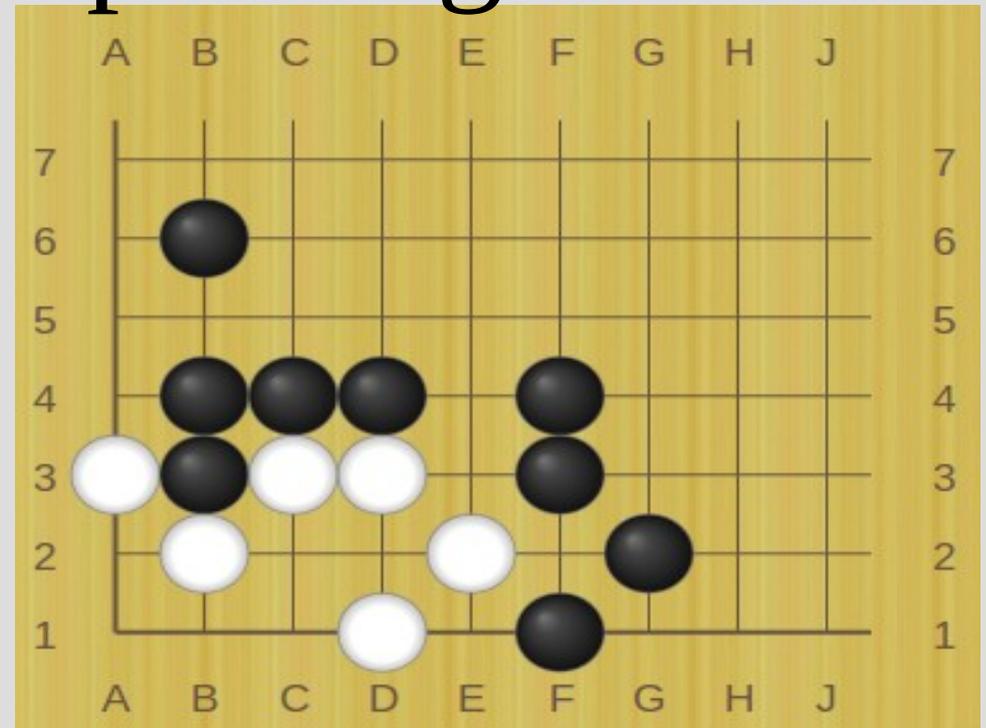
# BFS and DFS in trees

BFS prioritizes “exploring”

DFS prioritizes “exploiting”



White to move



Black to move

# BFS and DFS in trees

BFS benefits?

DFS benefits?

# BFS and DFS in trees

BFS benefits?

- if stopped before full search, can evaluate best found

DFS benefits?

- uses less memory on complete search

# BFS and DFS in graphs

BFS: shortest path from origin to any node

DFS: find graph structure

Both running time of  $O(V+E)$

# Breadth first search

BFS( $G, s$ ) // to find shortest path from  $s$

for all  $v$  in  $V$

$v.color = \text{white}$ ,  $v.d = \infty$ ,  $v.\pi = \text{NIL}$

$s.color = \text{grey}$ ,  $s.d = 0$

Enqueue( $Q, s$ )

while( $Q$  not empty)

$u = \text{Dequeue}(Q, s)$

    for  $v$  in  $G.adj[u]$

        if  $v.color == \text{white}$

$v.color = \text{grey}$ ,  $v.d = u.d + 1$ ,  $v.\pi = u$

            Enqueue( $Q, v$ )

$u.color = \text{black}$

# Breadth first search

Let  $\delta(s, v)$  be the shortest path from  $s$  to  $v$

After running BFS you can find this path as:  $v.\pi$  to  $(v.\pi).\pi$  to ...  $s$

(pseudo code on p. 601, recursion)

# BFS correctness

Proof: contradiction

Assume  $\delta(s, v) \neq v.d$

$v.d \geq \delta(s, v)$  (Lemma 22.2, induction)

Thus  $v.d > \delta(s, v)$

Let  $u$  be previous node on  $\delta(s, v)$

Thus  $\delta(s, v) = \delta(s, u) + 1$

and  $\delta(s, u) = u.d$

Then  $v.d > \delta(s, v) = \delta(s, u) + 1 = u.d + 1$

# BFS correctness

$$v.d > \delta(s,v) = \delta(s,u)+1 = u.d+1$$

Cases on color of  $v$  when  $u$  dequeue,  
all cases invalidate top equation

Case white: alg sets  $v.d = u.d + 1$

Case black: already removed

thus  $v.d \leq u.d$  (corollary 22.4)

Case grey: exists  $w$  that dequeued  $v$ ,  
 $v.d = w.d+1 \leq u.d+1$  (corollary 22.4)

# Depth first search

DFS(G)

for all  $v$  in  $V$

$v.color = white$ ,  $v.\pi = NIL$

time=0

for each  $v$  in  $V$

    if  $v.color == white$

        DFS-Visit(G,  $v$ )

