

# Paths and connectivity

pseudocode: \* what order to consider the edges?  
\* obscure way to implement things

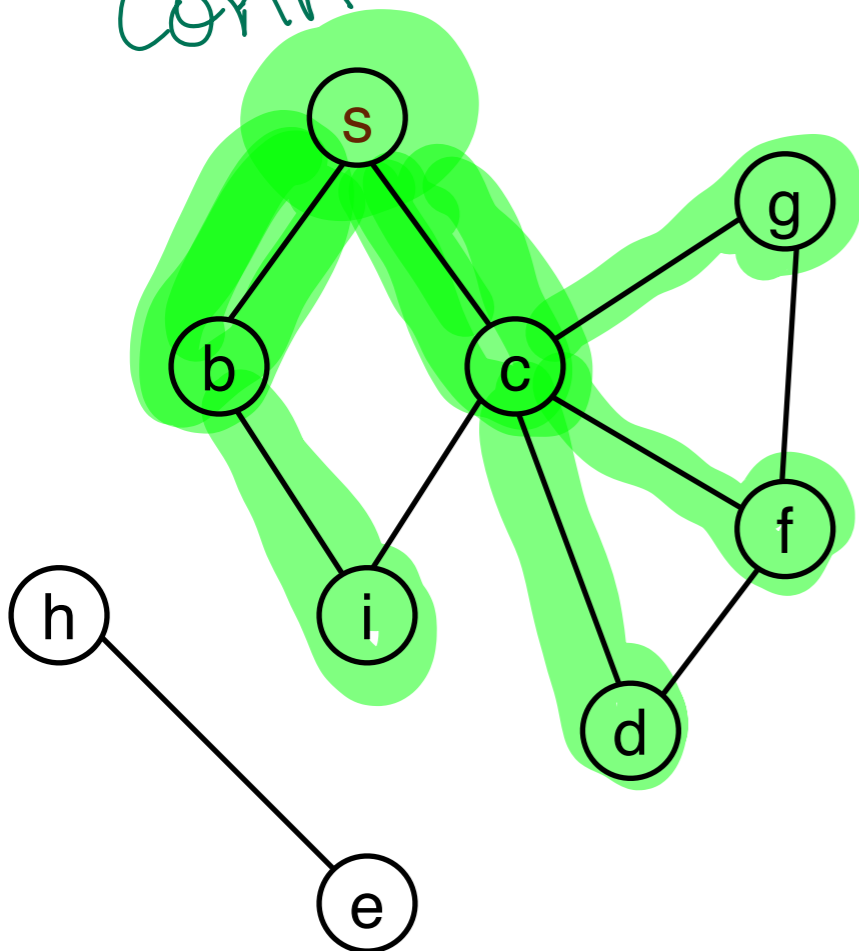
Question. Is there a path from  $s$  to  $e$ ?

Def. An undirected graph is **connected**, if there is a path between any pair of nodes.

Task: Given a source node  $s$ , find all nodes connected to  $s$ .

input: a graph  
a vertex

conn



---

**Algorithm 1:** GraphSearch( $G(V, E), s$ )

---

```
/* G is an undirected graph, s a source node
1 conn ← {s} /* set of nodes connected to s
2 while there is edge (u, v) with u in conn and v not in conn do
3   | add v to conn;
4 return conn
```

