

# NSUPS Bootcamp Week 7

Graph Theories

<http://bit.ly/bootcamp07>

# Graphs

A graph  $G = (V, E)$

$V$  = set of vertices,  $E$  = set of edges

*Dense* graph:  $|E| \approx |V|^2$ ; *Sparse* graph:  $|E| \approx |V|$

*Undirected graph*:

edge  $(u,v) = \text{edge } (v,u)$

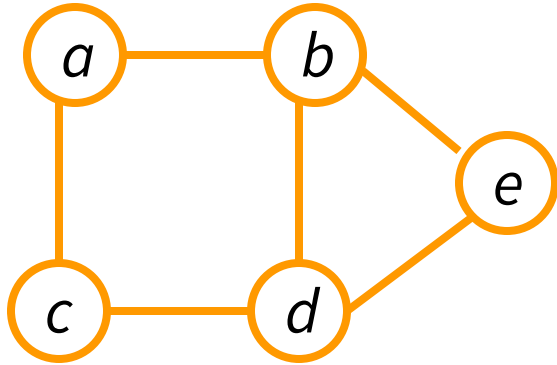
No self-loops

*Directed graph*:

Edge  $(u,v)$  goes from vertex  $u$  to vertex  $v$ , notated  $u \rightarrow v$

A *weighted graph* associates weights with either the edges or the vertices

# Adjacency Matrix Representation



	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>
<i>a</i>	0	1	1	0	0
<i>b</i>	1	0	0	1	1
<i>c</i>	1	0	0	1	0
<i>d</i>	0	1	1	0	1
<i>e</i>	0	1	0	1	0

Check for an edge in constant  
time

# Adjacency Matrix Representation

Memory required

$$O(V+V^2)=O(V^2)$$

Preferred when

The graph is **dense**:  $E = O(V^2)$

Advantage

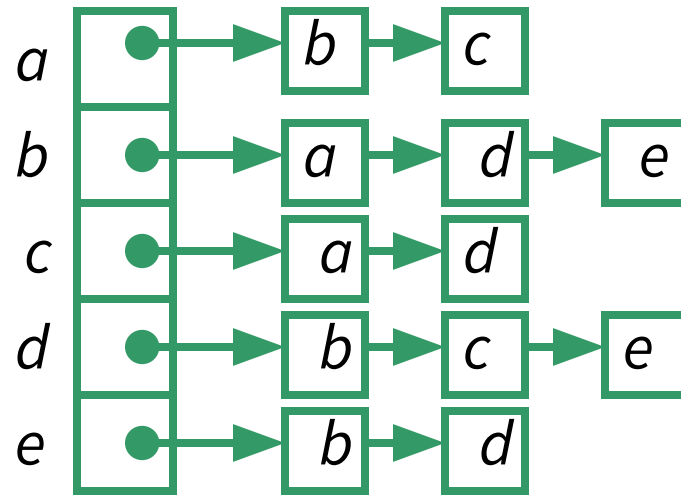
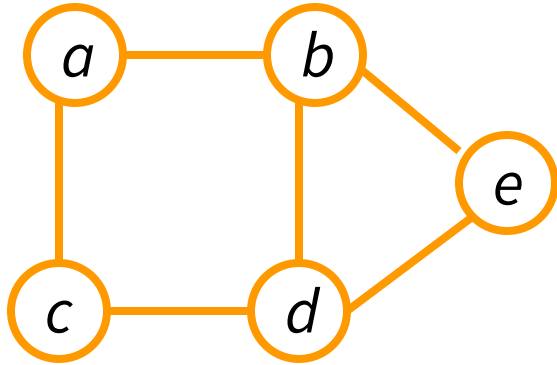
Can quickly determine if there is an edge between two vertices

Disadvantage

No quick way to determine the vertices adjacent **from** another vertex.

Listing all the neighbors of ANY vertex takes  $O(V)$  time.

# Adjacency List Representation



Space-efficient for sparse graphs

# Adjacency List Representation

Memory required

$$O(V + E)$$

$O(V)$  for sparse graphs since  $E=O(V)$

Preferred when

for **sparse** graphs:  $E = O(V)$

$O(V^2)$  for dense graphs since  $E=O(V^2)$

Disadvantage

No quick way to determine whether there is an edge between vertices  $u$  and  $v$

Advantage

Can quickly determine all the vertices adjacent **from** a given vertex,  $v$ : takes  $O(N(v))$  time, where  $N(v)$  is the number of neighbors of  $v$ .

# Graph Searching

Given: a graph  $G = (V, E)$ , directed or undirected

Goal: methodically explore every vertex and every edge

Ultimately: build a tree on the graph

- Pick a vertex as the root

- Choose certain edges to produce a tree

- Note: might also build a *forest* if graph is not connected

# Graph Traversals

Ways to traverse/search a graph

Visit every vertex exactly once

Breadth-First Search

Depth-First Search



# Breadth-First Search

“Explore” a graph, turning it into a tree

- One vertex at a time

- Expand frontier of explored vertices across the *breadth* of the frontier

Builds a tree over the graph

- Pick a *source vertex* to be the root

- Find (“discover”) all of its children, then their children, etc.

# Breadth-First Search

Associate vertex “colors” to guide the algorithm

- White vertices have not been discovered

  - All vertices start out white

- Grey vertices are discovered but not fully explored

  - They may be adjacent to white vertices

- Black vertices are discovered and fully explored

  - They are adjacent only to black and gray vertices

Explore vertices by scanning adjacency list of grey vertices

# BFS Trees

BFS tree is not necessarily unique for a given graph

Depends on the order in which neighboring vertices are processed

During the breadth-first search, assign an integer to each vertex

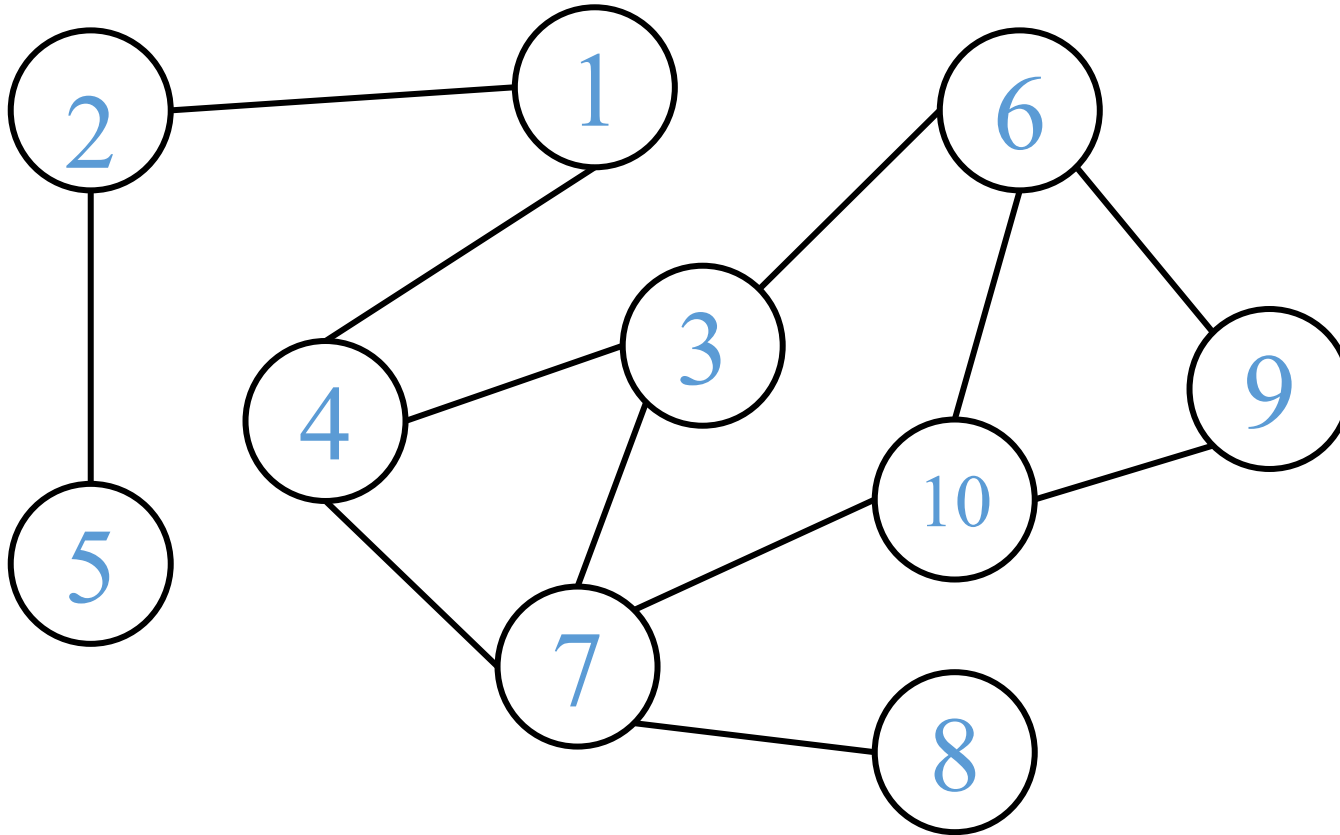
Indicate the distance of each vertex from the source  $s$

# Breadth-First Search

**BFS( $G, s$ )**

1.     **for** each vertex  $u \in G.V - \{s\}$
2.          $u.color = \text{WHITE}$
3.          $u.d = \infty$
4.          $u.\pi = \text{NIL}$
5.      $s.color = \text{GRAY}$
6.      $s.d = 0$
7.      $s.\pi = \text{NIL}$
8.      $Q = \emptyset$
9.      $\text{ENQUEUE}(Q, s)$
10.    **while**  $Q \neq \emptyset$
11.        $u = \text{DEQUEUE}(Q)$
12.       **for** each  $v \in G.Adj[u]$
13.          **if**  $v.color == \text{WHITE}$
14.              $v.color = \text{GRAY}$
15.              $v.d = u.d + 1$
16.              $v.\pi = u$
17.              $\text{ENQUEUE}(Q, v)$
18.        $u.color = \text{BLACK}$

## Breadth-First Search: Example



visited[]