# Depth first search

DFS-Visit( $G, u$ )

time=time+1

$u.d = \text{time}$ ,  $u.\text{color} = \text{grey}$

for each  $v$  in  $G.\text{adj}[u]$

if  $v.\text{color} == \text{white}$

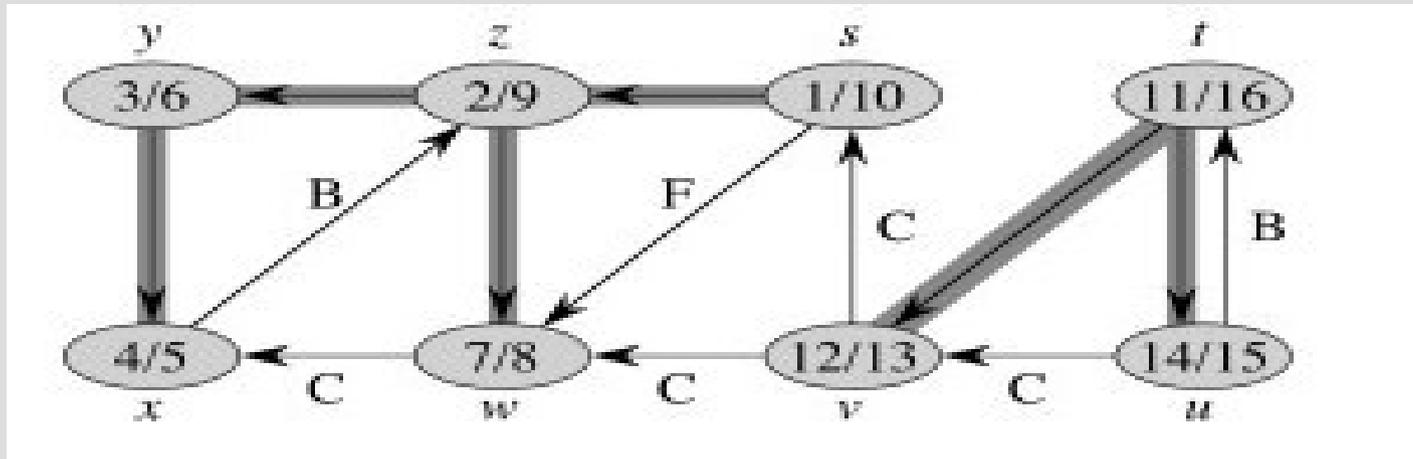
$v.\pi = u$

DFS-Visit( $G, v$ )

$u.\text{color} = \text{black}$ ,  $\text{time} = \text{time} + 1$ ,  $u.f = \text{time}$

# Depth first search

Edge markers:



Consider edge  $u$  to  $v$

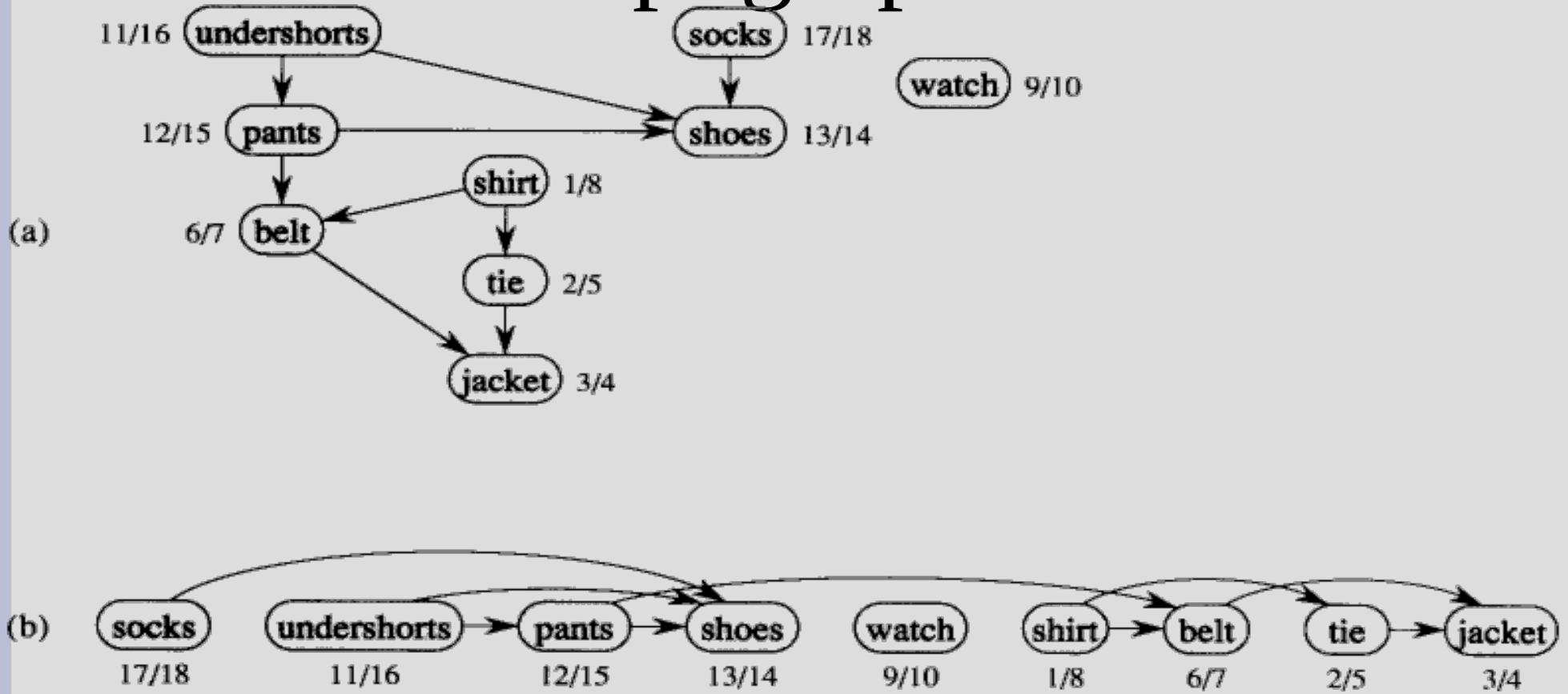
C = Edge to black node ( $u.d > v.f$ )