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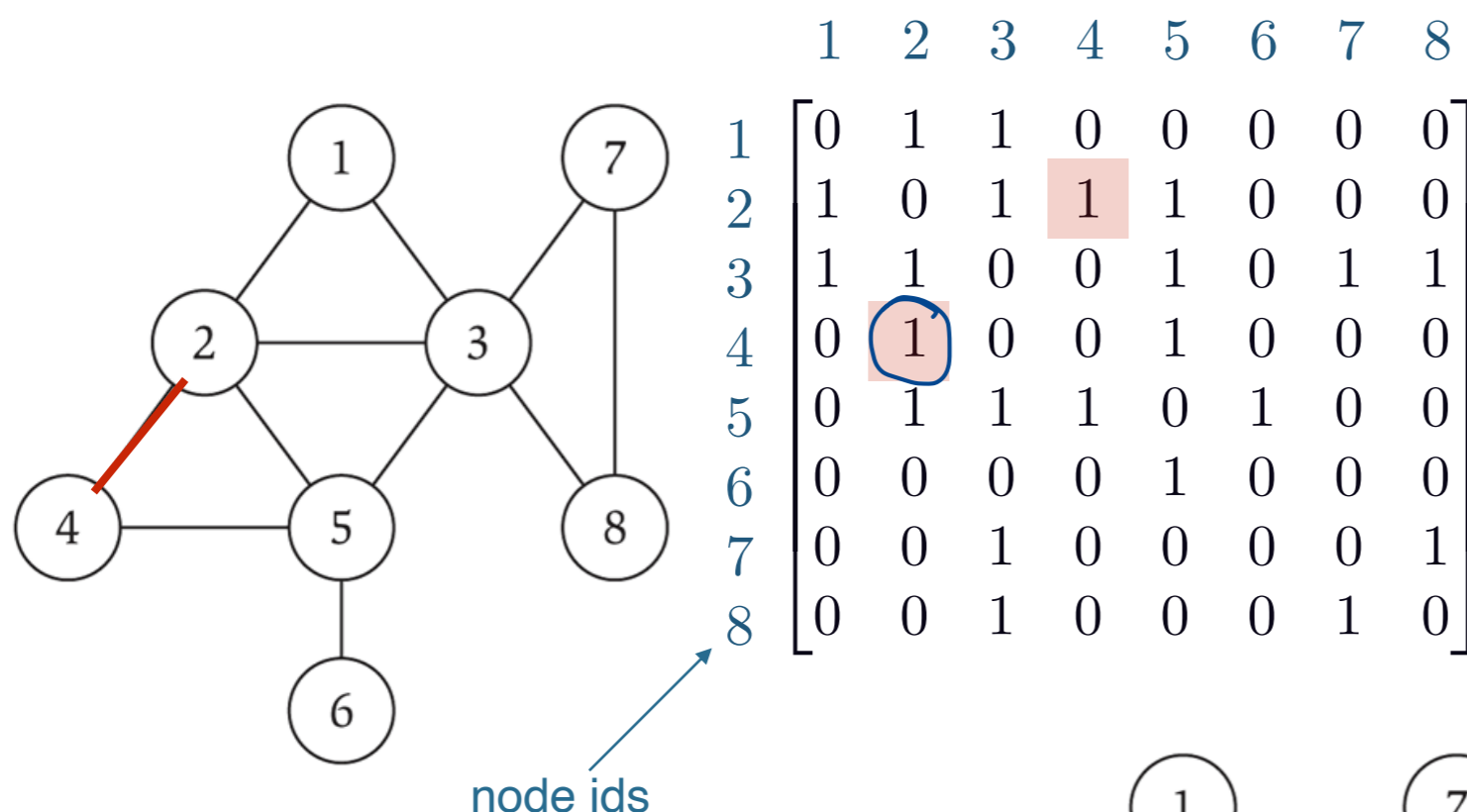
→ output: set of vertices

$s - c - d$  (simple path)

$s - c - g - f - c - d$   
(not simple path)

# Graph representation: adjacency matrix

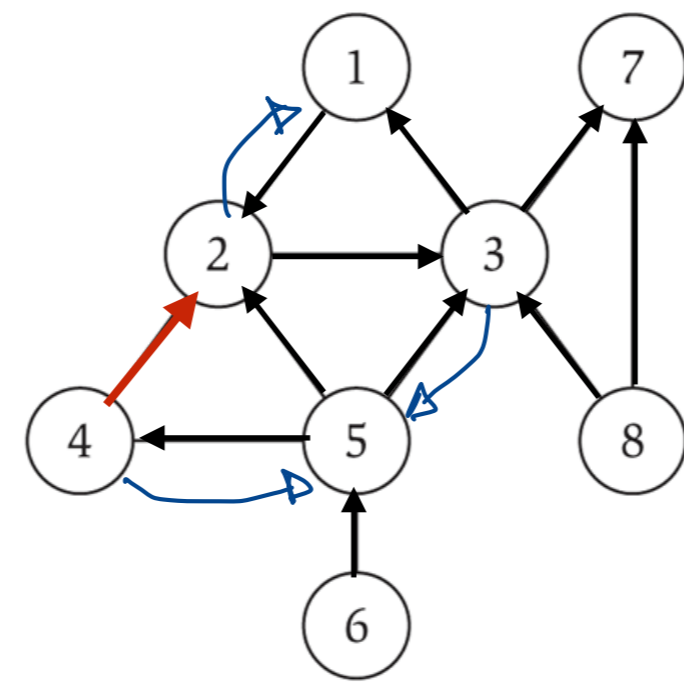
Adjacency matrix:  $n \times n$  binary matrix  $A$ , such that  $A[i,j]=1$  iff  $(i,j)$  is an edge