[1] = 0

[2] = 0

[3] = 0

[4] = 0

[5] = 0

[6] = 0

[7] = 0

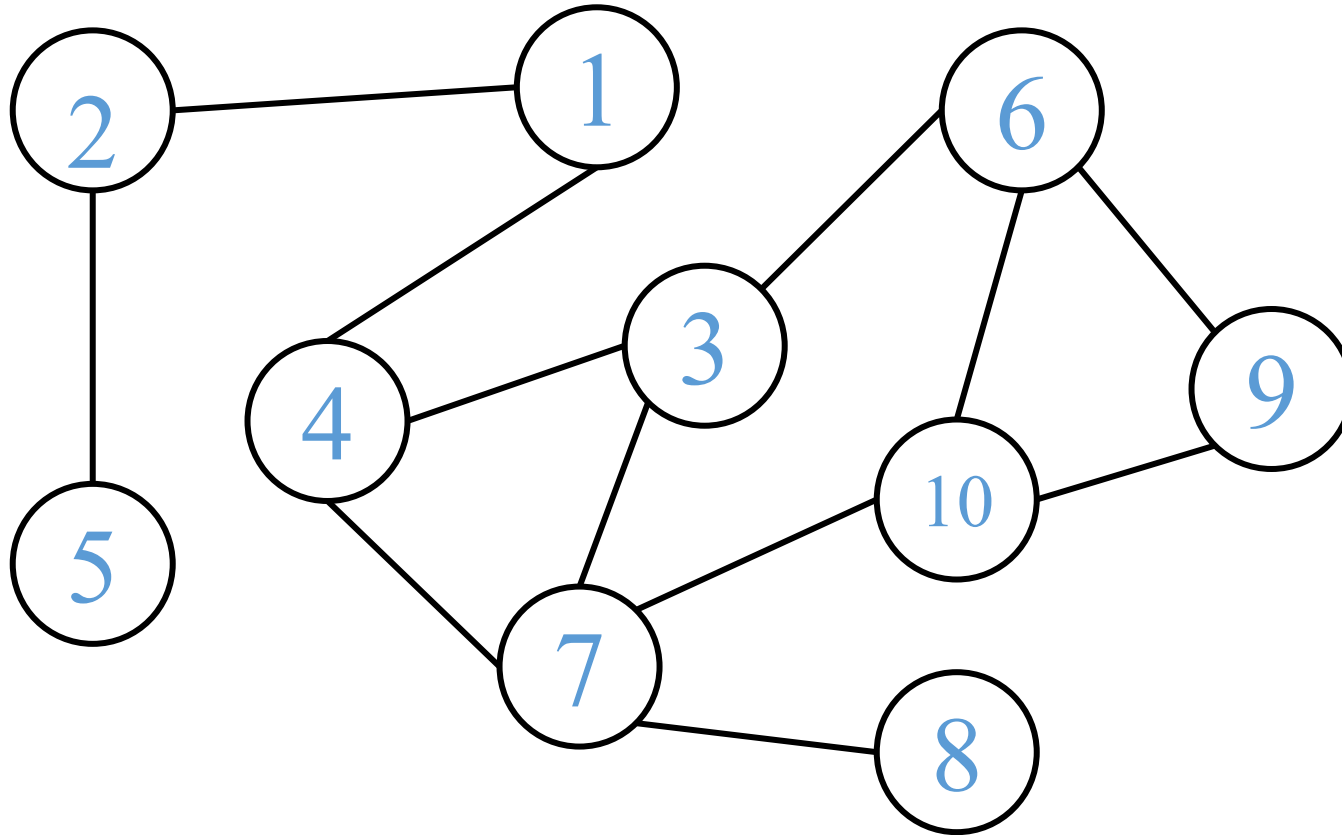
[8] = 0

[9] = 0

[10] = 0

***Q:***

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 0

[3] = 0

[4] = 0

[5] = 0

[6] = 0

[7] = 0

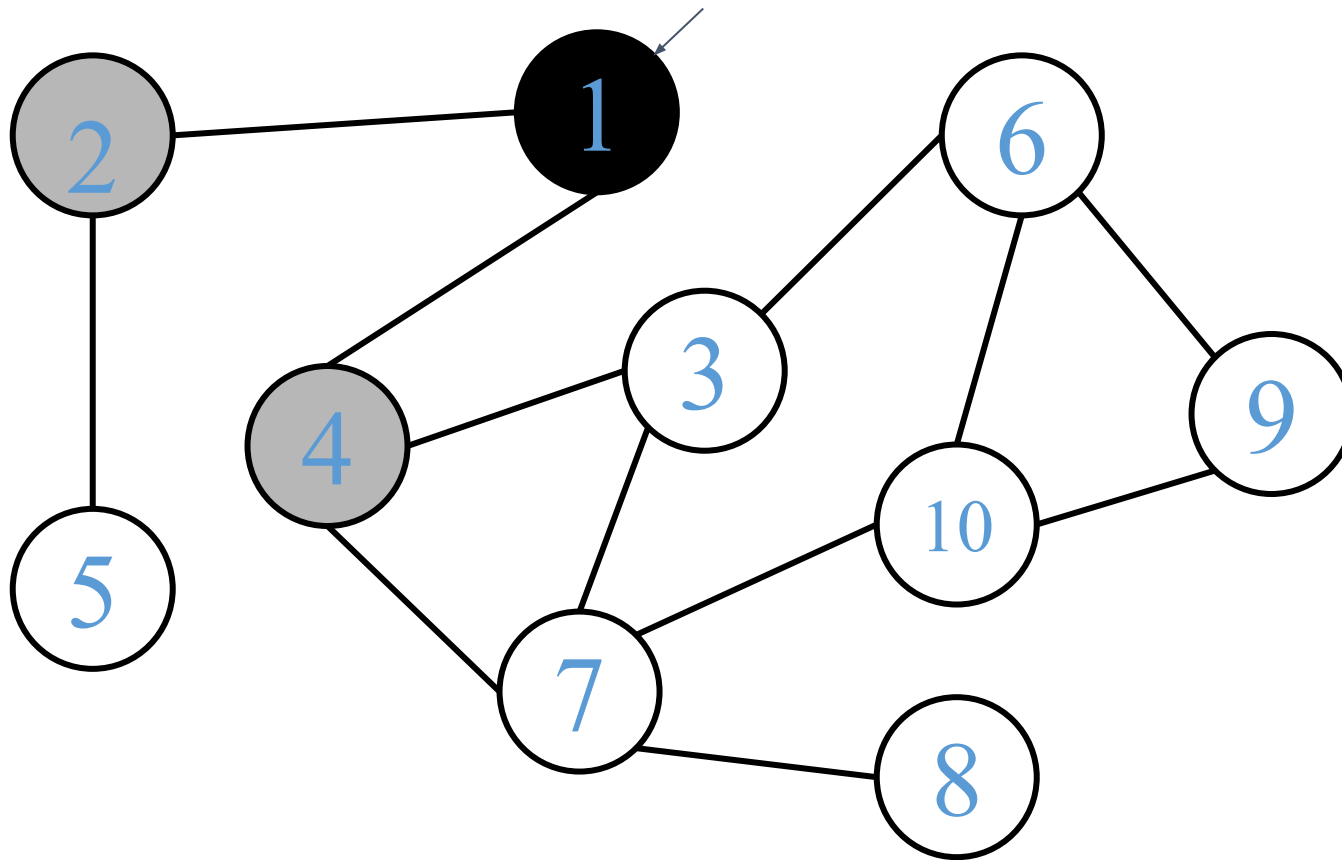
[8] = 0

[9] = 0

[10] = 0

***Q:*** *1*

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 0

[3] = 0

[4] = 0

[5] = 0

[6] = 0

[7] = 0

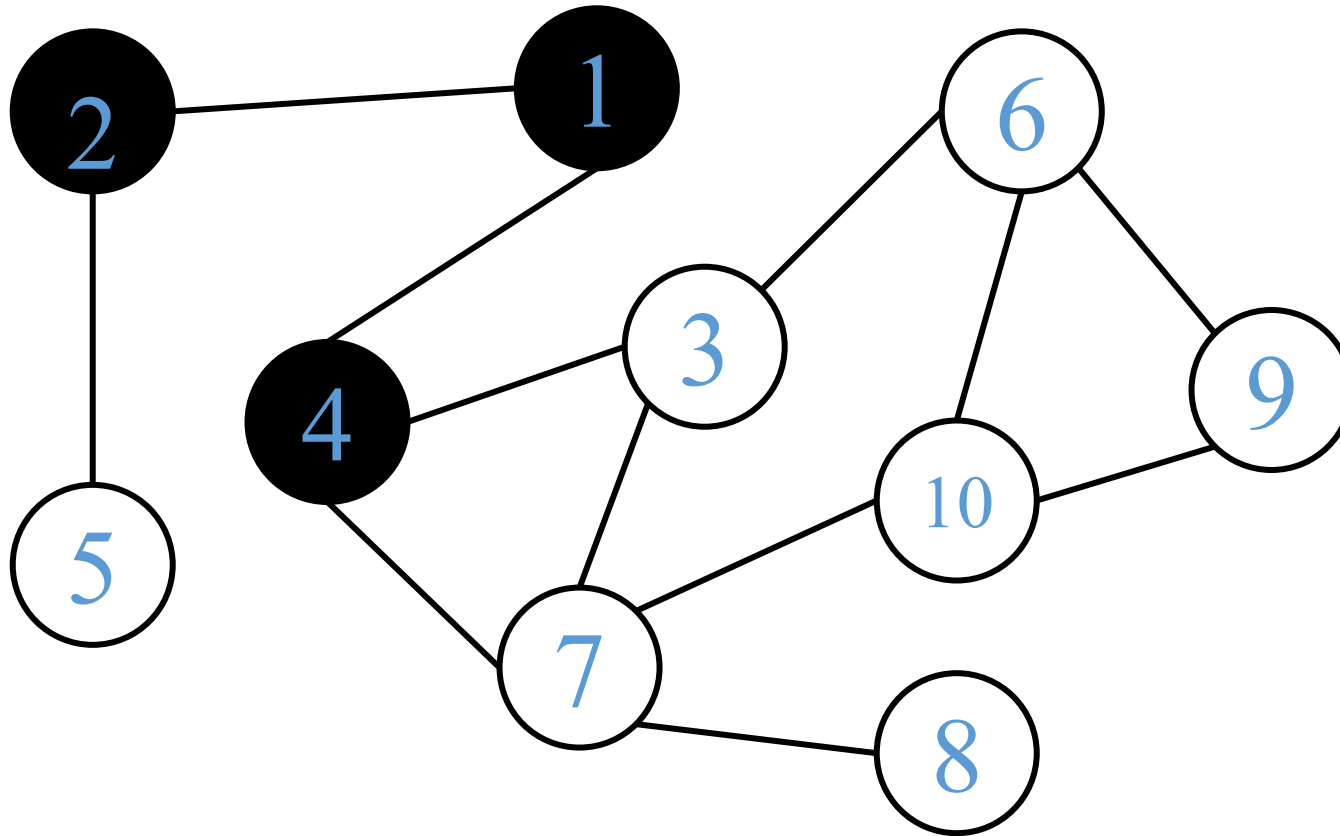
[8] = 0

[9] = 0

[10] = 0

Top of Q: 1

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 0

[4] = 1

[5] = 0

[6] = 0

[7] = 0

[8] = 0

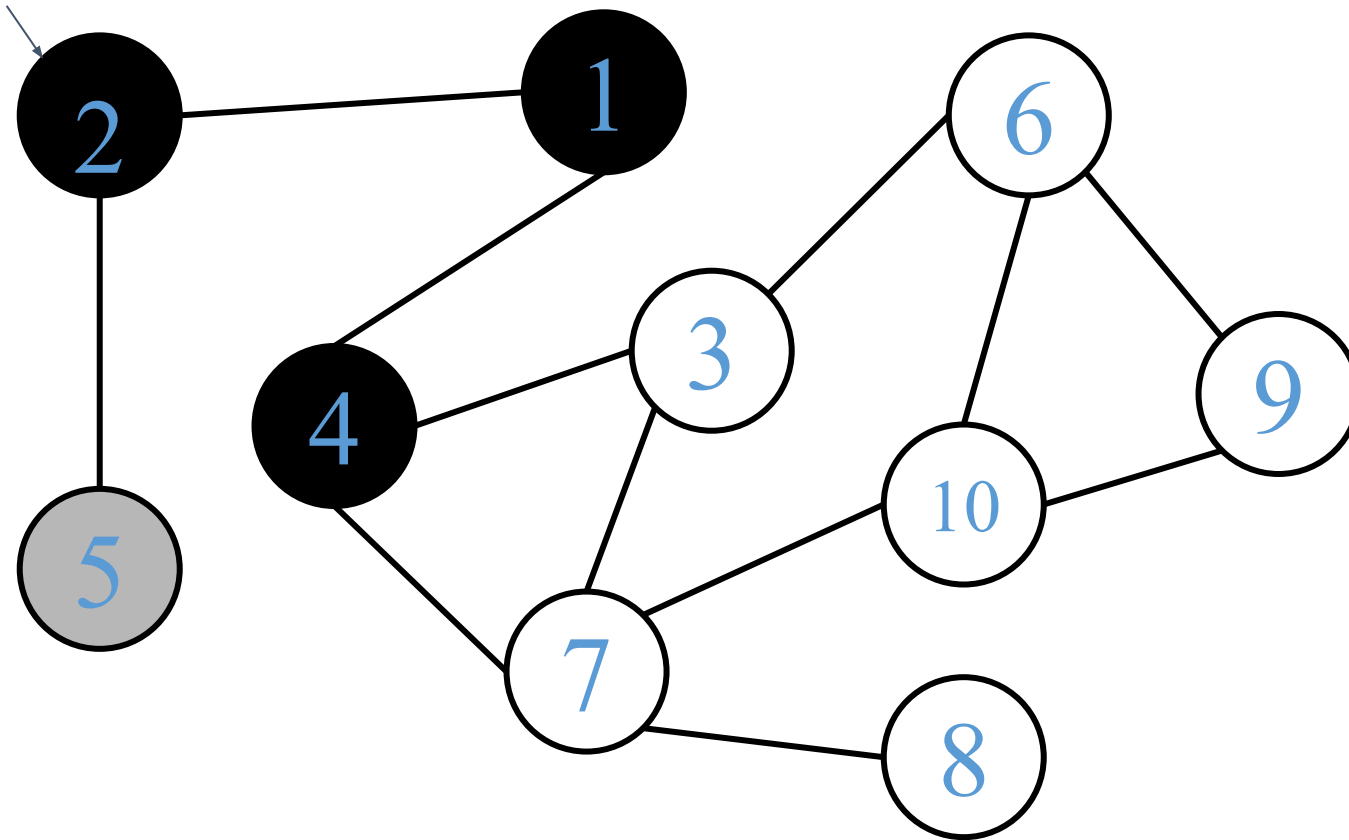
[9] = 0

[10] = 0

Top of Q42



## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 0

[4] = 1

[5] = 0

[6] = 0

[7] = 0

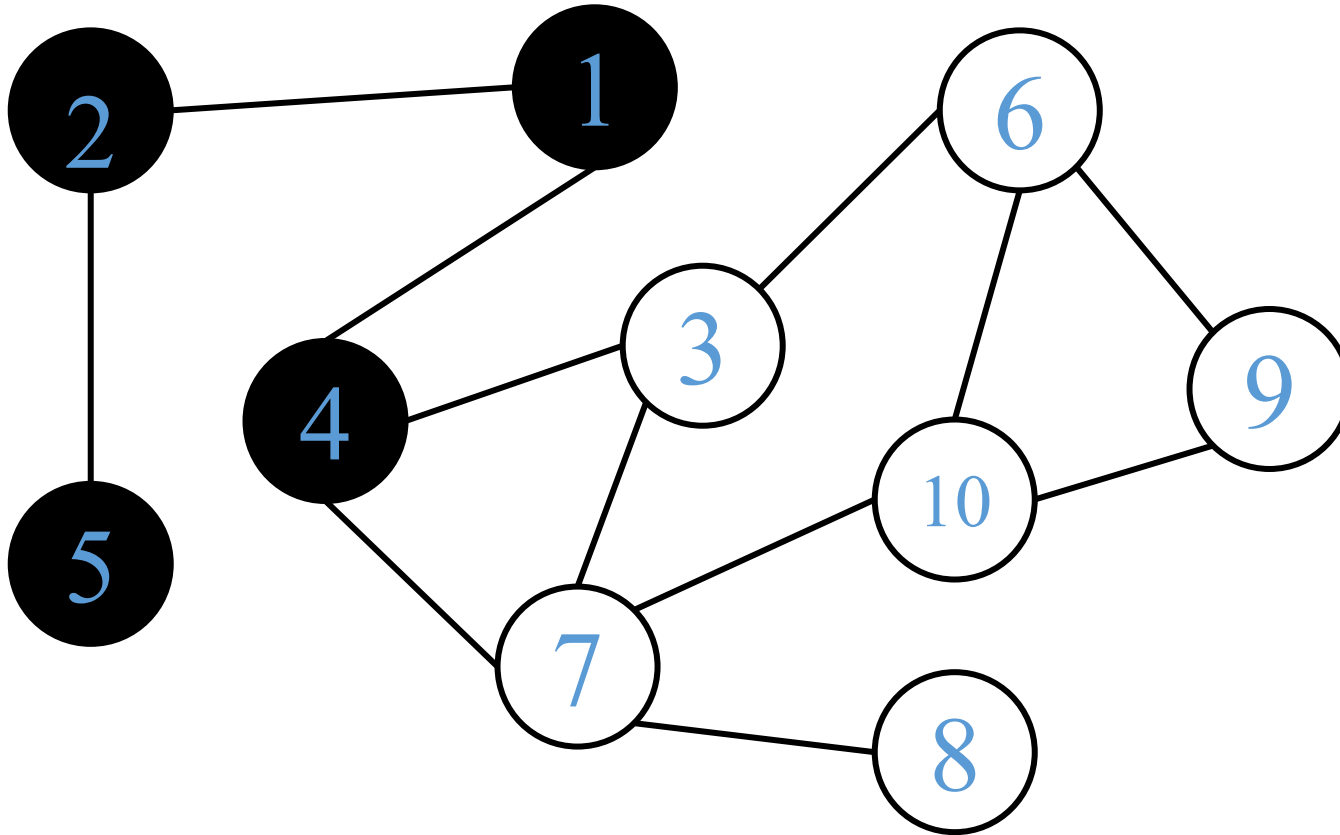
[8] = 0

[9] = 0

[10] = 0

Qp of Q42

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 0