B = Edge to grey node ( $u.f < v.f$ )

F = Edge to black node ( $u.f > v.f$ )

# Depth first search

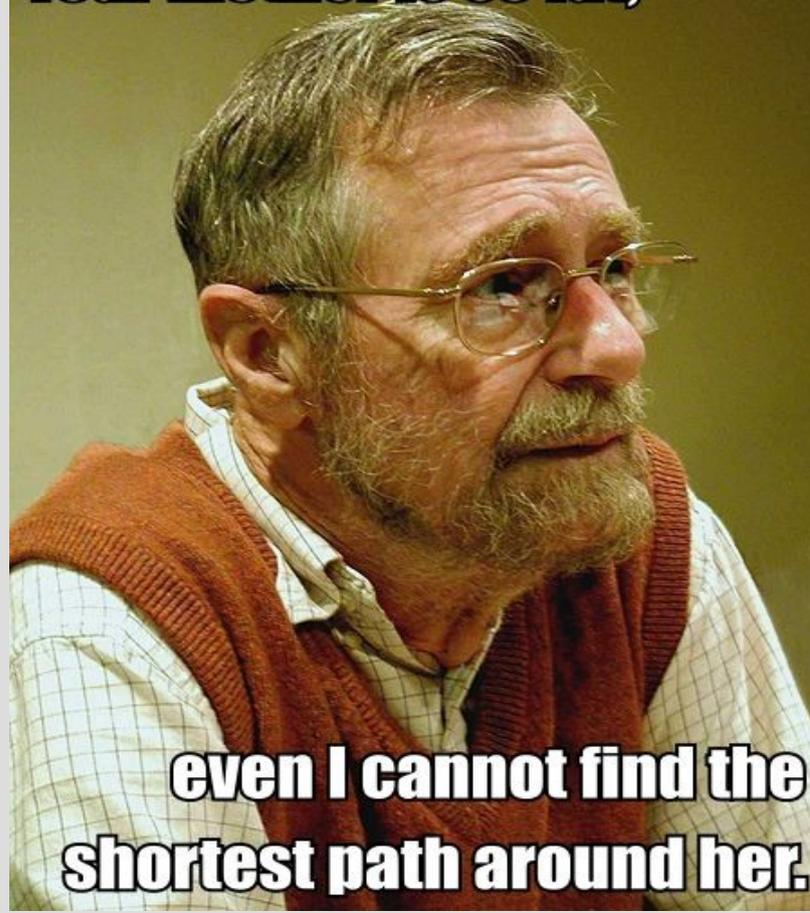
DFS can do topographical sort



Run DFS, sort in decreasing finish time

# Weighted graphs

**Your mother is so fat,**



**even I cannot find the  
shortest path around her.**

# Weighted graph

Edges in weighted graph are assigned a weight:  $w(v_1, v_2)$ ,  $v_1, v_2$  in  $V$

If path  $p = \langle v_0, v_1, \dots, v_k \rangle$  then the weight is:  $w(p) = \sum_{i=0}^k w(v_{i-1}, v_i)$

Shortest Path:

$\delta(u, v) = \min\{w(p) : v_0 = u, v_k = v\}$

# Shortest paths

Today we will look at single-source shortest paths

This finds the shortest path from some starting vertex,  $s$ , to any other vertex on the graph (if it exists)

This creates  $G_\pi$ , the shortest path tree

# Shortest paths

Optimal substructure: Let  $\delta(v_0, v_k) = p$ ,  
then for all  $0 \leq i \leq j \leq k$ ,  $\delta(v_i, v_j) = p_{i,j} =$   
 $\langle v_i, v_{i+1}, \dots, v_j \rangle$

Proof?