*symmetric*

undirected graph

*There is no edge connecting 2 to 4*  
directed graph

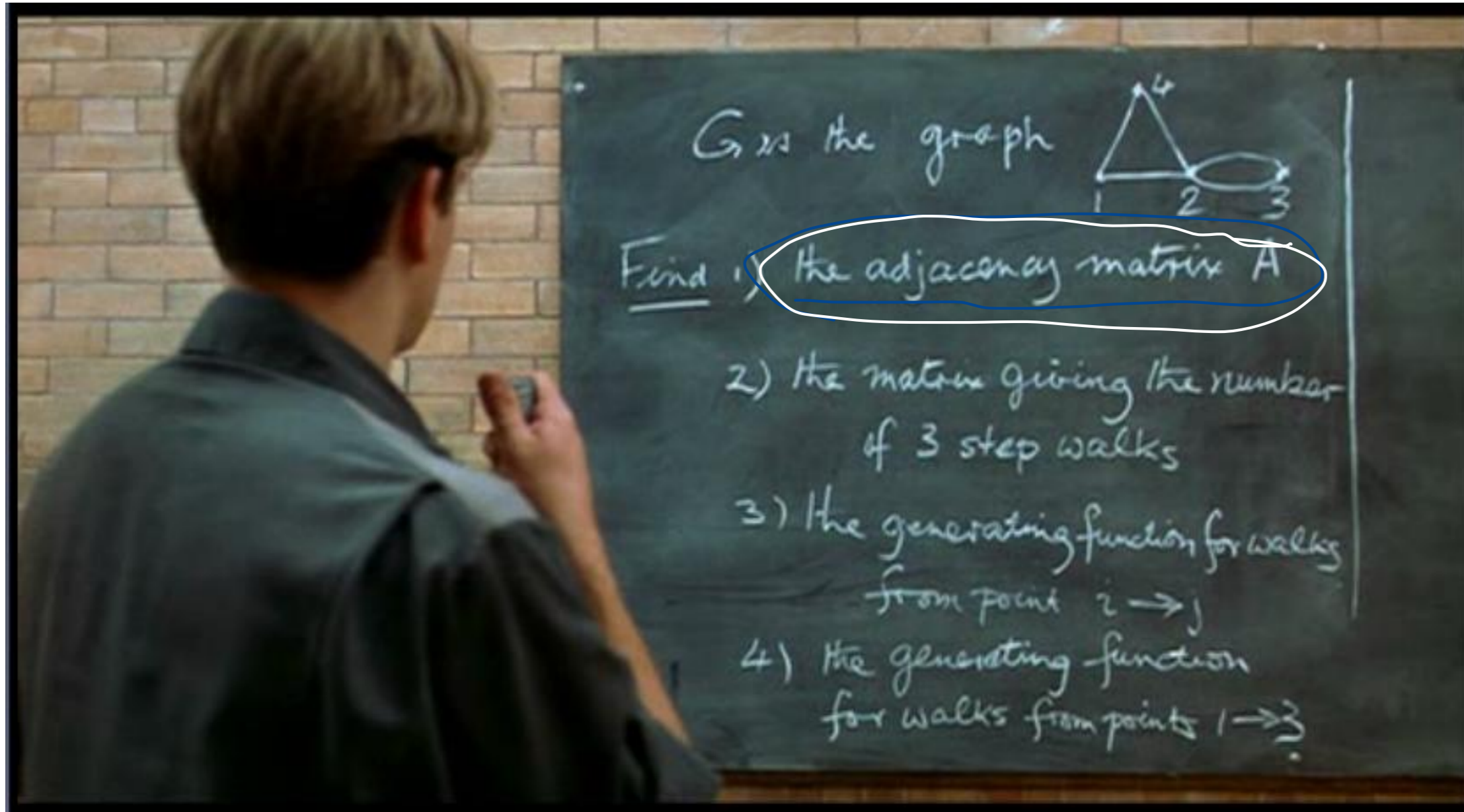


	1	2	3	4	5	6	7	8
1	0	1	0	0	0	0	0	0
2	0	0	1	0	0	0	0	0
3	1	0	0	0	0	0	1	0
4	0	1	0	0	0	0	0	0
5	0	1	1	1	0	0	0	0
6	0	0	0	0	1	0	0	0
7	0	0	0	0	0	0	0	0
8	0	0	1	0	0	0	1	0

*asymmetric*

directed adjacency matrix:  
rows: node of origin of edge  $e$   
cols: destination node