[4] = 1

[5] = 1

[6] = 0

[7] = 0

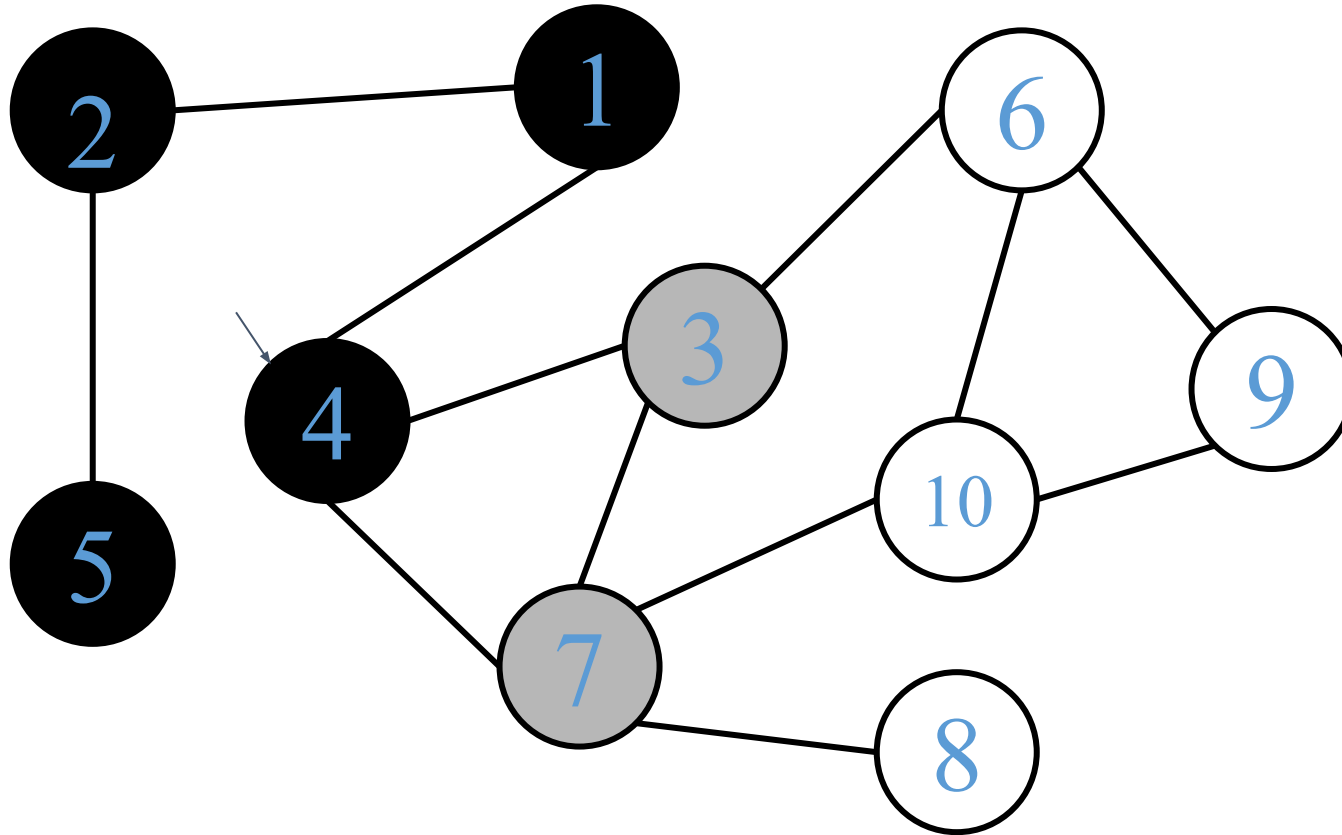
[8] = 0

[9] = 0

[10] = 0

Top of Q54

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 0

[4] = 1

[5] = 1

[6] = 0

[7] = 0

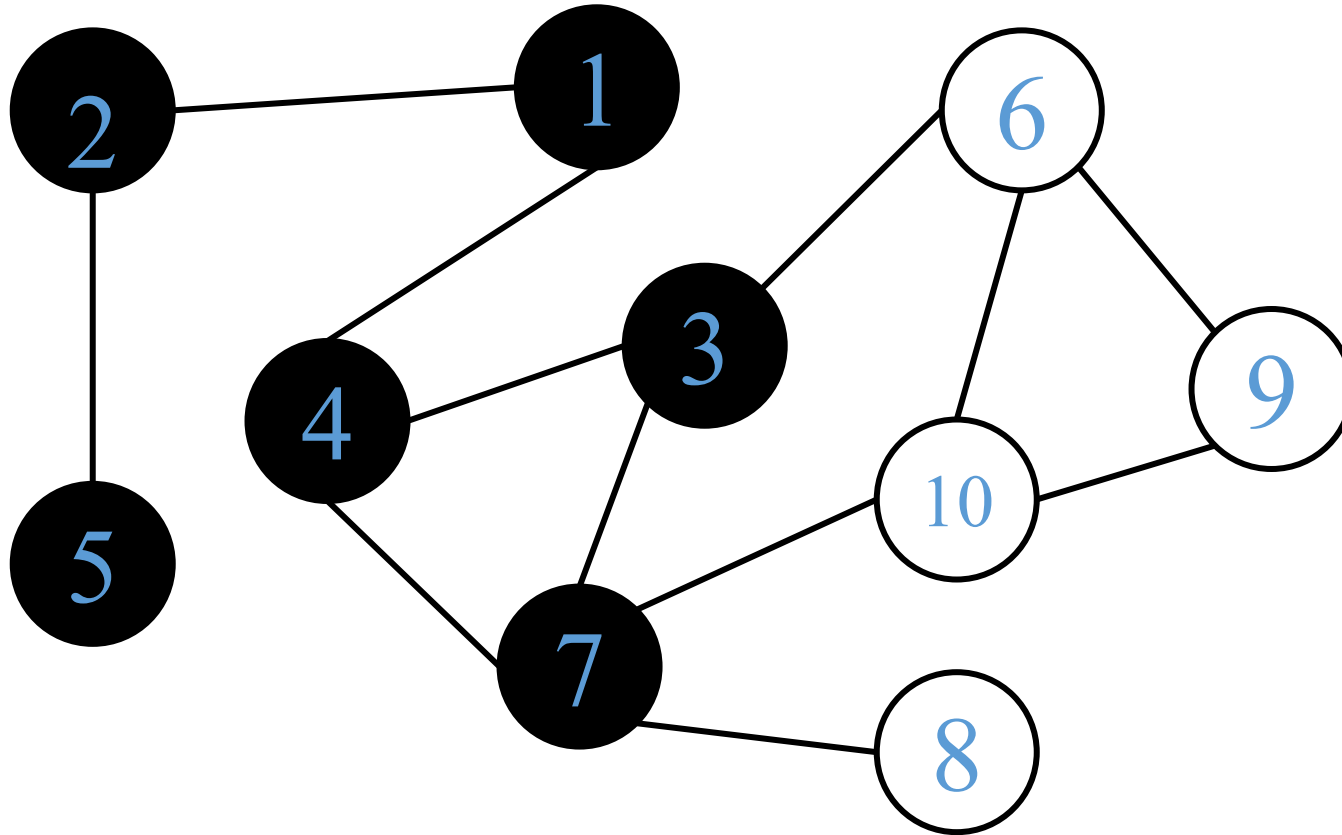
[8] = 0

[9] = 0

[10] = 0

Top of Q: 4

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 0

[7] = 1

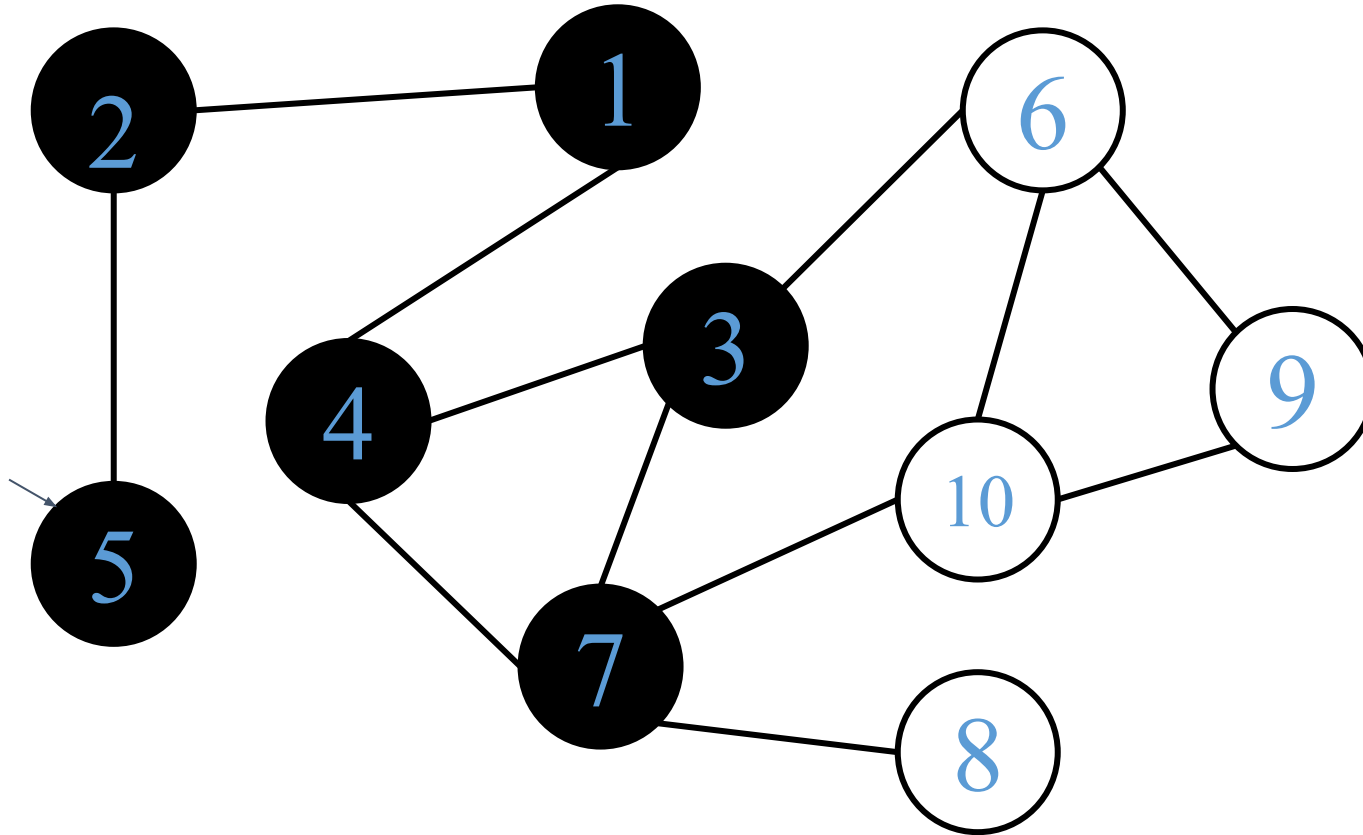
[8] = 0

[9] = 0

[10] = 0

Top of Q: 3 5 7

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 0

[7] = 1

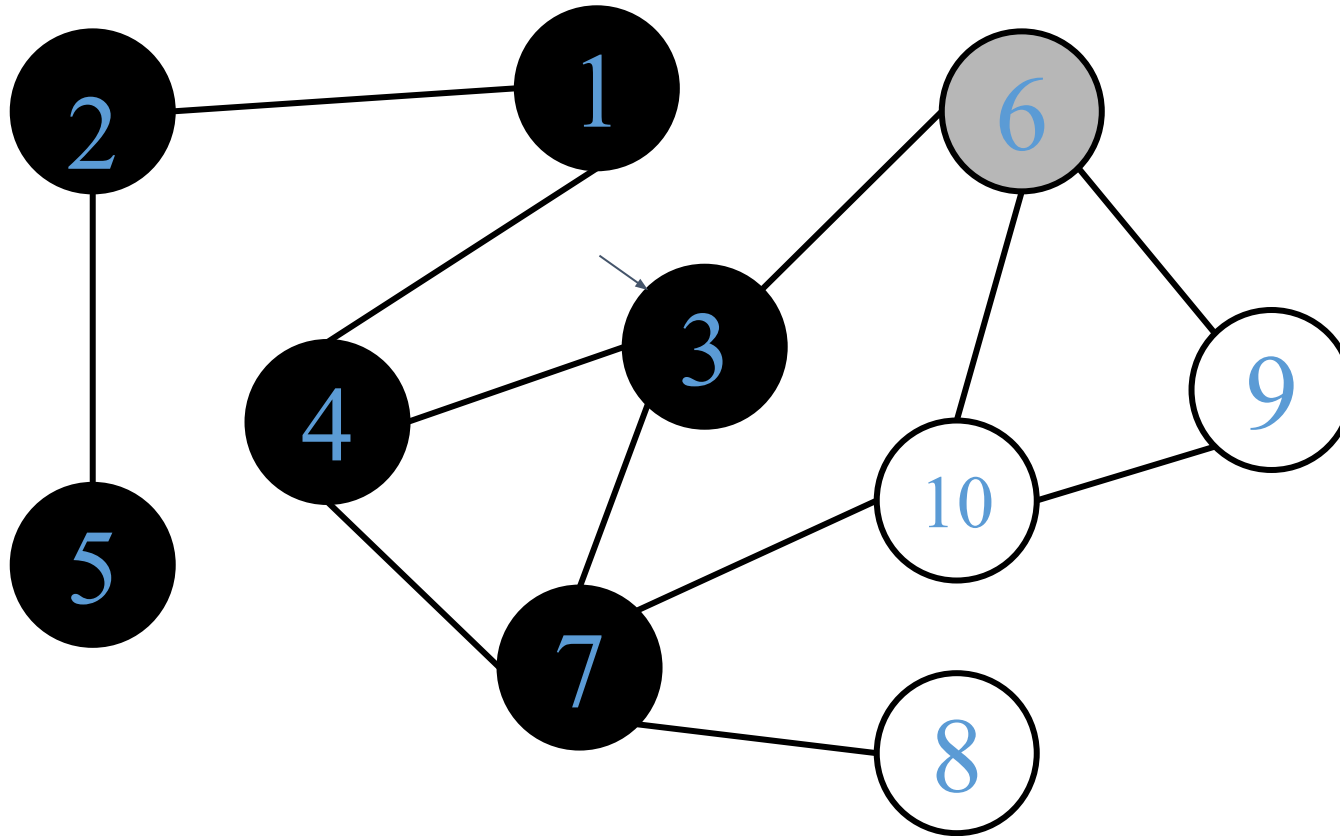
[8] = 0

[9] = 0

[10] = 0

Top of Q: 3 5 7

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 0

[7] = 1

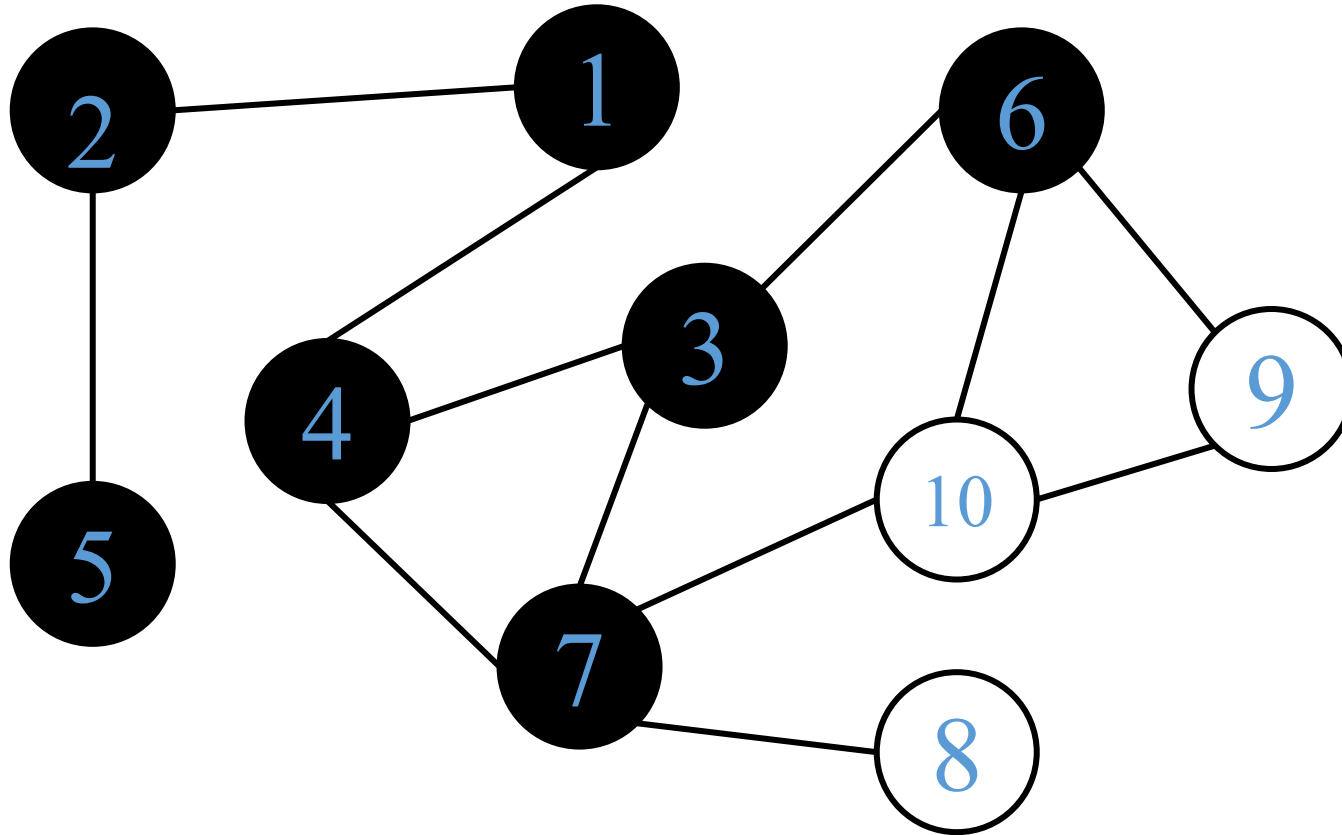
[8] = 0

[9] = 0

[10] = 0

Top of Q: 3

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 0

[7] = 1

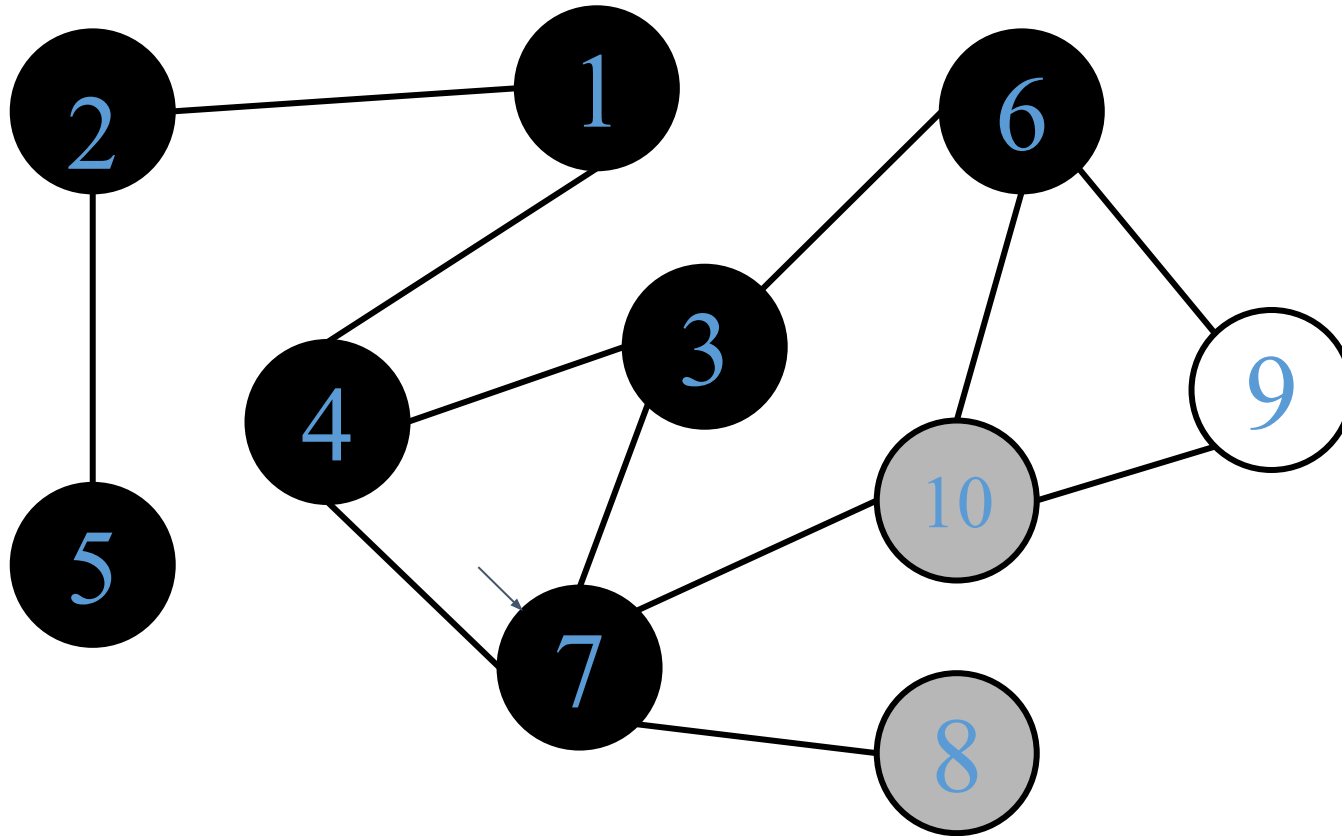
[8] = 0

[9] = 0