Where have we seen this before?

# Shortest paths

Optimal substructure: Let  $\delta(v_0, v_k) = p$ , then for all  $0 \leq i \leq j \leq k$ ,  $\delta(v_i, v_j) = p_{i,j} = \langle v_i, v_{i+1}, \dots, v_j \rangle$

Proof? Contradiction!

Suppose  $w(p'_{i,j}) < p_{i,j}$ , then let

$p'_{0,k} = p_{0,i} p'_{i,j} p_{j,k}$  then  $w(p'_{0,k}) < w(p)$

# Relaxation

We will only do relaxation on the values  $v.d$  (min weight) for vertex  $v$

Relax( $u, v, w$ )

if( $v.d > u.d + w(u, v)$ )

$v.d = u.d + w(u, v)$

$v.\pi = u$

# Relaxation

We will assume all vertices start with  $v.d = \infty, v.\pi = \text{NIL}$  except  $s, s.d = 0$

This will take  $O(|V|)$  time

This will not effect the asymptotic runtime as it will be at least  $O(|V|)$  to find single-source shortest path

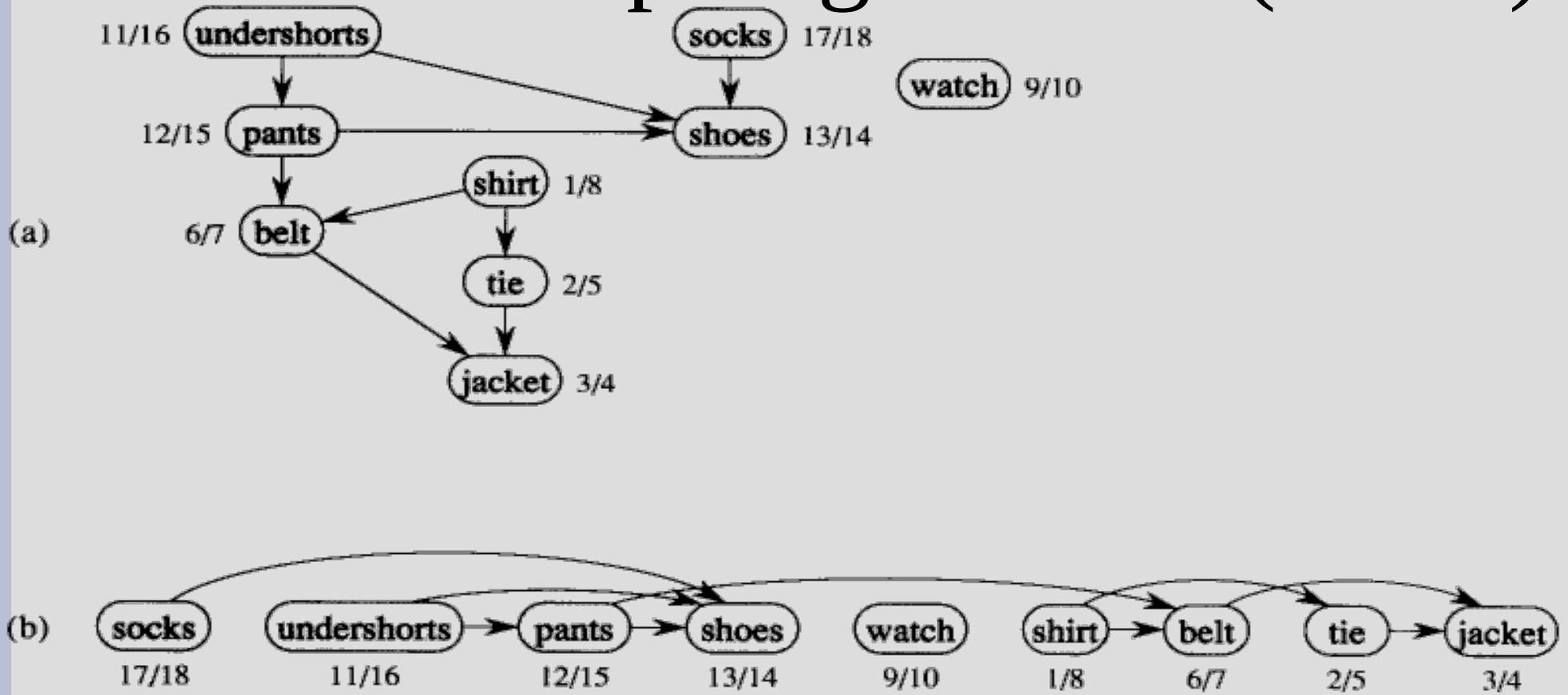
# Relaxation

Relaxation properties:

1.  $\delta(s,v) \leq \delta(s,u) + \delta(u,v)$  (triangle inequality)
2.  $v.d \geq \delta(s,v)$ ,  $v.d$  is monotonically decreasing
3. if no path,  $v.d = \delta(s,v) = \infty$
4. if  $\delta(s,v)$ , when  $(v.\pi).d = \delta(s,v.\pi)$  then  $\text{relax}(v.\pi, v, w)$  causes  $v.d = \delta(s,v)$
5. if  $\delta(v_0, v_k) = p_{0,k}$ , then when relaxed in order  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$  then  $v_{k.d} = \delta(v_0, v_k)$  even if other relax happen
6. when  $v.d = \delta(s,v)$  for all  $v$  in  $V$ ,  $G_\pi$  is shortest path tree rooted at  $s$

# Directed Acyclic Graphs

DFS can do topological sort (DAG)



Run DFS, sort in decreasing finish time

# Directed Acyclic Graphs

DAG-shortest-paths( $G, w, s$ )

topologically sort  $G$

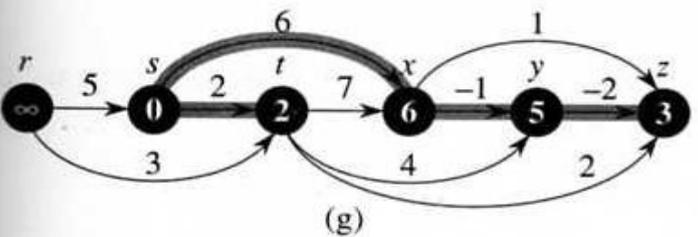
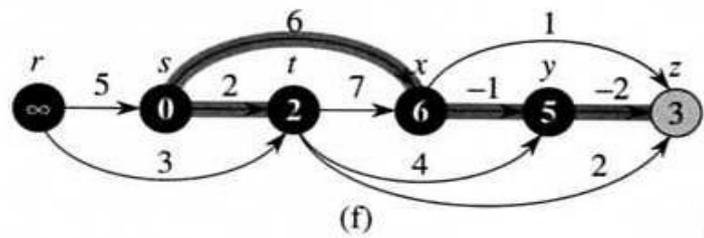
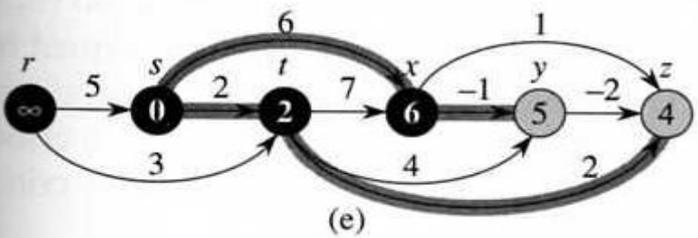
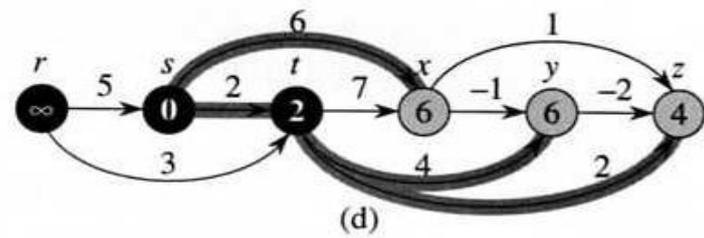
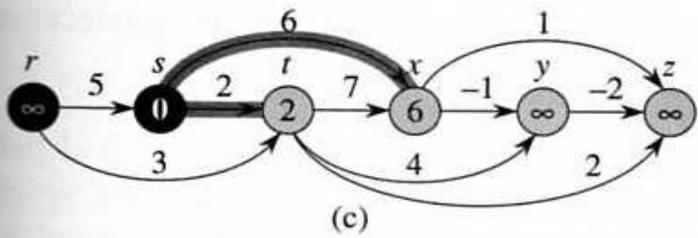
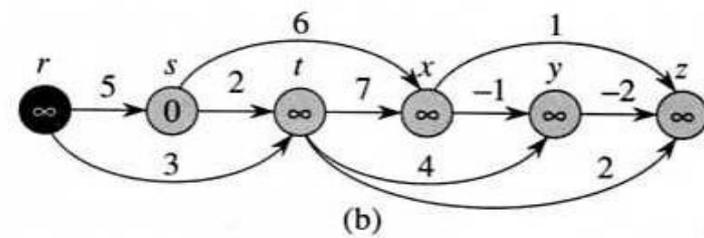
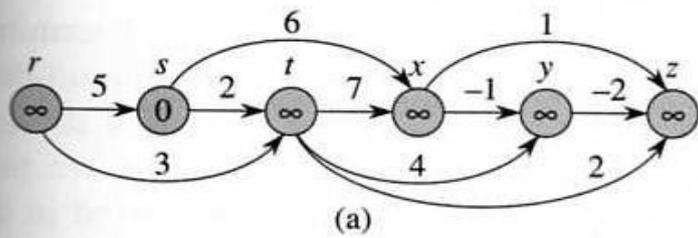
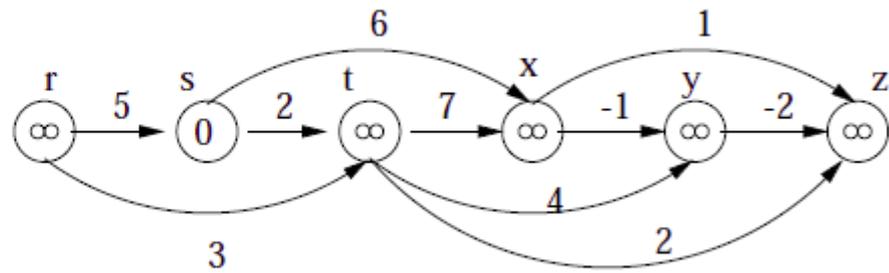
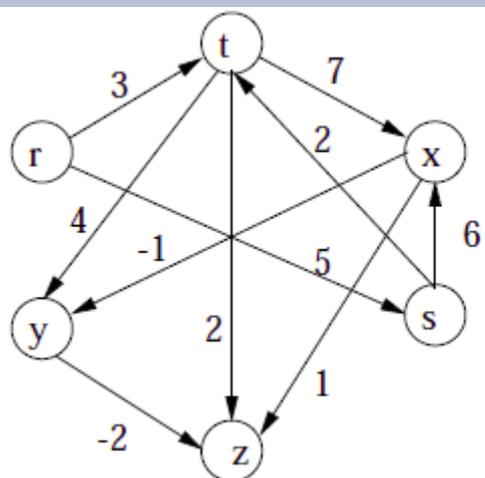
initialize graph from  $s$