## Adjacency matrix in Good Will Hunting.

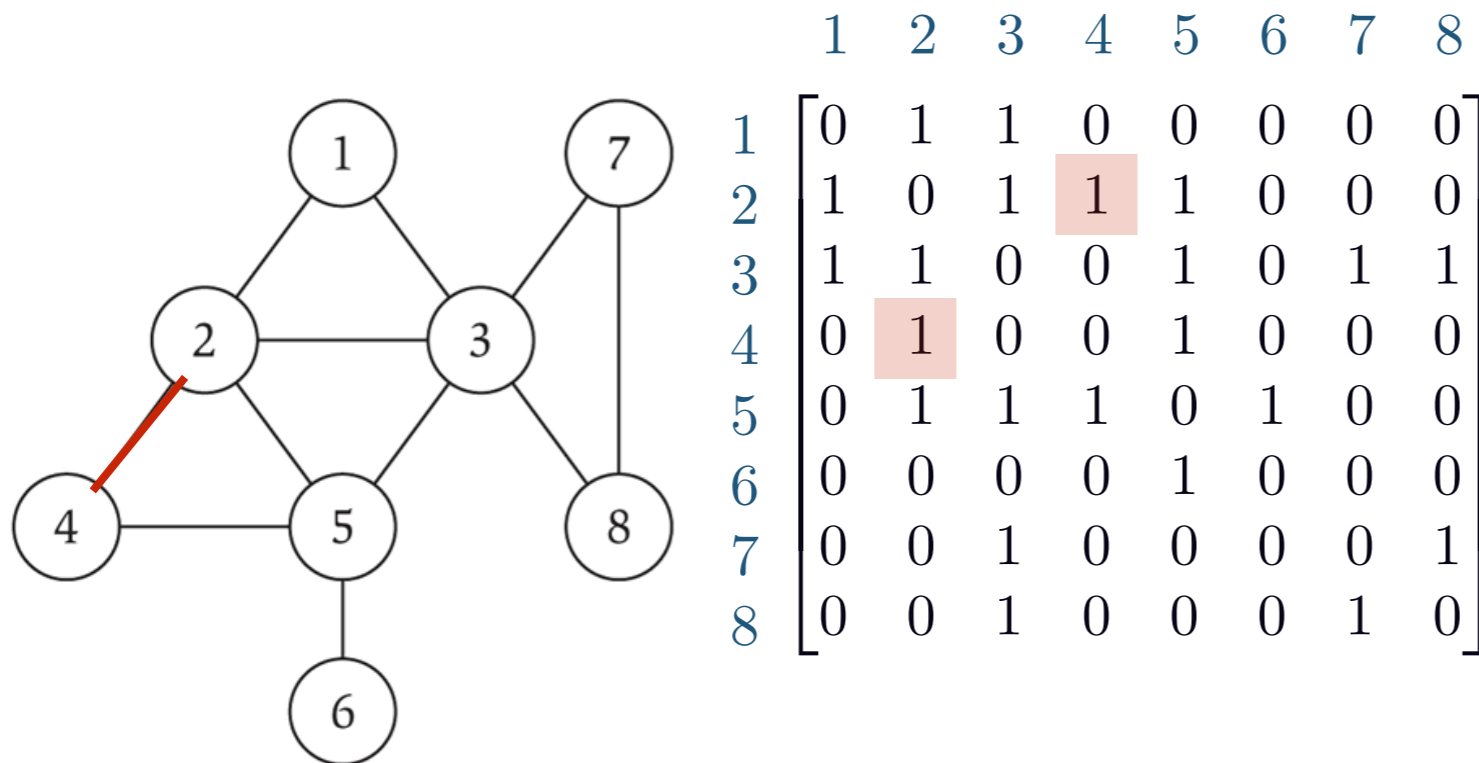


<https://www.imdb.com/title/tt0119217/>

<https://graph-theory.blogspot.com/2008/11/good-will-hunting-problem.html>

# Operations on an adjacency matrix

Adjacency matrix:  $n \times n$  binary matrix  $A$ , such that  $A[i,j]=1$  iff  $(i,j)$  is an edge



How do we perform a lookup? e.g. check whether  $(u,v)$  is an edge

Check the value of the matrix  $A[u,v]$  at the intersection of row  $u$  and column  $v$ .

How many operations is that?