[10] = 0

Top of Q: 6 7

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 1

[7] = 1

[8] = 0

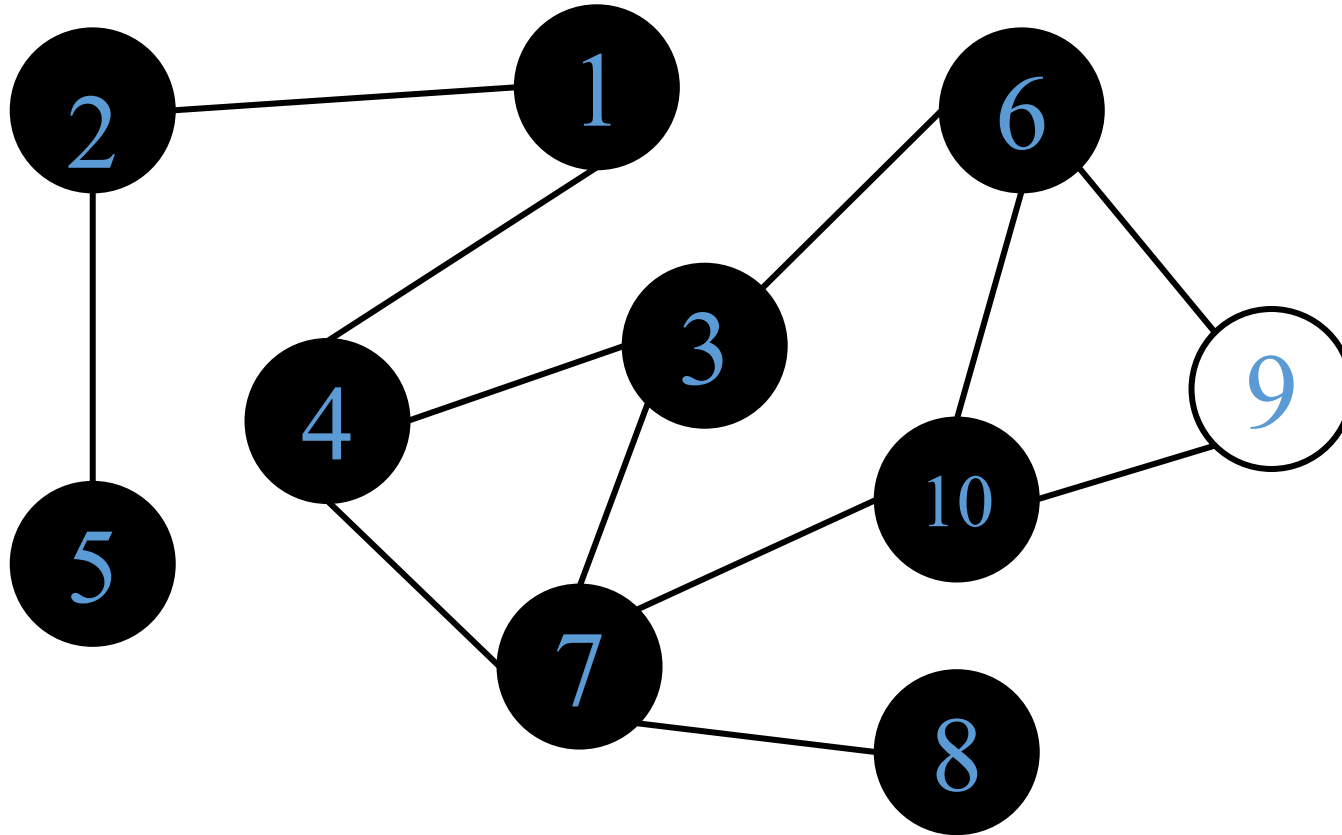
[9] = 0

[10] = 0

Top of Q: 6, 7



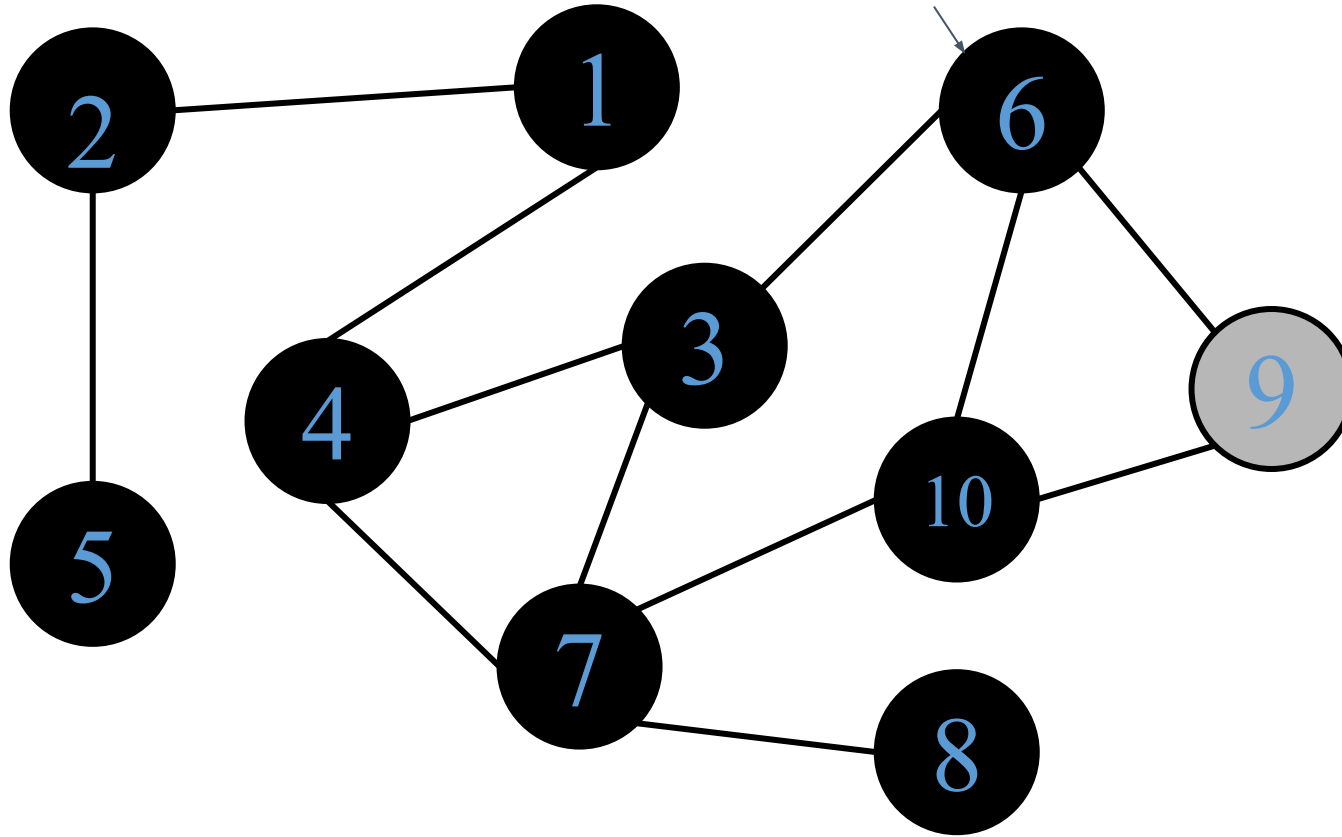
## Breadth-First Search: Example



visited[]  
[1] = 1  
[2] = 1  
[3] = 1  
[4] = 1  
[5] = 1  
[6] = 1  
[7] = 1  
[8] = 1  
[9] = 0  
[10] = 1

Top of Q: 8

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 1

[7] = 1

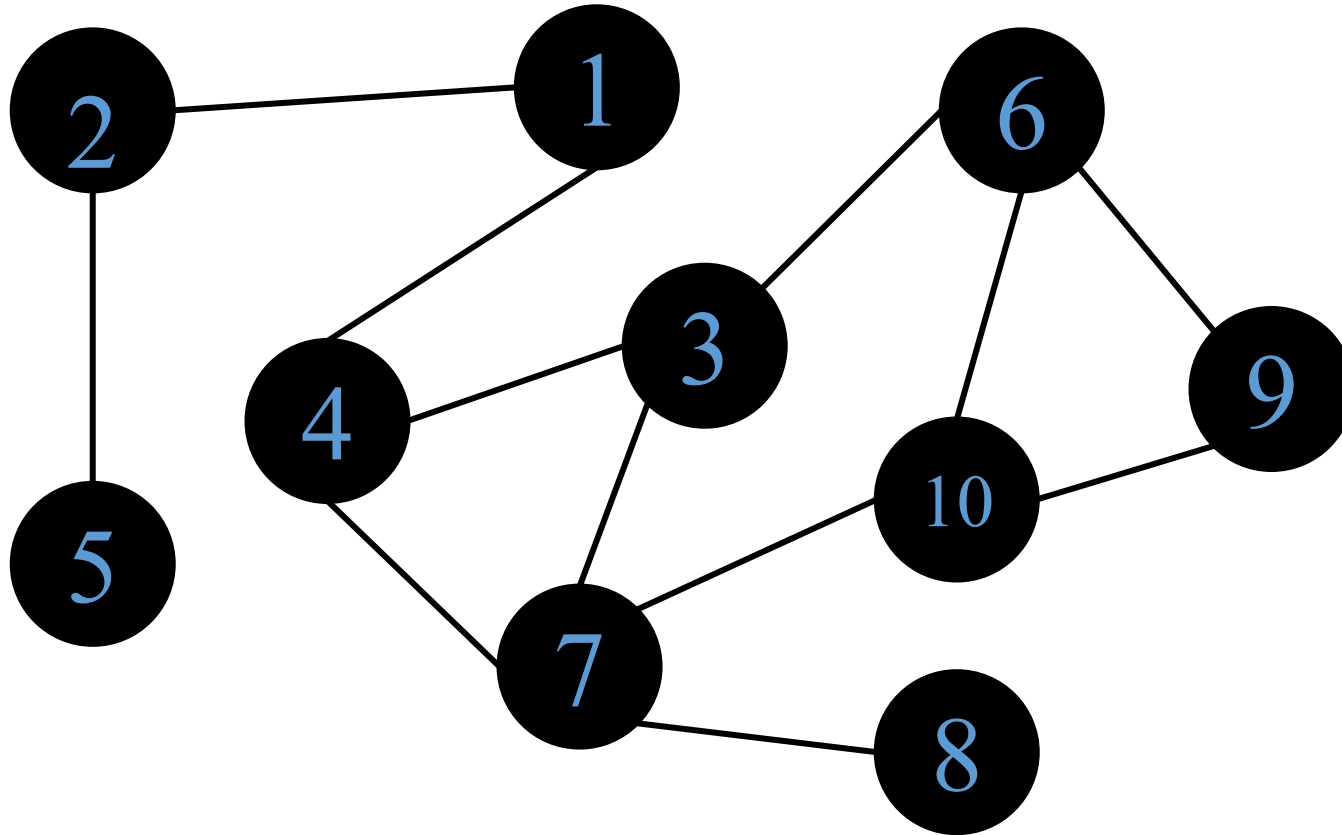
[8] = 1

[9] = 0

[10] = 1

Top of Q: 8

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 1

[7] = 1

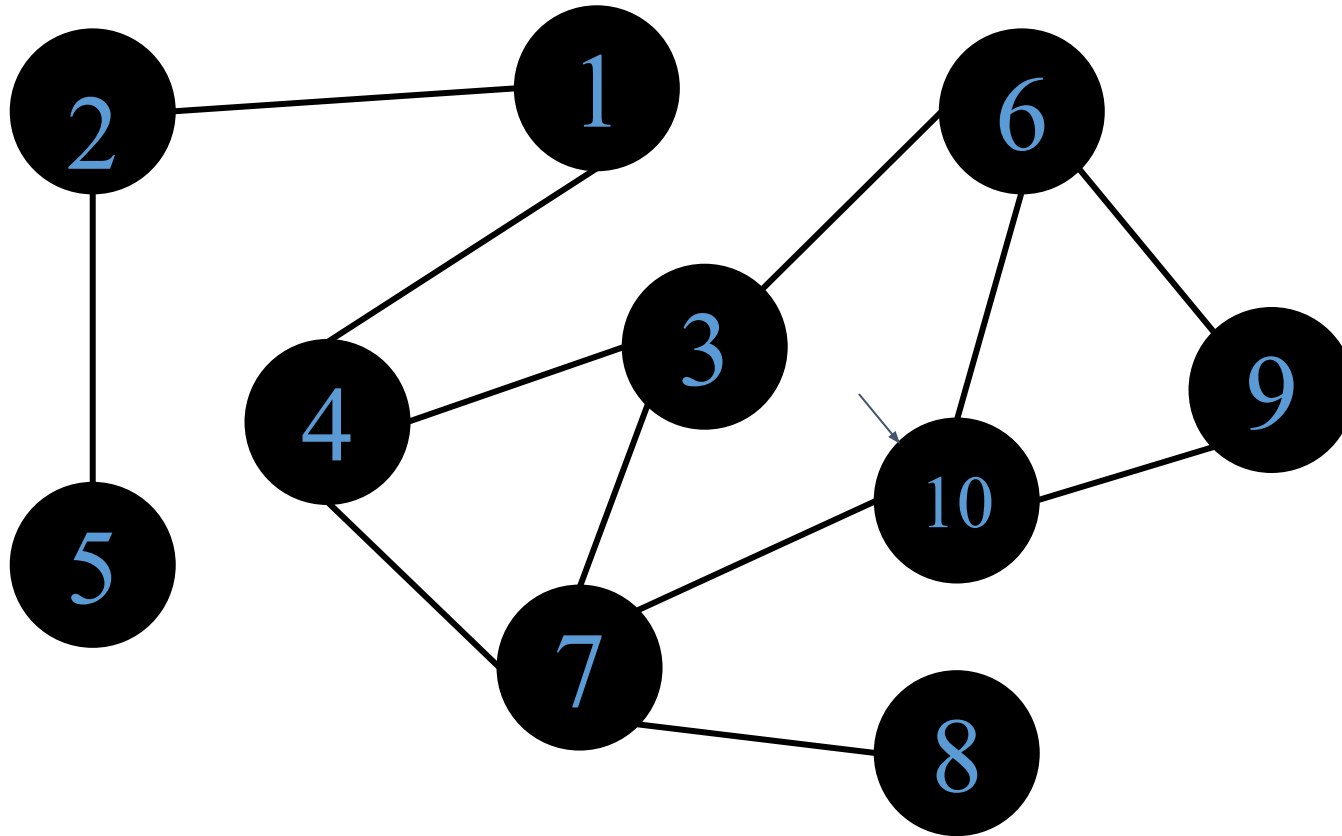
[8] = 1

[9] = 1

[10] = 1

Q: [1] [2] [3] [4] [5] [6] [7] [8] [9] [10]

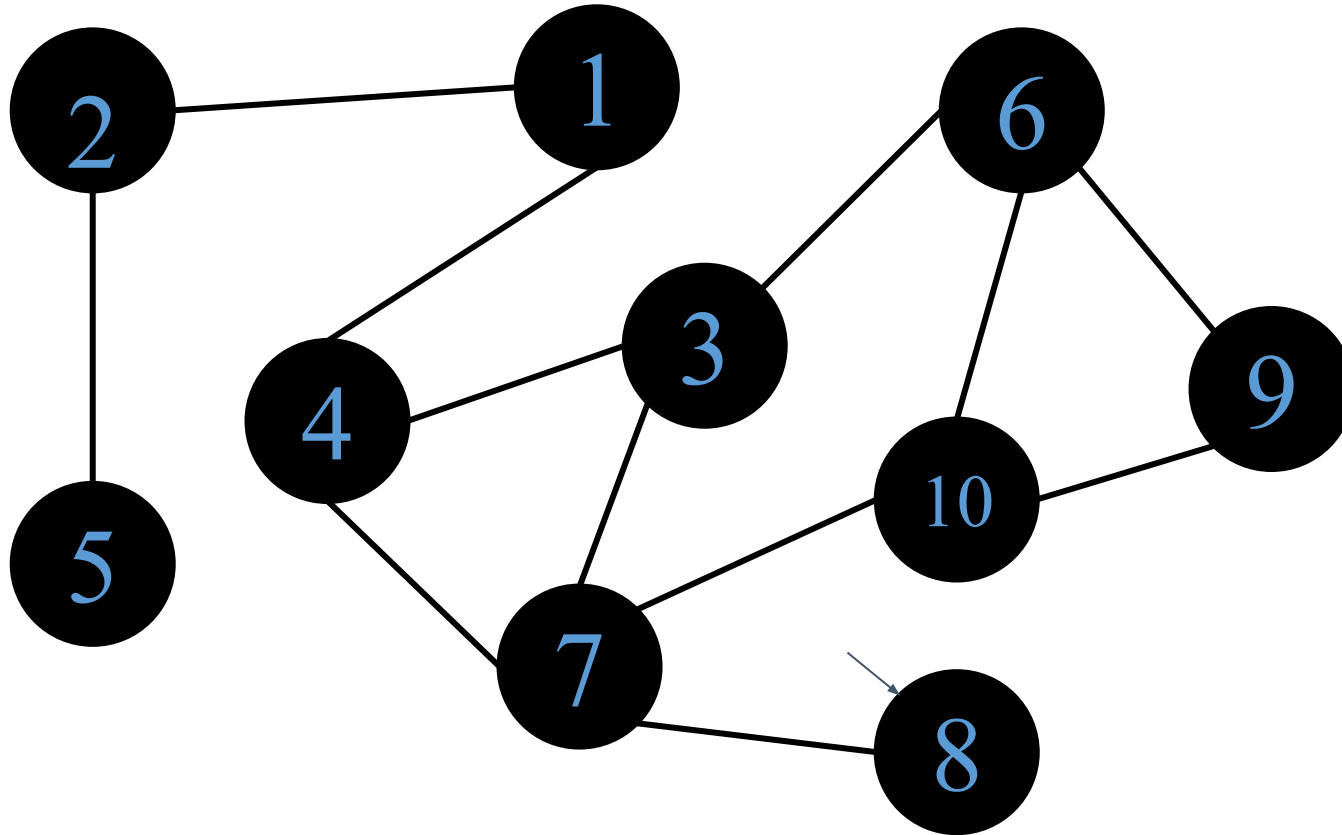
## Breadth-First Search: Example



visited[]  
[1] = 1  
[2] = 1  
[3] = 1  
[4] = 1  
[5] = 1  
[6] = 1  
[7] = 1  
[8] = 1  
[9] = 1  
[10] = 1