for each  $u$  in  $V$  in topological order

    for each  $v$  in  $G.Adj[u]$

        Relax( $u, v, w$ )

Runtime:  $O(|V| + |E|)$



# Directed Acyclic Graphs

Correctness:

Prove it!

# Directed Acyclic Graphs

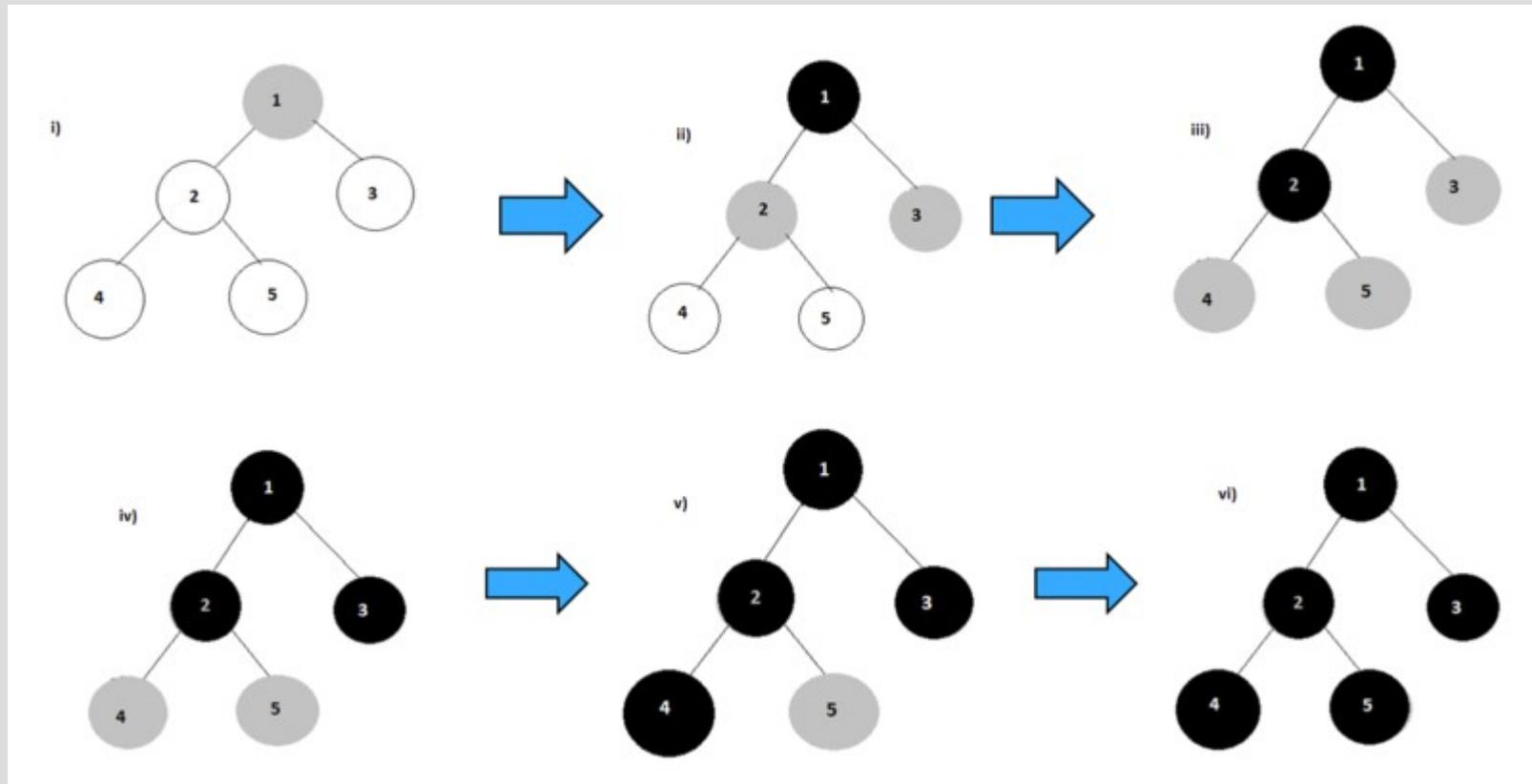
Correctness:

By definition of topological order,  
When relaxing vertex  $v$ , we have  
already relaxed any preceding  
vertices

So by relaxation property 5, we have  
found the shortest path to all  $v$

# BFS (unweighted graphs)

Create FIFO queue to explore unvisited nodes



# Dijkstra

Dijkstra's algorithm is the BFS equivalent for non-negative weight graphs



# Dijkstra

Dijkstra( $G, w, s$ )

initialize  $G$  from  $s$

$Q = G.V$ ,  $S = \text{empty}$

while  $Q$  not empty

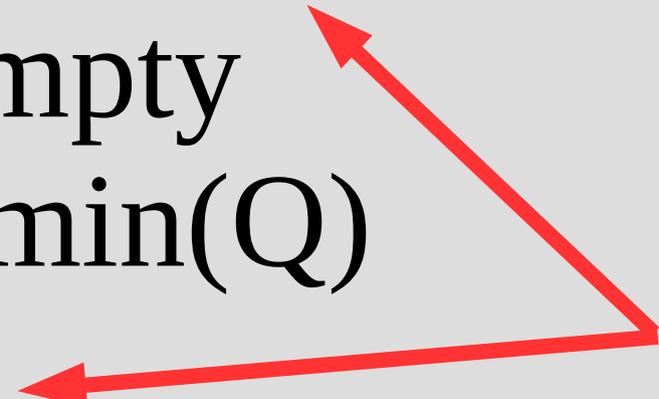
$u = \text{Extract-min}(Q)$

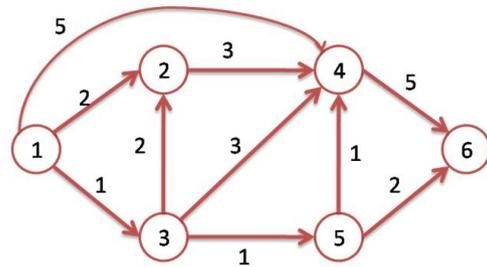
$S = S \cup \{u\}$

for each  $v$  in  $G.\text{Adj}[u]$

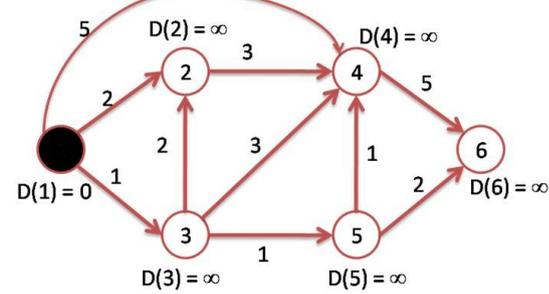
relax( $u, v, w$ )