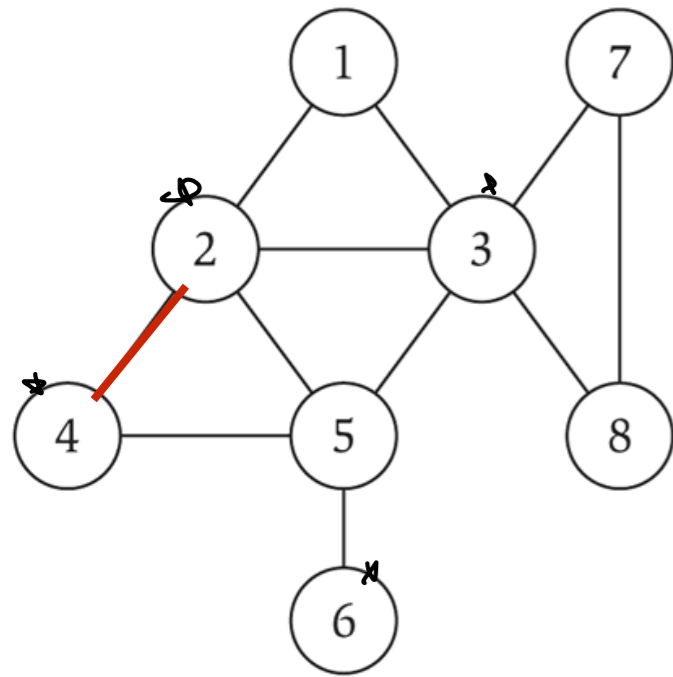
$\Theta(1)$  = constant time

if  $A[u,v] = 1 \rightarrow$  connected

if  $A[u,v] = 0 \rightarrow$  disconnected

# Operations on an adjacency matrix – TopHat

Adjacency matrix:  $n \times n$  binary matrix  $A$ , such that  $A[i,j]=1$  iff  $(i,j)$  is an edge



	1	2	3	4	5	6	7	8
1	0	1	1	0	0	0	0	0
2	1	0	1	1	1	0	0	0
3	1	1	0	0	1	0	1	1
4	0	1	0	0	1	0	0	0
5	0	1	1	1	0	1	0	0
6	0	0	0	0	1	0	0	0
7	0	0	1	0	0	0	0	1
8	0	0	1	0	0	0	1	0

Question 1.: (multiple choice!)

Complexity of listing all neighbors of vertex  $v$ ?

A.  $\Theta(1)$

B.  $\Theta(n)$

C.  $\Theta(n^2)$

Complexity of listing every edge in  $G$ ?

D.  $\Theta(n)$

E.  $\Theta(n^2)$

F.  $\Theta(m)$

# Operations on an adjacency matrix – directed

How would you implement the following operations and what is their complexity?

- Lookup: verify whether the pair  $(u,v)$  form a directed edge

*Check if the intersection of row  $u$  and column  $v$  is 1*

- Find the out-neighbors of  $v$ , or find the out-degree of  $v$ . (degree = number of edges directed away from  $v$ )