Q: 1 2 3 4 5 6 7 8 9 10

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 1

[7] = 1

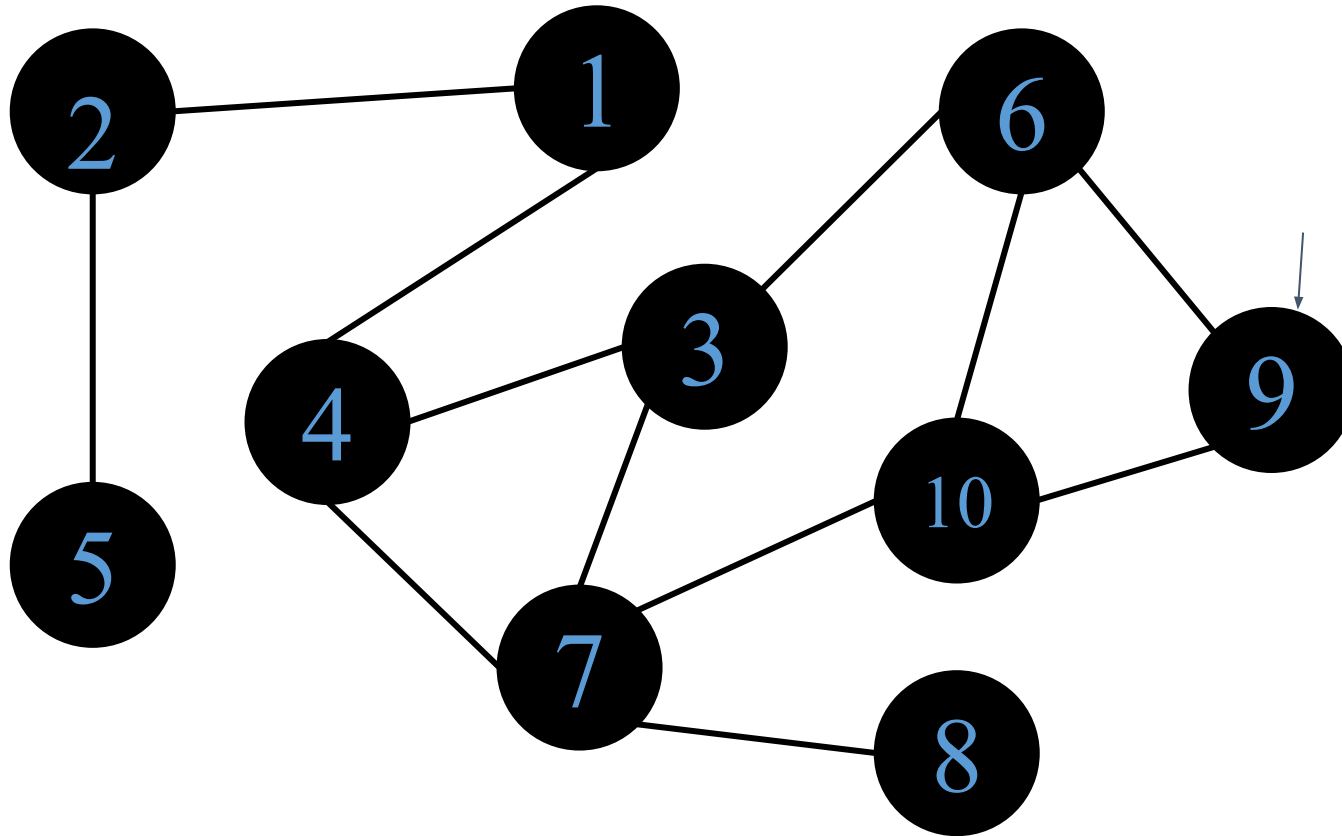
[8] = 1

[9] = 1

[10] = 1

Top of Q: 8

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 1

[7] = 1

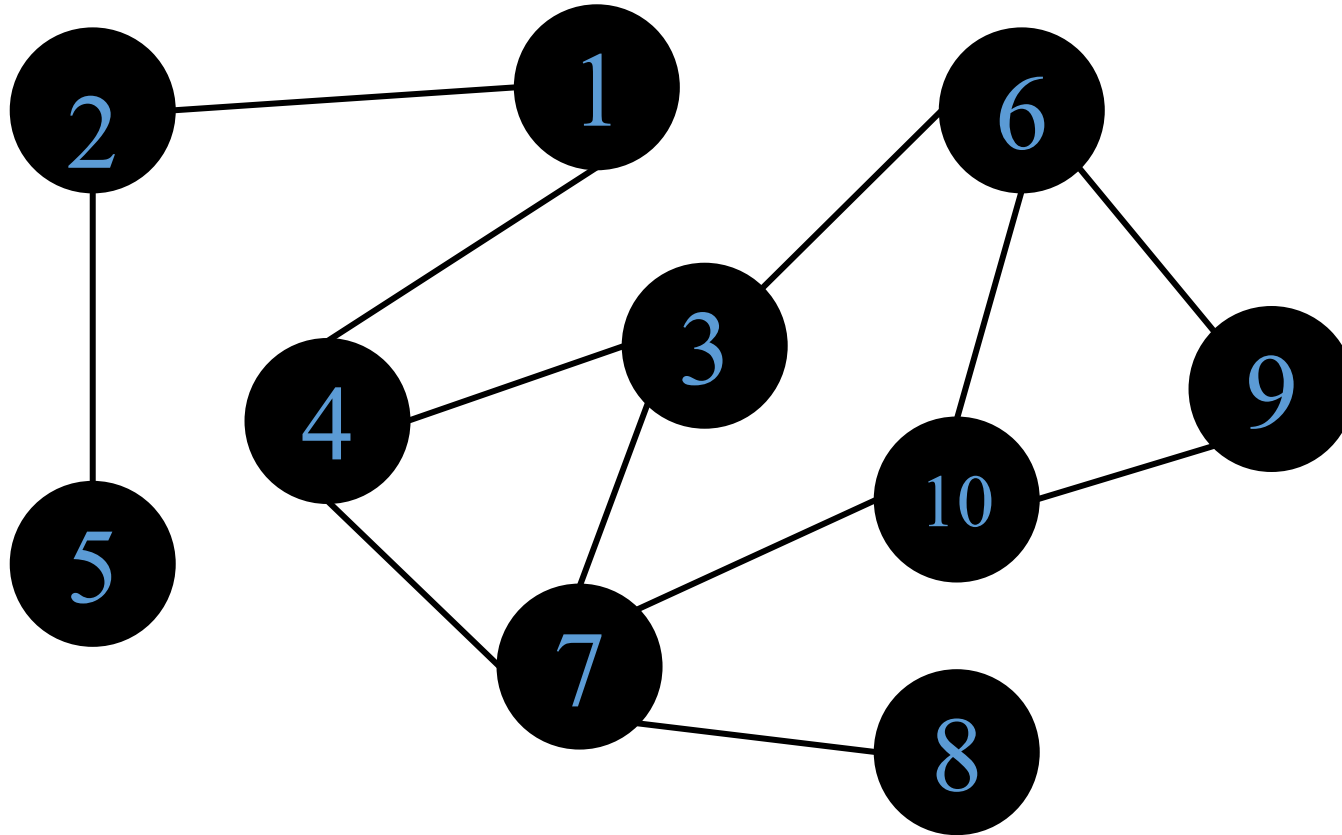
[8] = 1

[9] = 1

[10] = 1

Top of Q: 9

## Breadth-First Search: Example



visited[]

[1] = 1

[2] = 1

[3] = 1

[4] = 1

[5] = 1

[6] = 1

[7] = 1

[8] = 1

[9] = 1

[10] = 1

*Q:*

# BFS Running Time

Initialization of each vertex takes  $O(V)$  time

Every vertex is enqueued once and dequeued once, taking  $O(V)$  time

When a vertex is dequeued, all its neighbors are checked to see if they are unvisited, taking time proportional to number of neighbors of the vertex, and summing to  $O(E)$  over all iterations

Total time is  $O(V+E)$



# Breadth-First Search: Properties

BFS calculates the *shortest-path distance* to the source node

Shortest-path distance  $\delta(s,v)$  = minimum number of edges from  $s$  to  $v$ , or  $\infty$  if  $v$  not reachable from  $s$

BFS builds *breadth-first tree*, in which paths to root represent shortest paths in  $G$

Thus we can use BFS to calculate shortest path from one vertex to another in  $O(V+E)$  time

# Depth-First Search

*Depth-first search* is another strategy for exploring a graph

- Explore “deeper” in the graph whenever possible

- Edges are explored out of the most recently discovered vertex  $v$  that still has unexplored edges

- When all of  $v$ 's edges have been explored, backtrack to the vertex from which  $v$  was discovered

# Depth-First Search

Vertices initially colored white

Then colored gray when discovered

Then black when finished

# DFS Tree

Actually might be a DFS forest (collection of trees)

Keep track of parents

# Depth-First Search

## DFS ( $G$ )

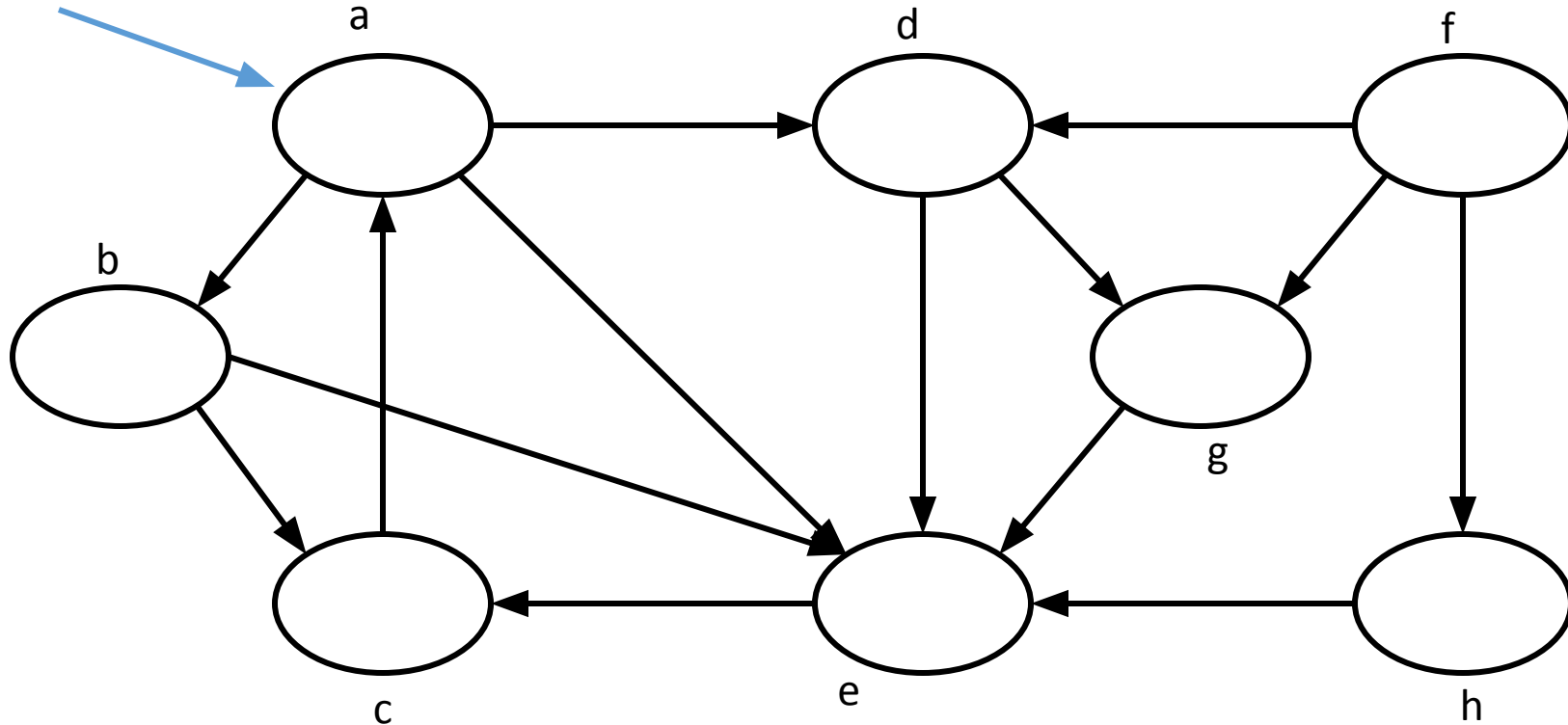
1. **for** each vertex  $u \in G.V$
2.      $u.color = \text{WHITE}$
3.      $u.\pi = \text{NIL}$
4.  $time = 0$
5. **for** each vertex  $u \in G.V$
6.     **if**  $u.color == \text{WHITE}$
7.         DFS-VISIT( $G, u$ )

## DFS-VISIT( $G, u$ )

1.  $time = time + 1$
2.  $u.d = time$
3.  $u.color = \text{GRAY}$
4. **for** each  $v \in G.Adj[u]$
5.     **if**  $v.color == \text{WHITE}$
6.          $v.\pi = u$
7.         DFS-VISIT( $G, v$ )
8.  $u.color = \text{BLACK}$
9.  $time = time + 1$
10.  $u.f = time$