$S$  optional

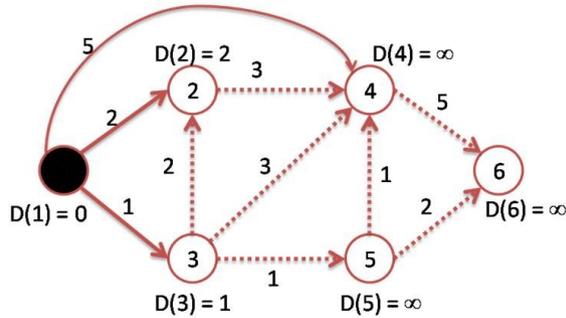




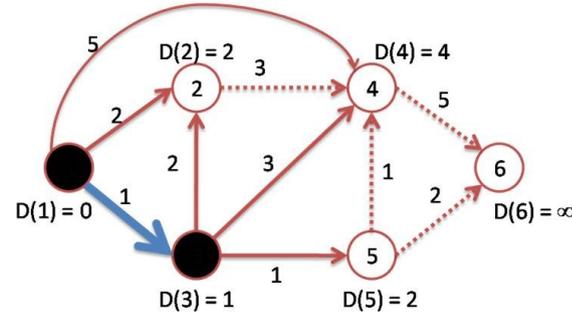
(a) Network Model



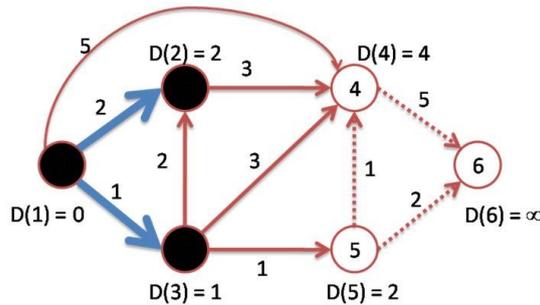
(b) Distance initialized



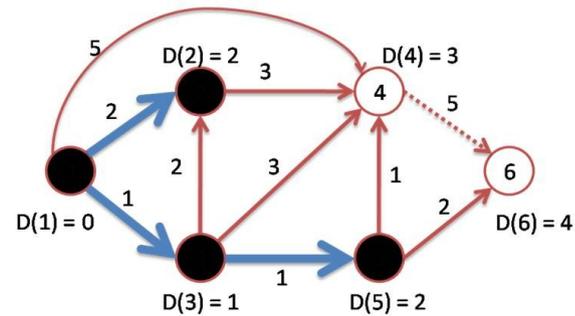
(c) Distance to adjacent nodes updated



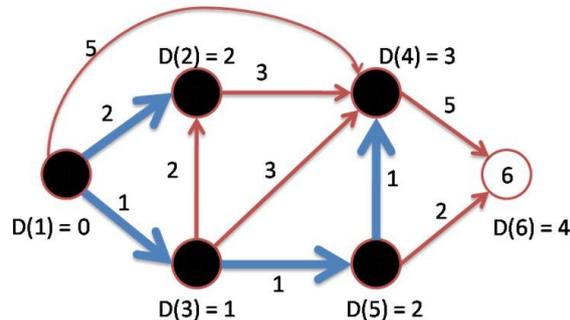
(d) Node 3 selected



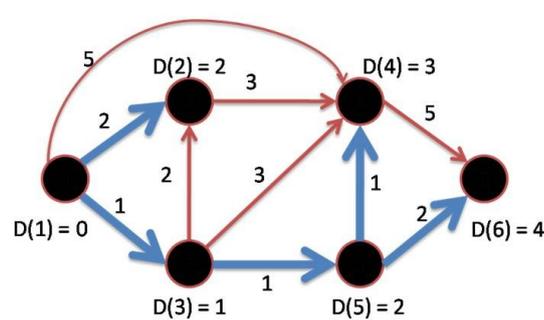
(e) Node 2 selected



(f) Node 5 selected



(g) Node 4 selected



(h) Shortest path found

# Dijkstra

Runtime?

# Dijkstra

Runtime:

Extract-min() run  $|V|$  times

Relax runs Decrease-key()  $|E|$  times

Both take  $O(\lg n)$  time

So  $O((|V| + |E|) \lg |V|)$  time

(can get to  $O(|V| \lg |V| + E)$  using  
Fibonacci heaps)

# Dijkstra

Runtime note:

If  $G$  is almost fully connected,

$$|E| \approx |V|^2$$

Use a simple array to store v.d

$$\text{Extract-min}() = O(|V|)$$

$$\text{Decrease-key}() = O(1)$$

$$\text{total: } O(|V|^2 + E)$$

# Dijkstra

Correctness: (p.660)

Sufficient to prove when  $u$  added to  $S$ ,  $u.d = \delta(s,u)$

Base:  $s$  added to  $S$  first,  $s.d=0=\delta(s,s)$

Termination: Loop ends after  $Q$  is empty, so  $V=S$  and we done

# Dijkstra

Step: Assume  $v$  in  $S$  has  $v.d = \delta(s, v)$

Let  $y$  be the first vertex outside  $S$   
on path of  $\delta(s, u)$

We know by relaxation property 4,  
that  $\delta(s, y) = y.d$  (optimal sub-structure)

$y.d = \delta(s, y) \leq \delta(s, u) \leq u.d$ , as  $w(p) \geq 0$

# Dijkstra

Step: Assume  $v$  in  $S$  has  $v.d = \delta(s, v)$

But as  $u$  was picked before  $y$ ,

$u.d \leq y.d$ , combined with  $y.d \leq u.d$

$y.d = u.d$

Thus  $y.d = \delta(s, y) = \delta(s, u) = u.d$