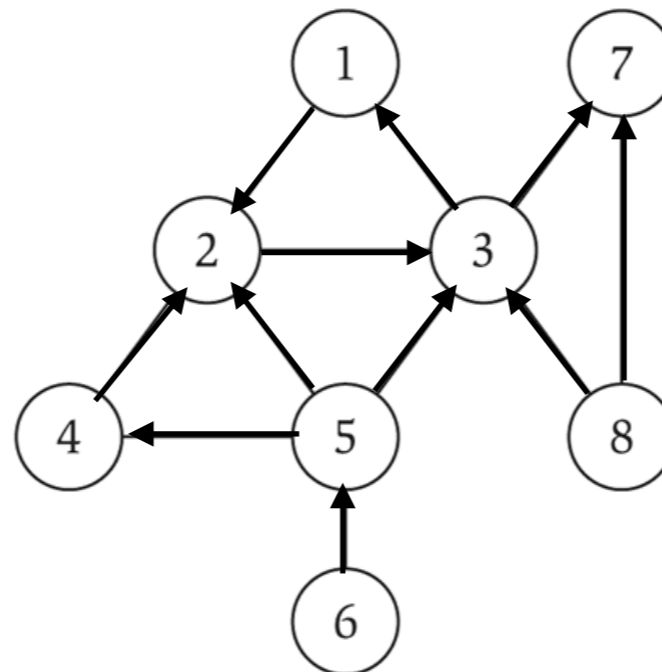
*Count the amount of 1 in the row  $v$*

- Find the in-neighbors of  $v$

*Count the amount of 1's in the column  $v$*

- List every directed edge in  $G$

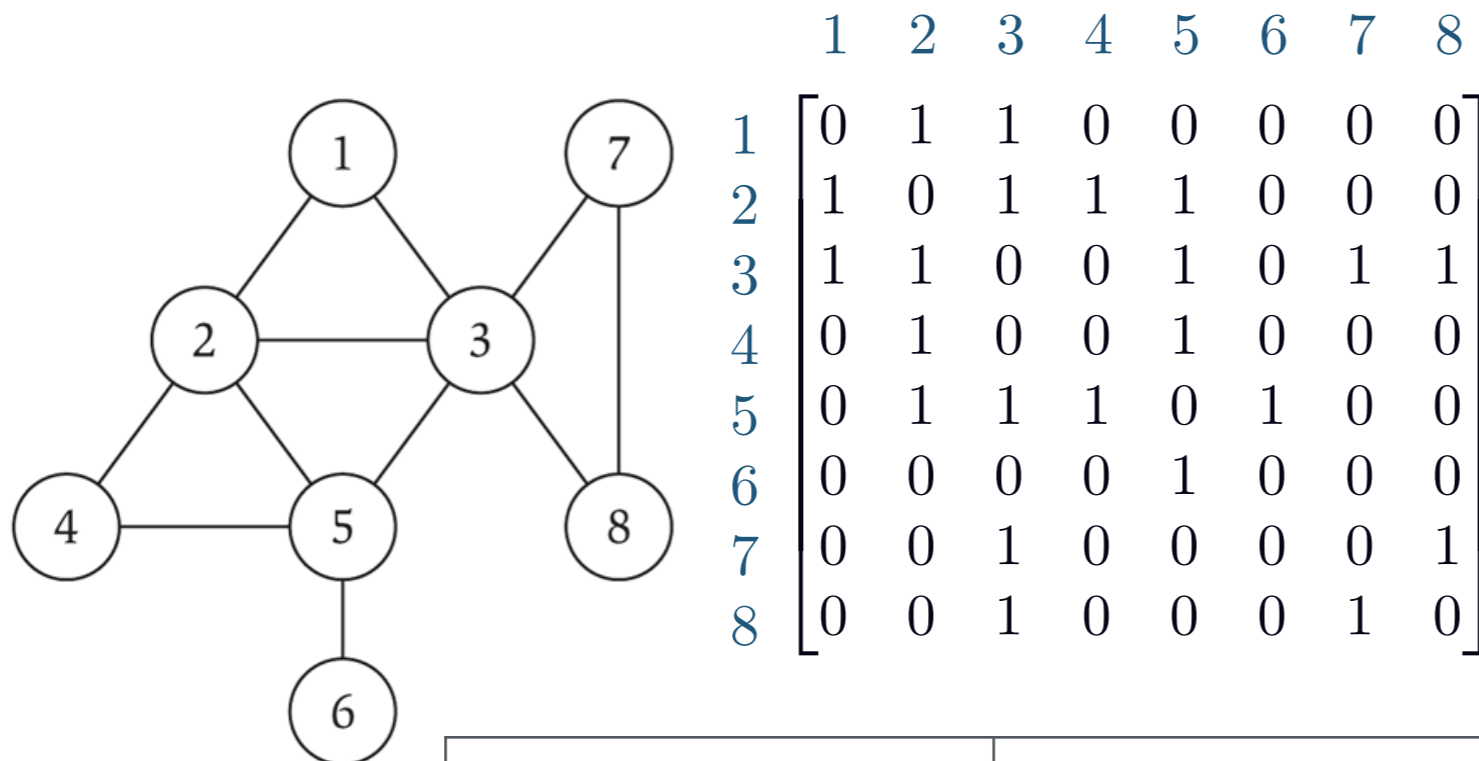
*count = 0  
for  $i=1$  to  $n$  do:  
  for  $j=1$  to  $n$  do:  
    if  $A[i,j] == 1$  then:  
      count += 1  
return count*



	1	2	3	4	5	6	7	8
1	0	1	0	0	0	0	0	0
2	0	0	1	0	0	0	0	0
3	1	0	0	0	0	0	1	0
4	0	1	0	0	0	0	0	0
5	0	1	1	1	0	0	0	0
6	0	0	0	0	1	0	0	0
7	0	0	0	0	0	0	0	0
8	0	0	1	0	0	0	1	0

# Graph representation: adjacency matrix

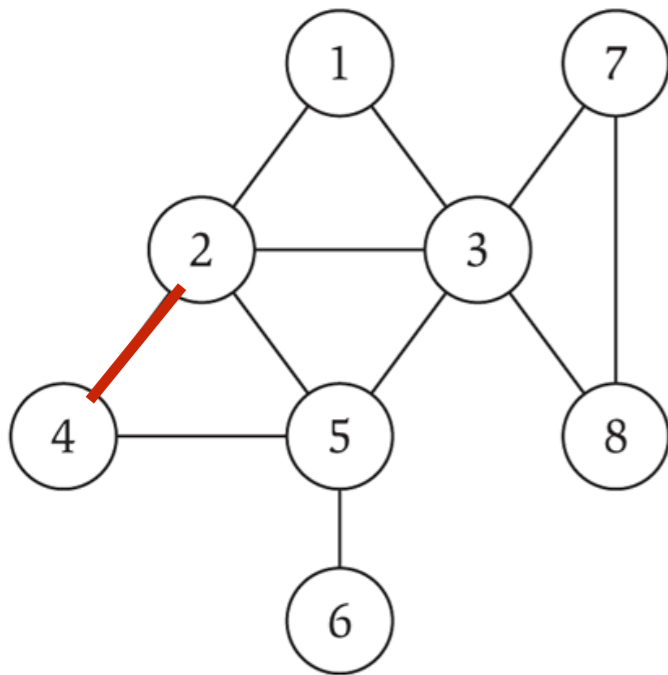
Adjacency matrix:  $n \times n$  binary matrix  $A$ , such that  $A[i,j]=1$  iff  $(i,j)$  is an edge