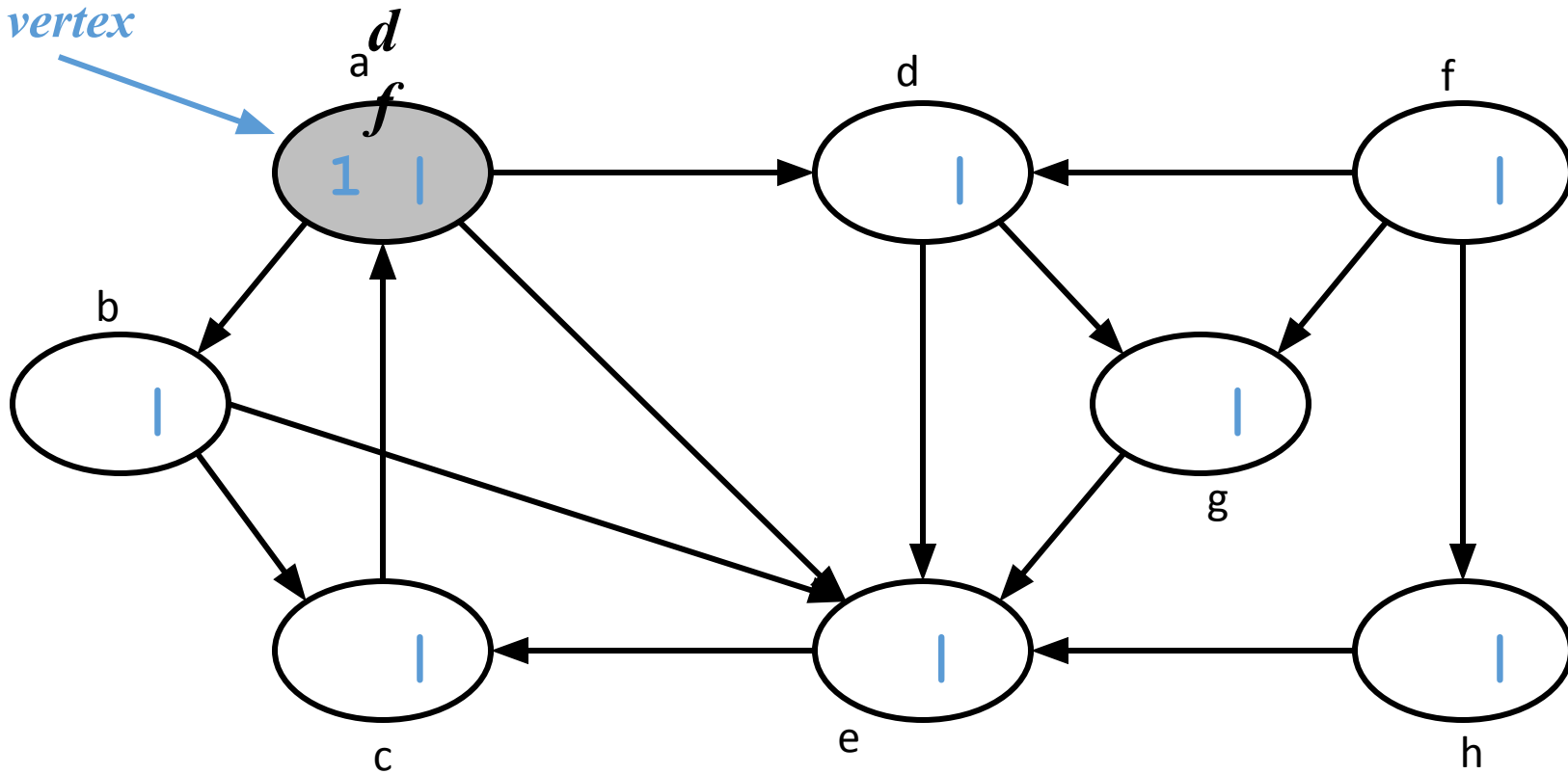
# DFS Example

*source  
vertex*



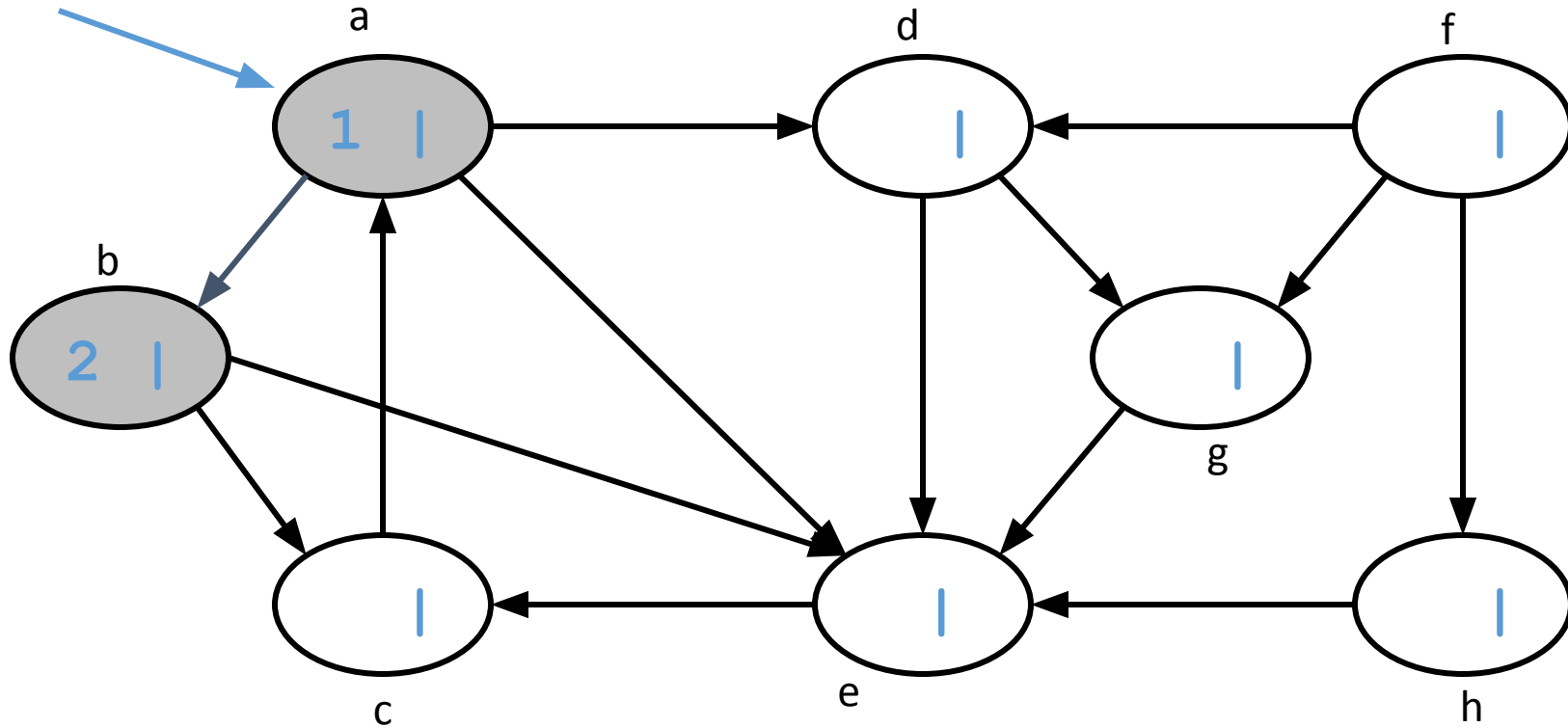
# DFS Example

*source  
vertex*



# DFS Example

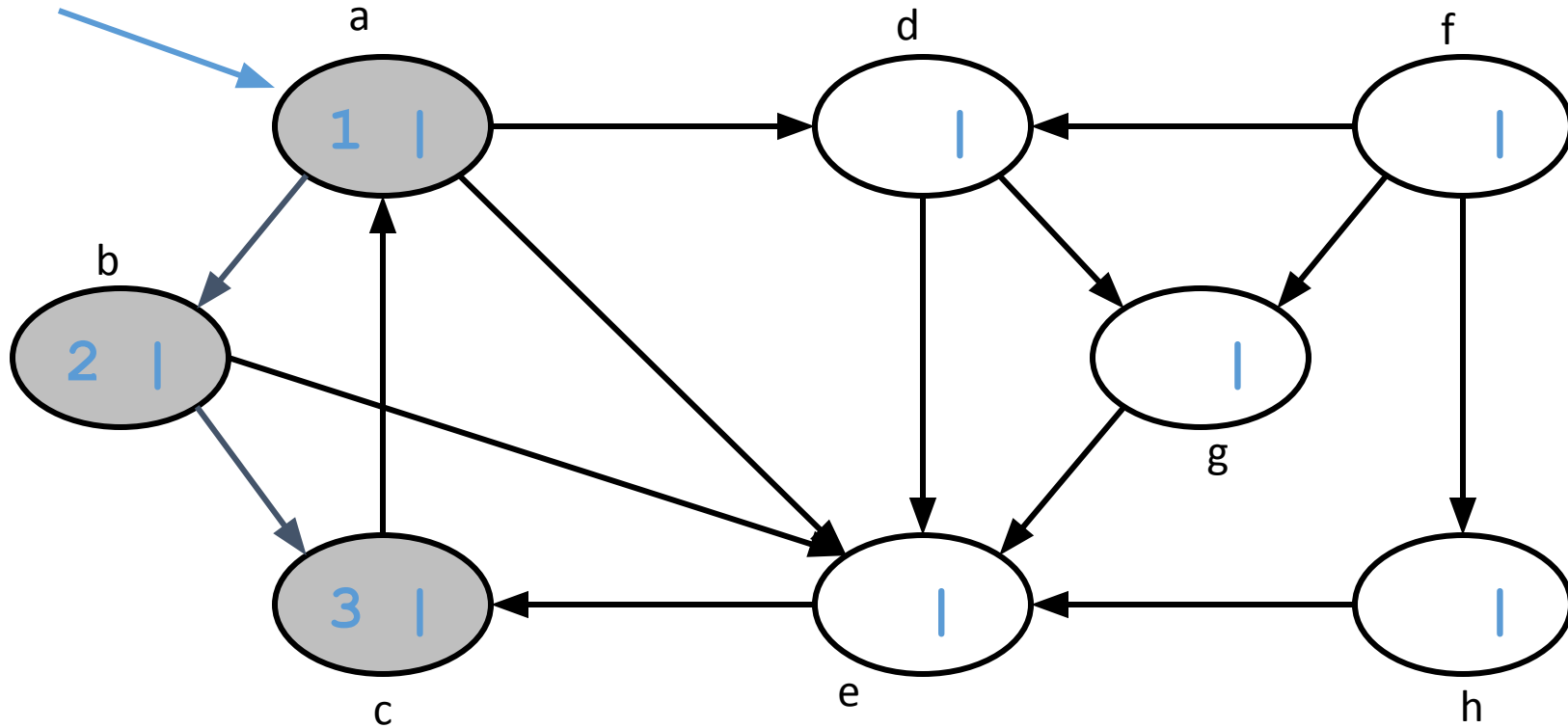
*source  
vertex*





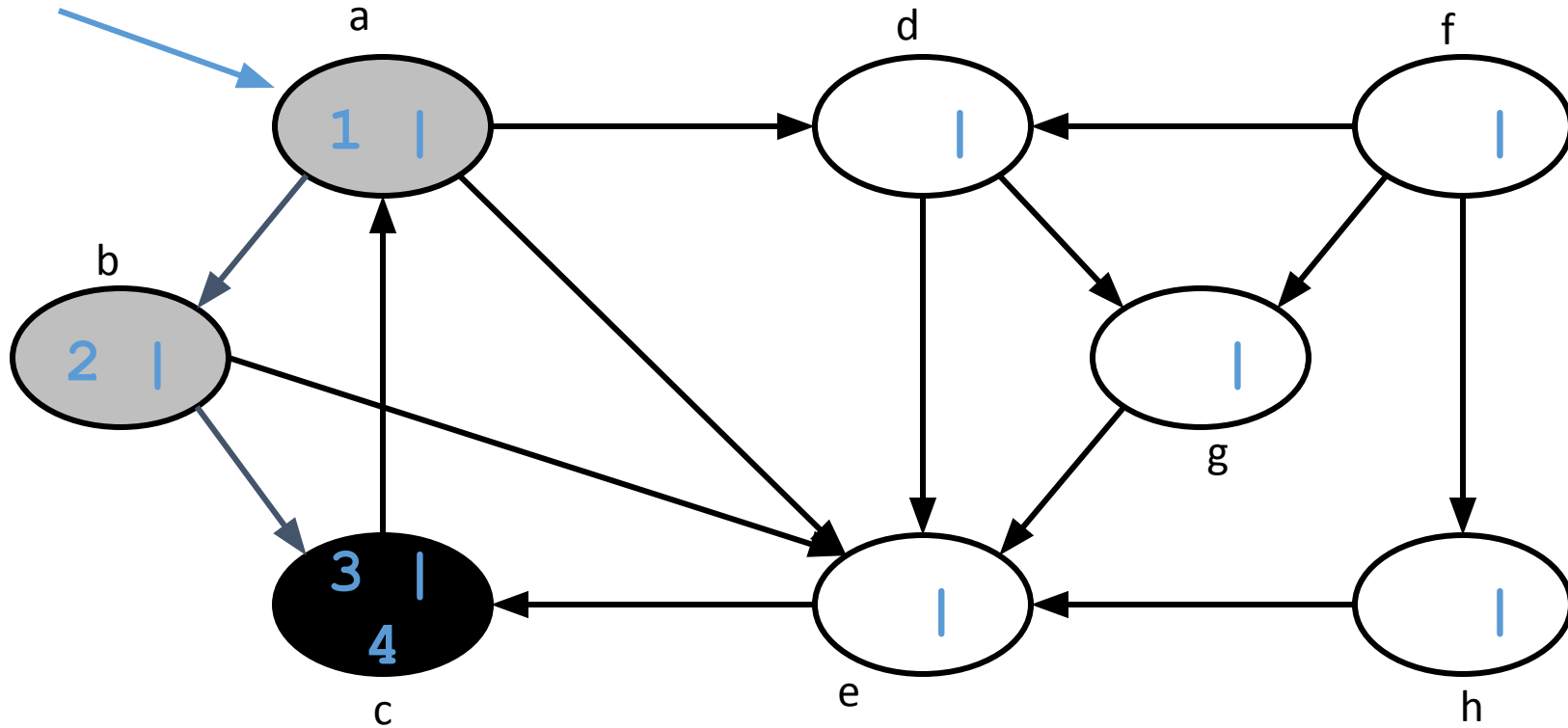
# DFS Example

*source  
vertex*



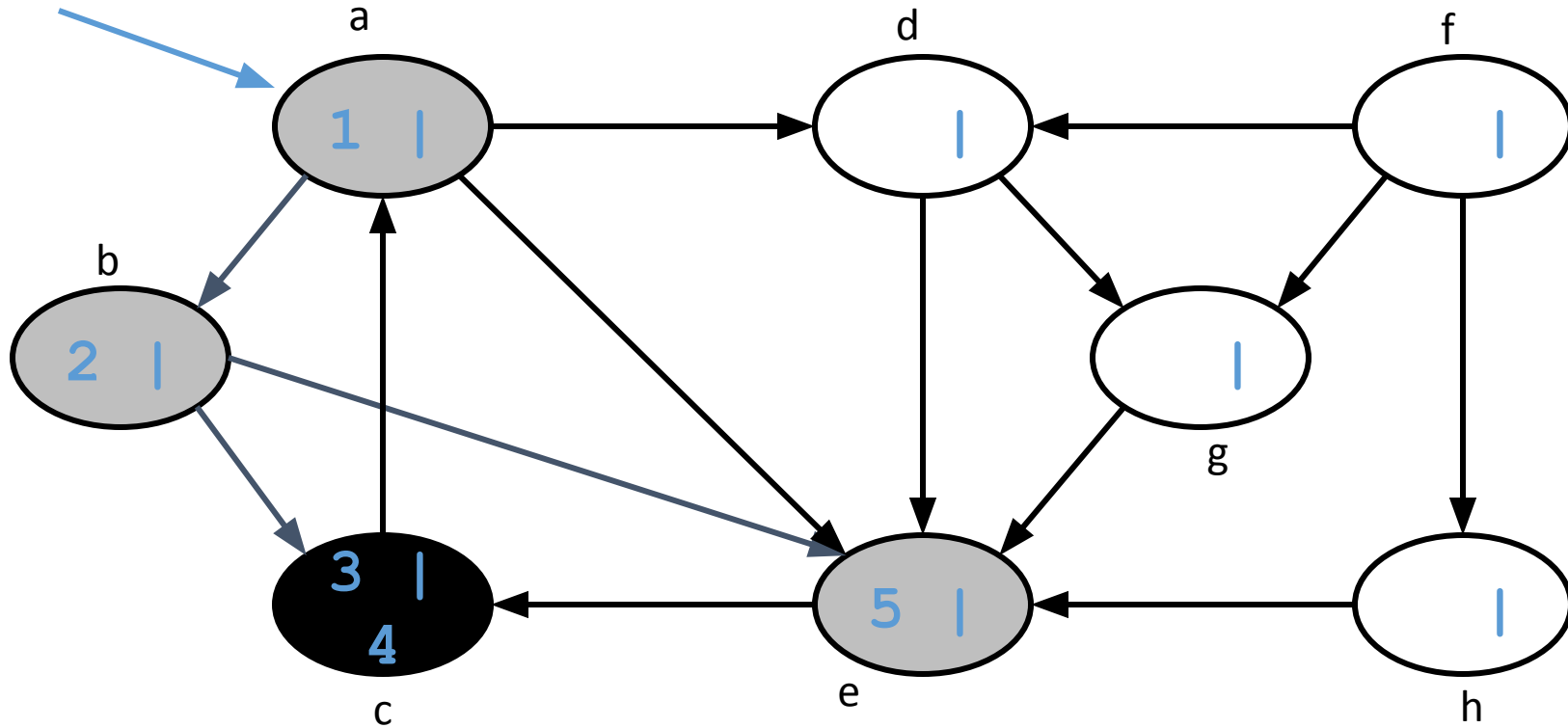
# DFS Example

*source  
vertex*



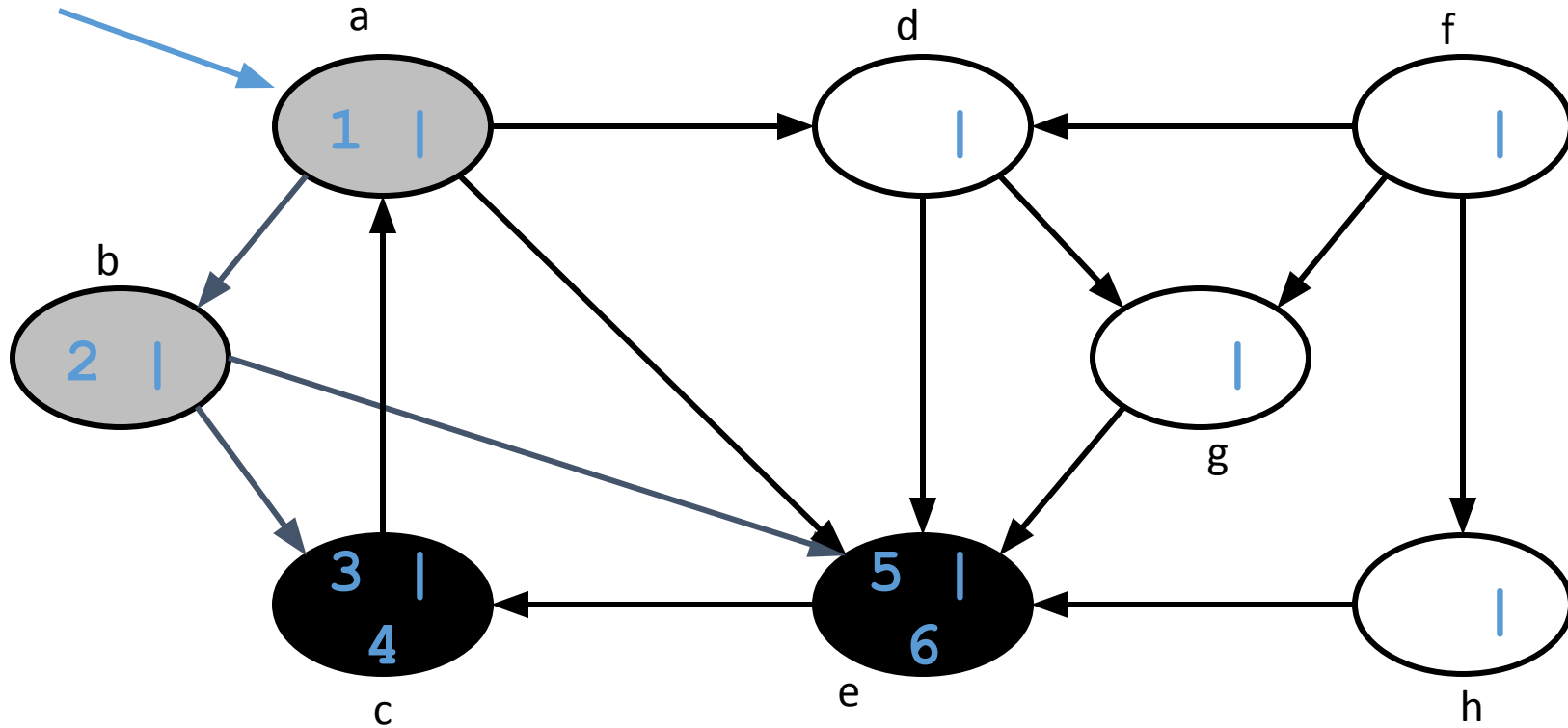
# DFS Example

*source  
vertex*



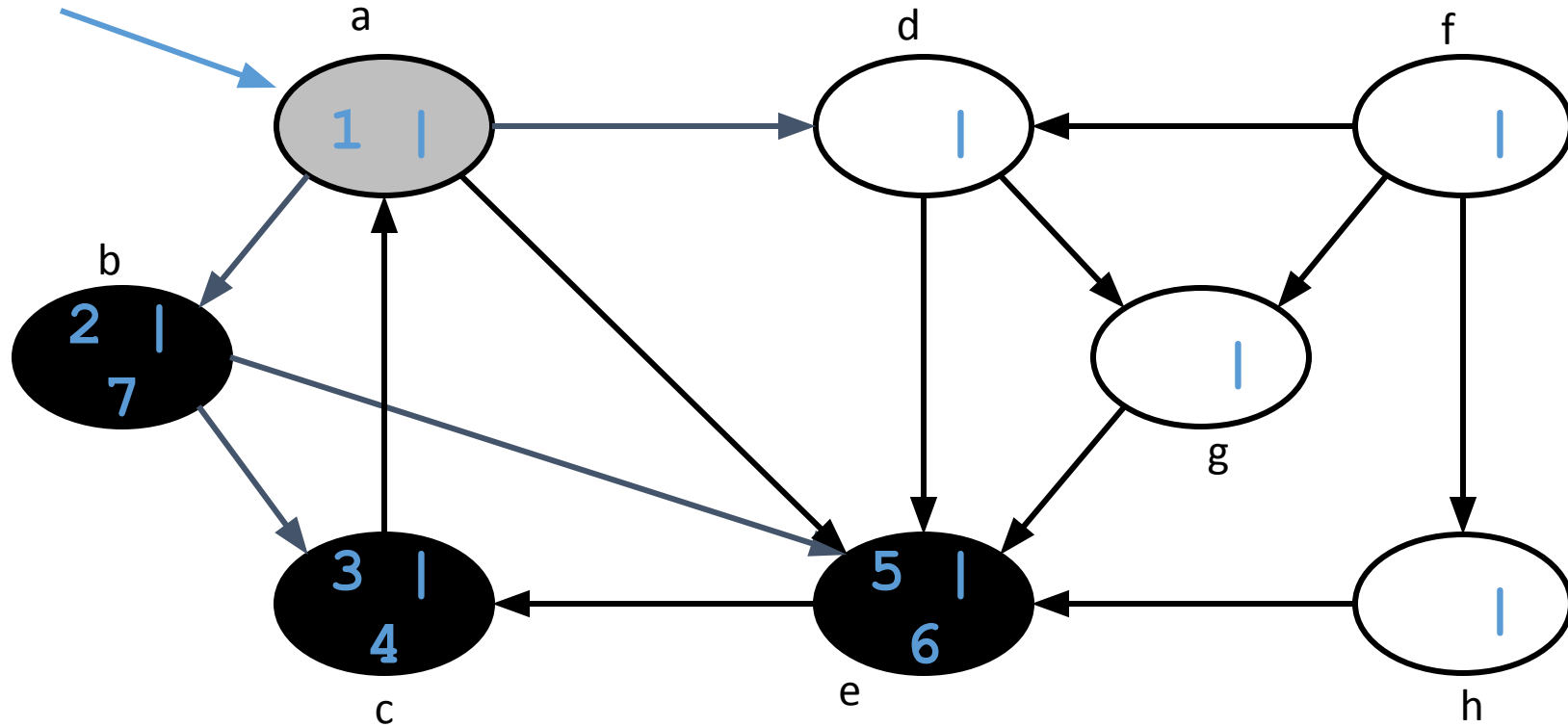
# DFS Example

*source  