<b>space complexity</b>	$\Theta(n^2)$
<b>Operation</b>	<b>time complexity</b>
check edge $(u,v)$	$\Theta(1)$
list all neighbors of $u$	$\Theta(n)$
list all edges	$\Theta(n^2)$

# Graph representation: adjacency lists

Adjacency list: For each node  $v$  there is a record listing the nodes to which  $v$  is connected. *↳ Uses less space than an adjacency matrix because we only store connected neighbors*



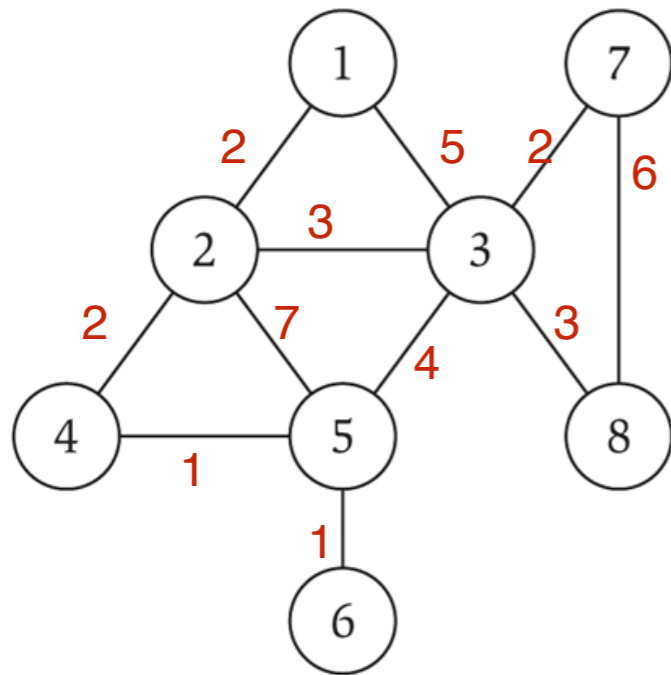
NodeId	Neighbors
1	2      3
2	1      4      5      3
3	2      1      5      7      8
4	2      5
5	2      4      3      6
6	5
7	3      8
8	3      7

Note: vertices in an adjacency list need not appear in any particular order.

*↳ doesn't need to*

# Graph representation: adjacency list – weighted graphs

**Adjacency list:** For each node  $v$  there is a record listing the (node, weight) pairs to which  $v$  is connected.



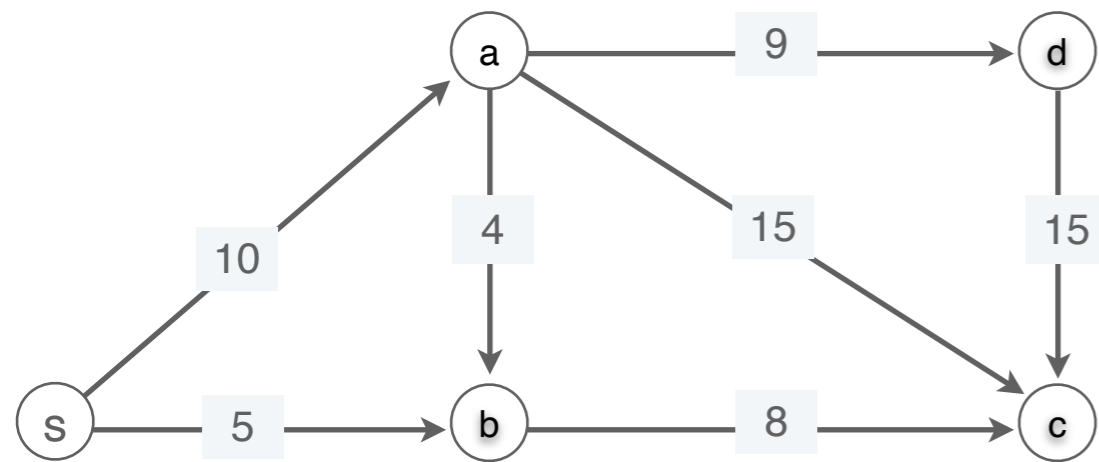
NodeId	Neighbors				
1	2   2	3   5			
2	1   2	4   2	5   7	3   3	
3	2   3	1   5	5   4	7   2	8   3
4	2   2	5   1			
8	3   3	7   6			
7	3   2	8   7			
5	2   7	4   1	3   4	6   1	
6	5   1				

neighbor id      edge weight

If  $(u,v)$  is an edge with weight  $w$ , then we store  $(v,w)$  instead of  $(v)$  in  $u$ 's neighbor list