vertex*



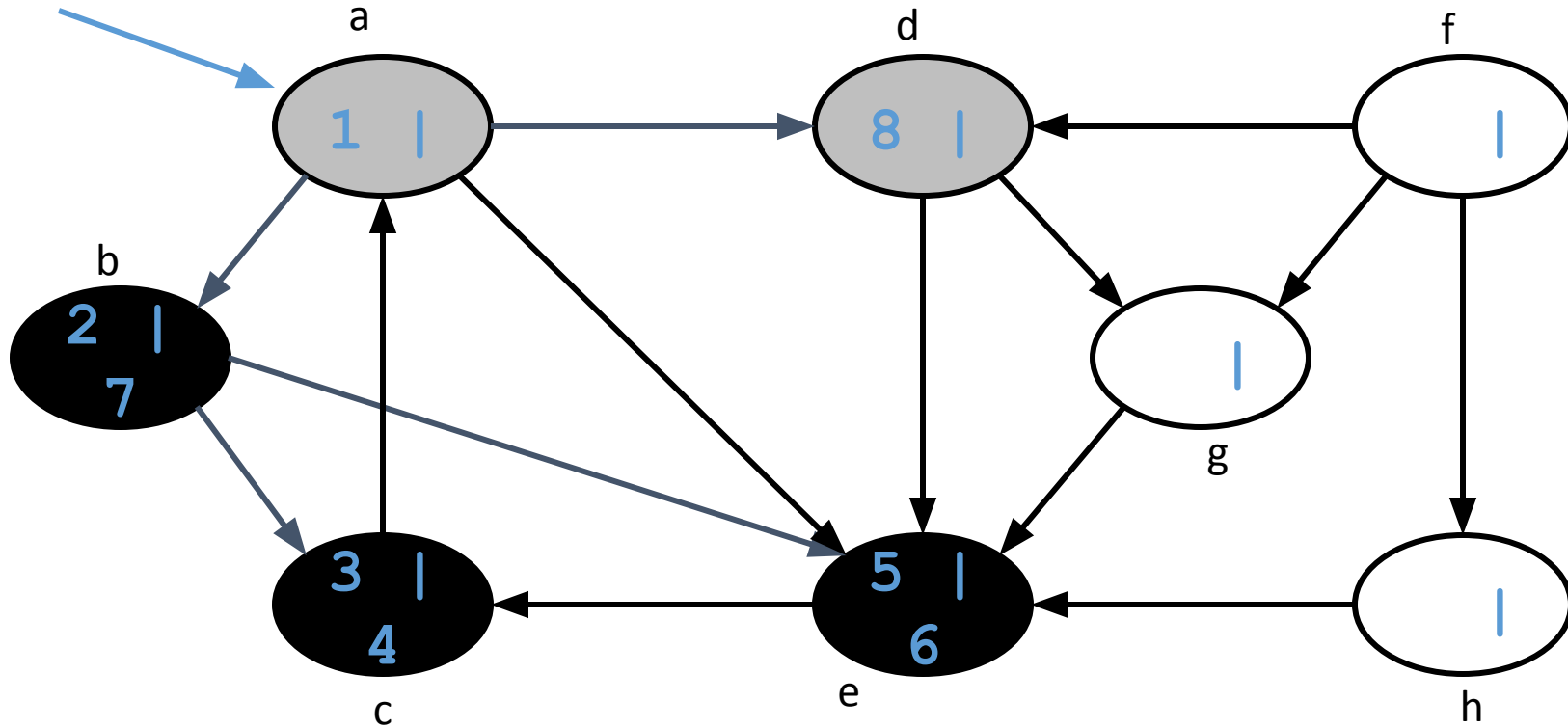
# DFS Example

*source  
vertex*



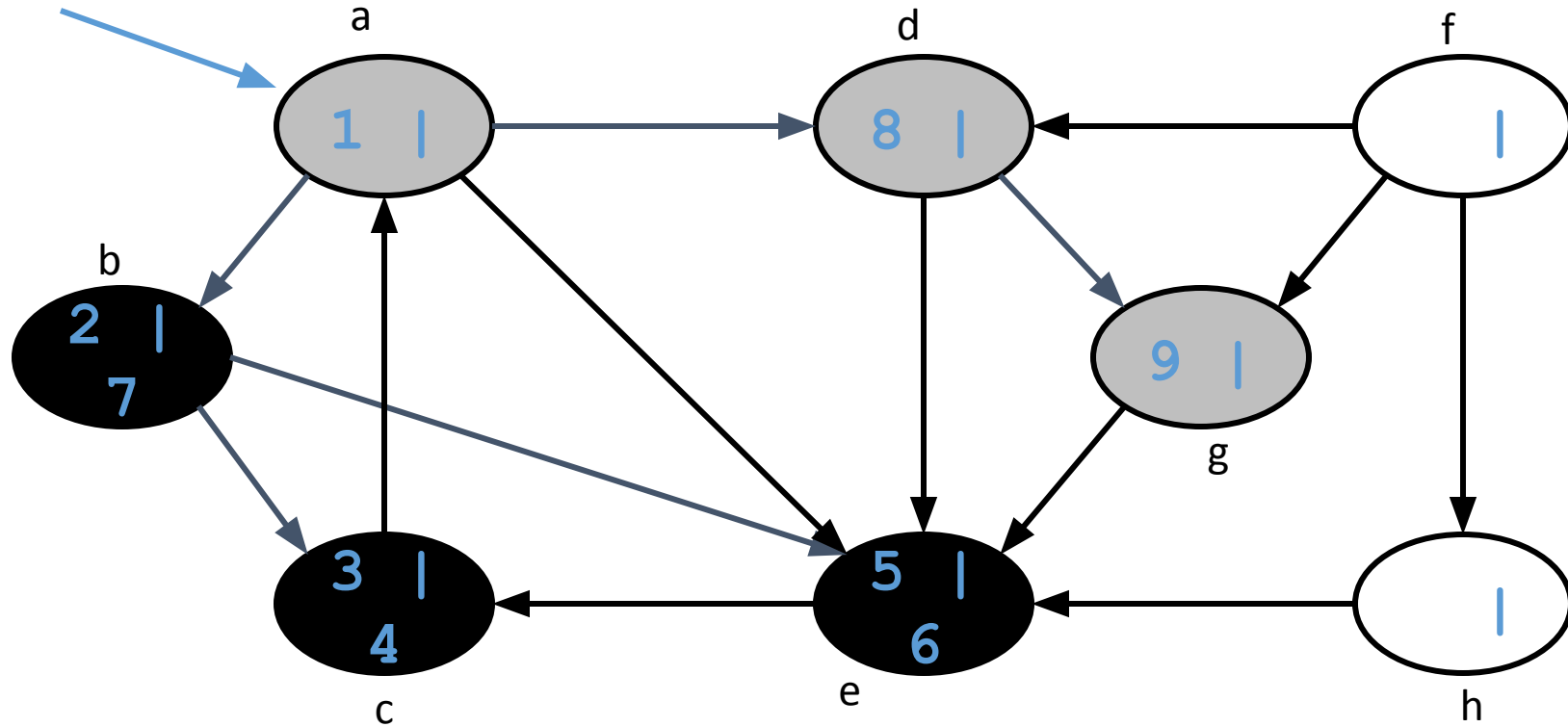
# DFS Example

*source  
vertex*



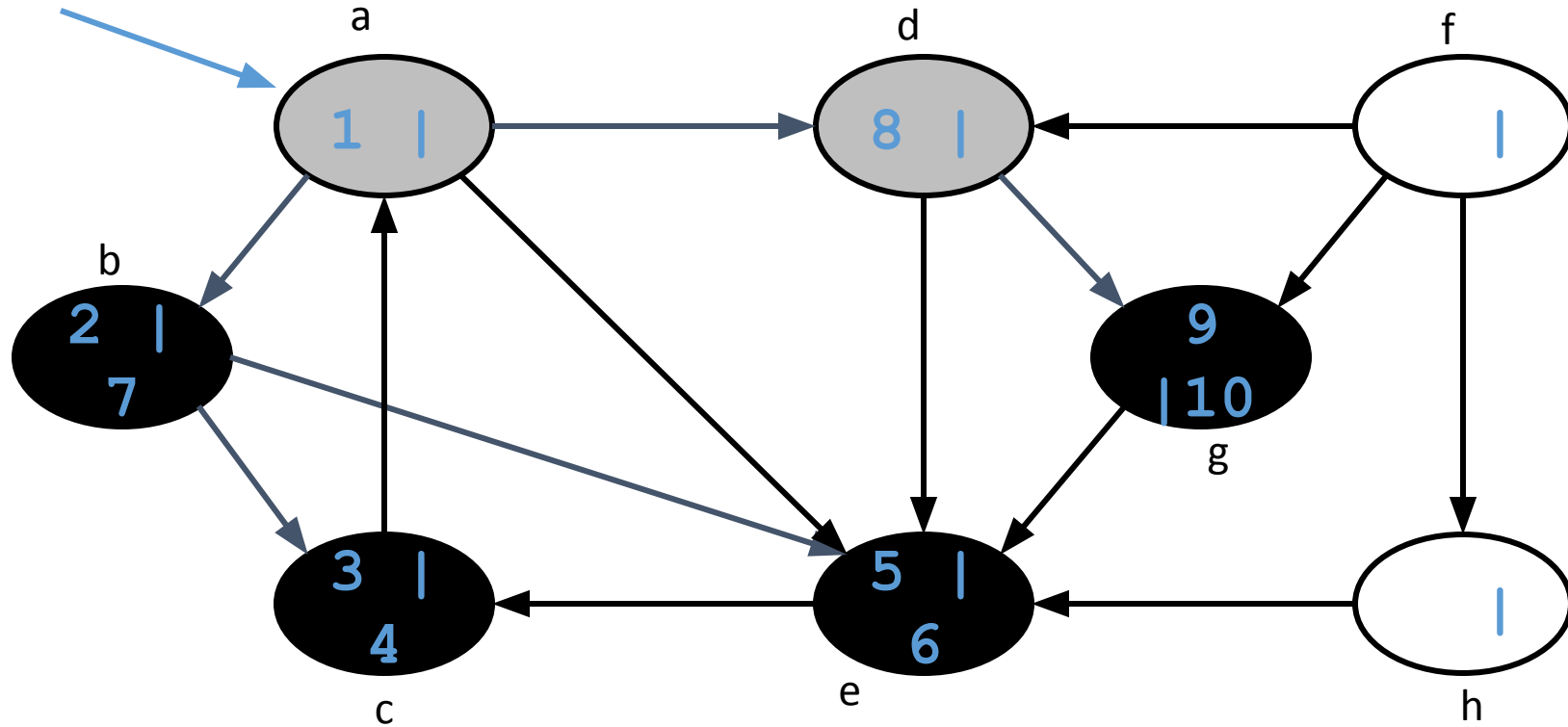
# DFS Example

*source  
vertex*



# DFS Example

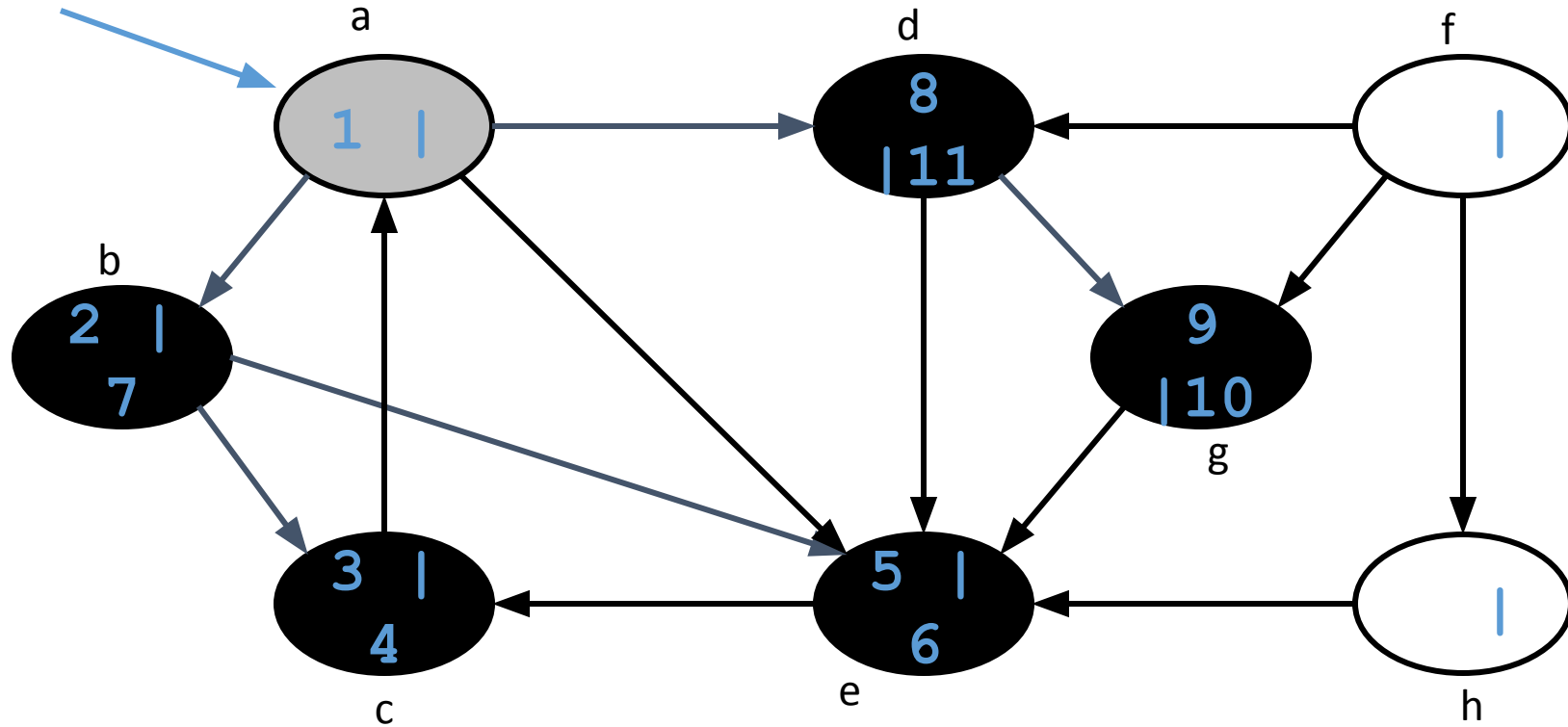
*source  
vertex*





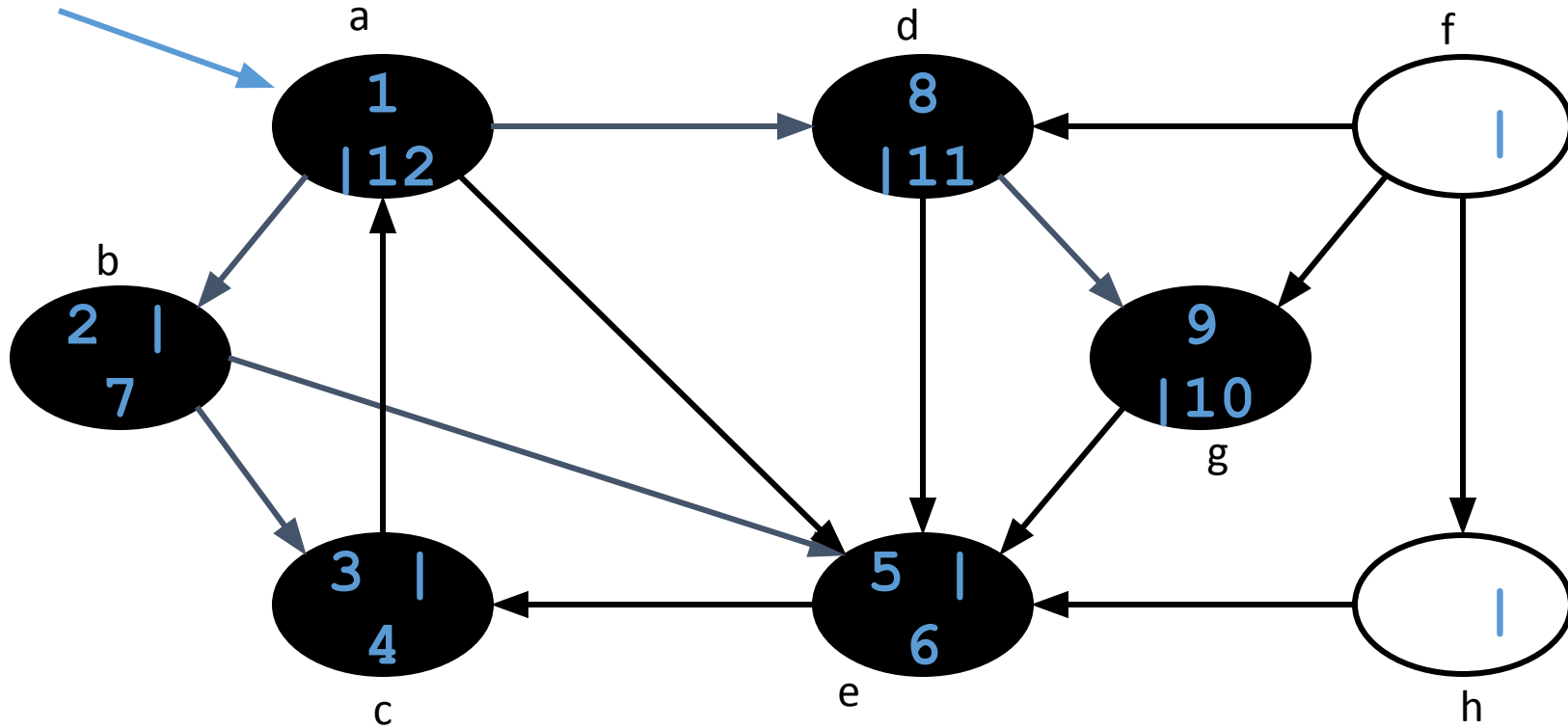
# DFS Example

*source  
vertex*



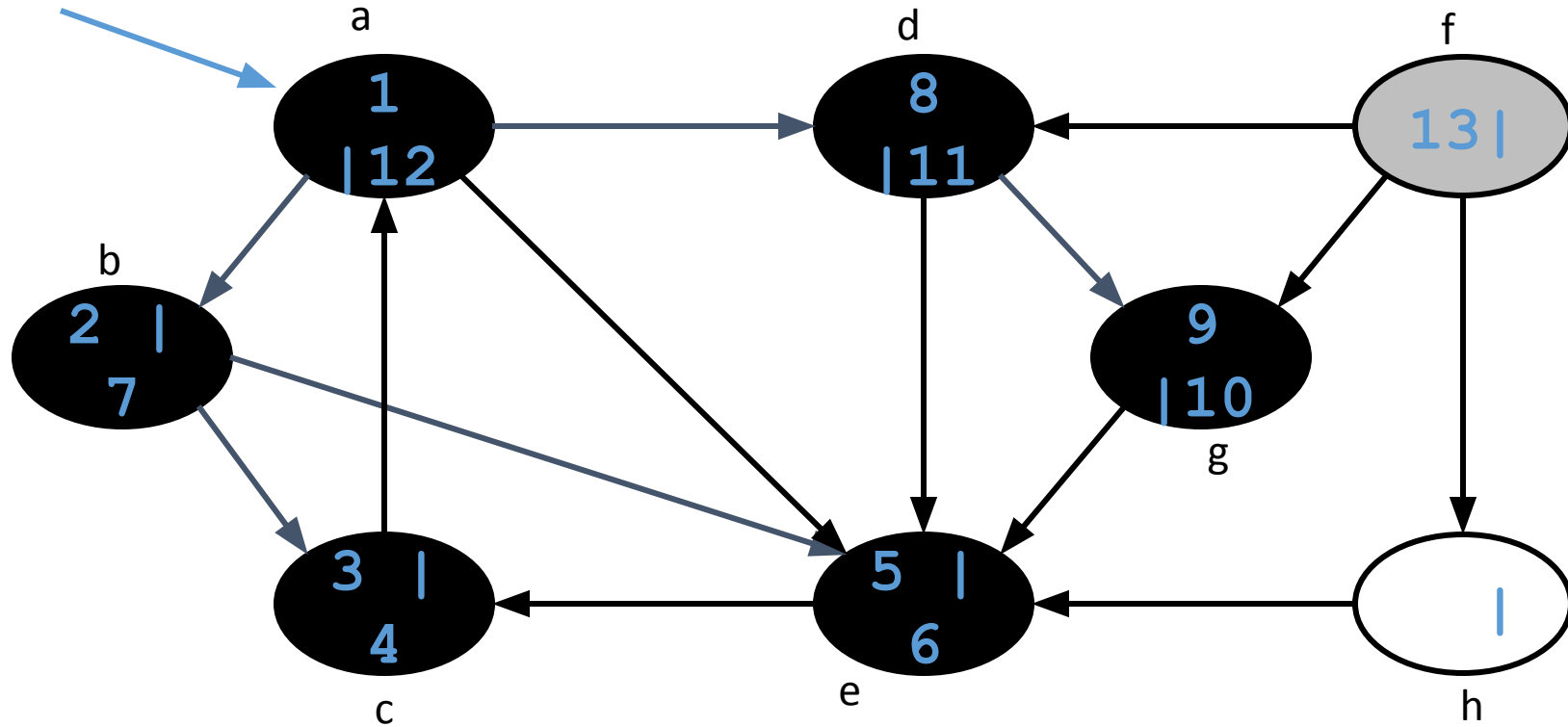
# DFS Example

*source  
vertex*



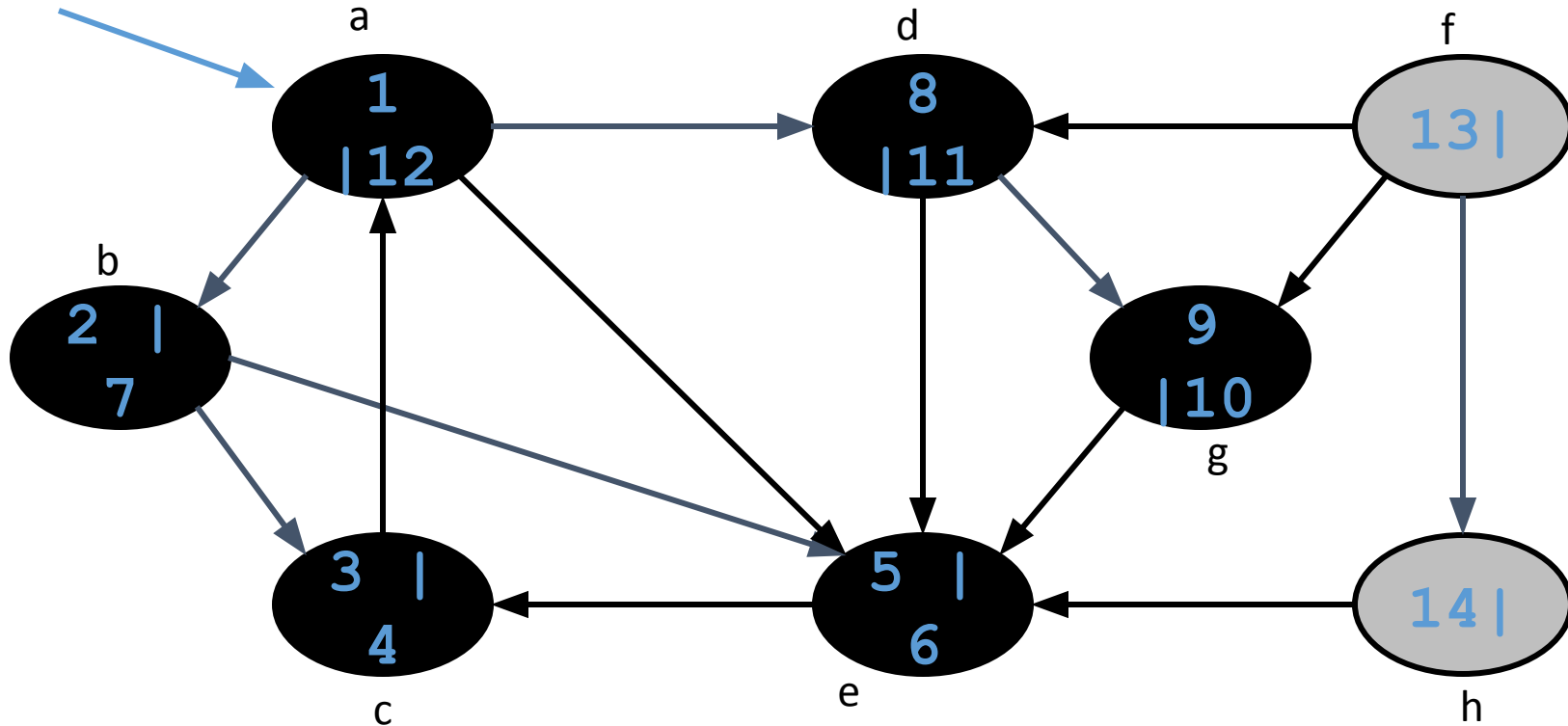
# DFS Example

*source  
vertex*



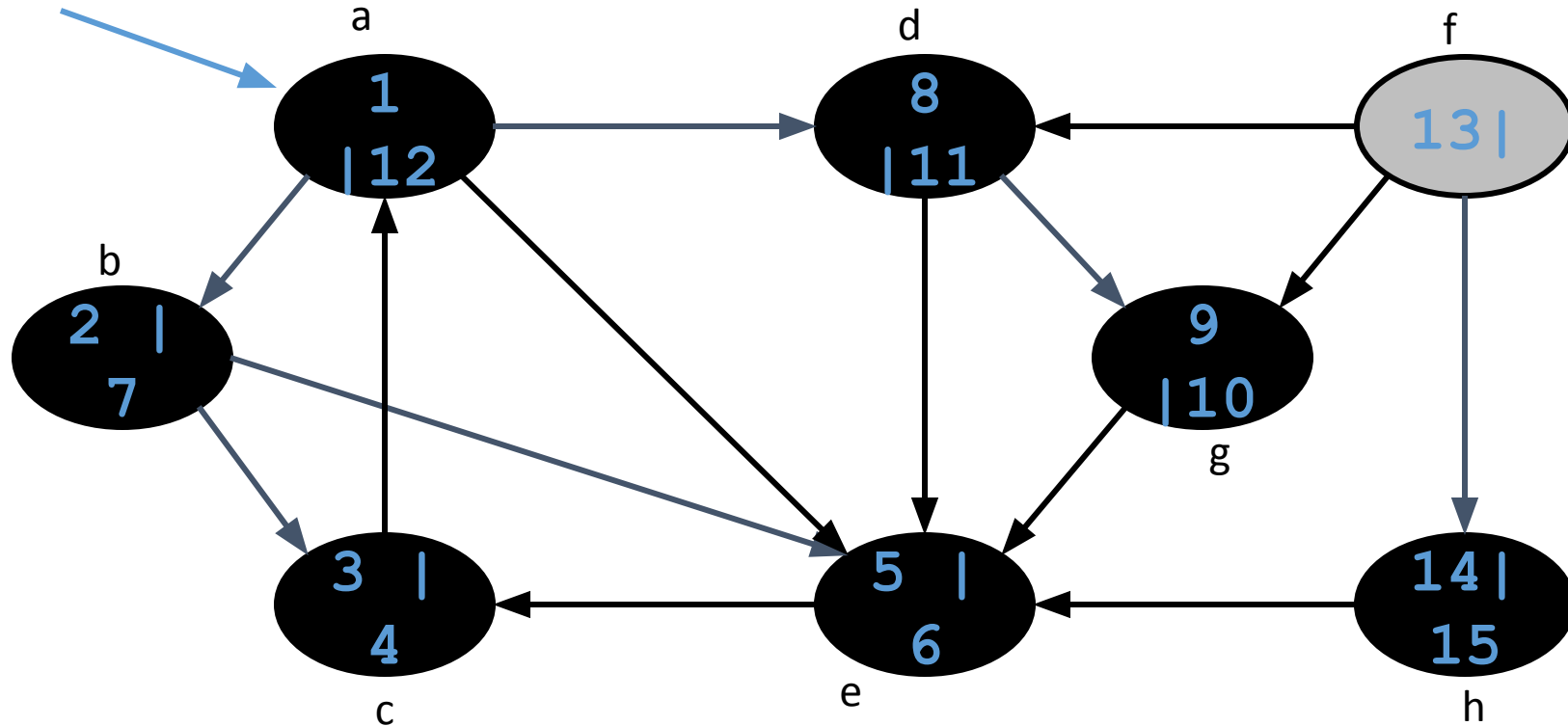
# DFS Example

*source  
vertex*



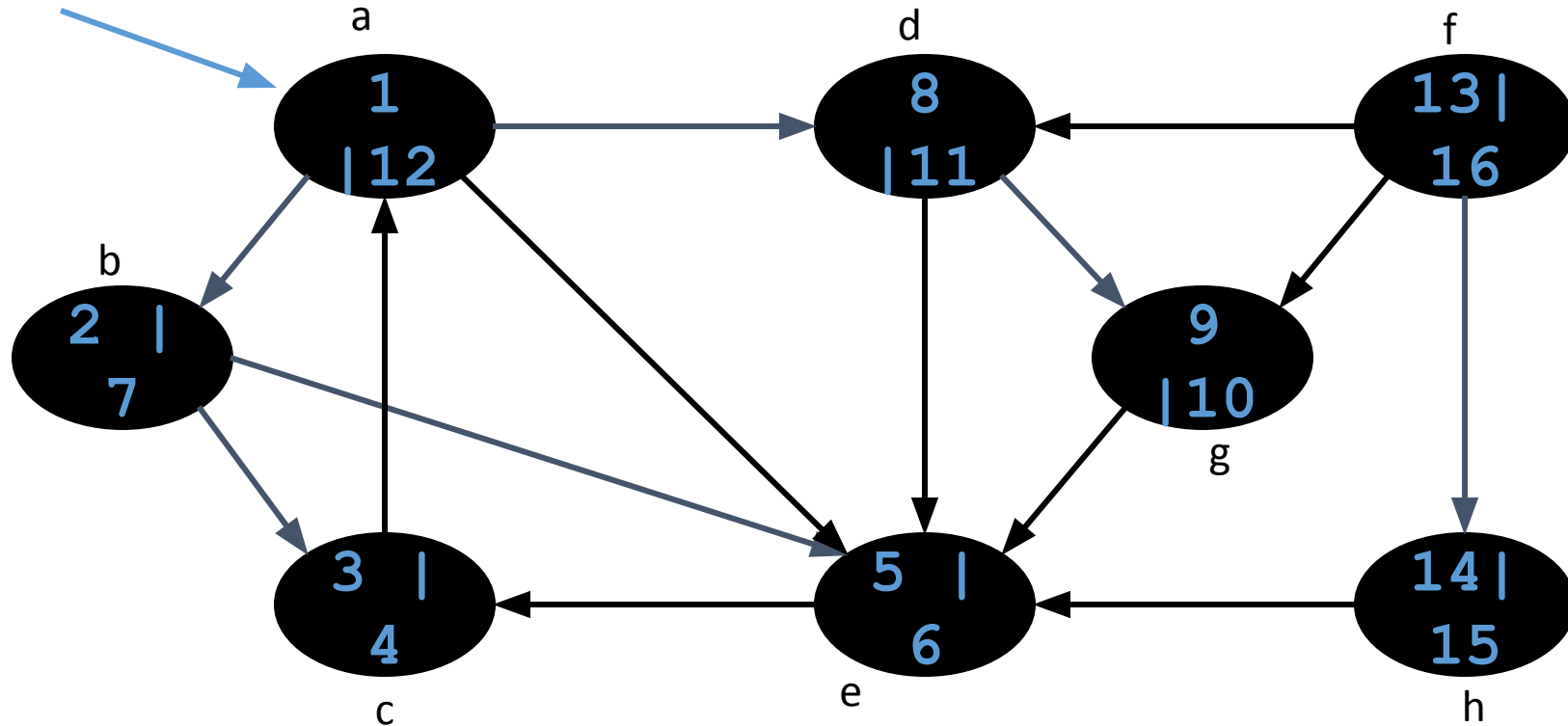
# DFS Example

*source  
vertex*



# DFS Example

*source*  
*vertex*



# Running Time of DFS

initialization takes  $O(V)$  time

second for loop in non-recursive wrapper considers each vertex, so  $O(V)$  iterations

one recursive call is made for each vertex

in recursive call for vertex  $u$ , all its neighbors are checked;  
total time in all recursive calls is  $O(E)$

Total time is  $O(V+E)$