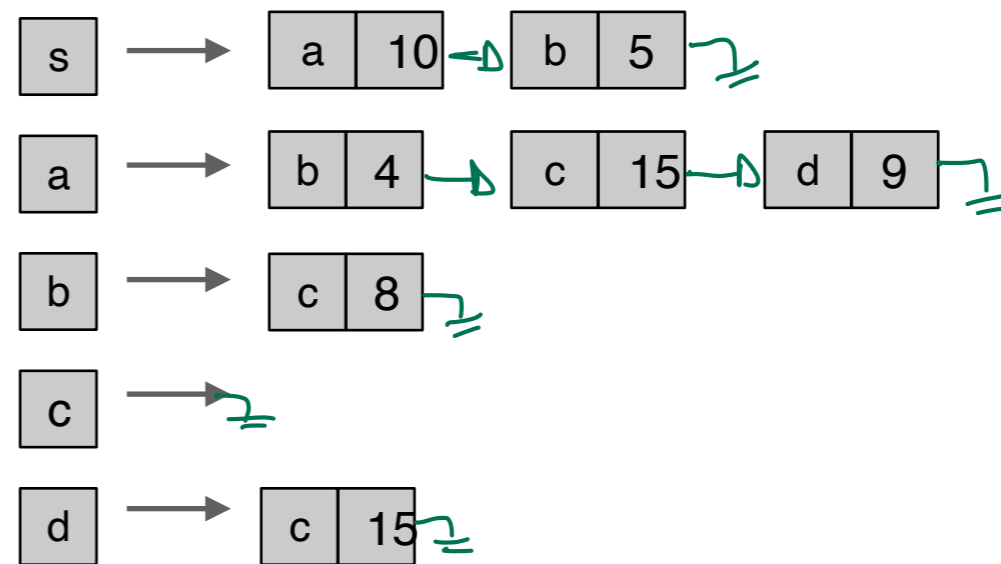
# Graph representation: adjacency list

Adjacency list: For each node  $v$  there is a record listing the nodes *towards* which  $v$  has a directed edge.

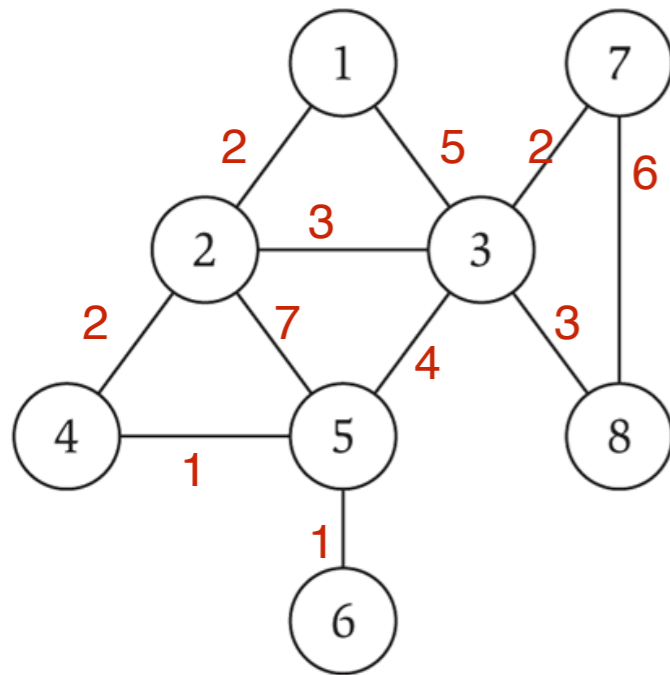


NodeId

Neighbors



# TopHat



Node id	Neighbors
1	2   2    3   5
2	1   2    3   3    4   2    5   7
3	1   5    2   3    4   2    5   4    7   2    8   3
4	2   2    3   3    5   1
5	2   7    3   4    4   1    6   1
6	5   1
7	1   2    2   3    3   3    4   2    5   4    6   1
8	3   3    4   2    5   4    6   1

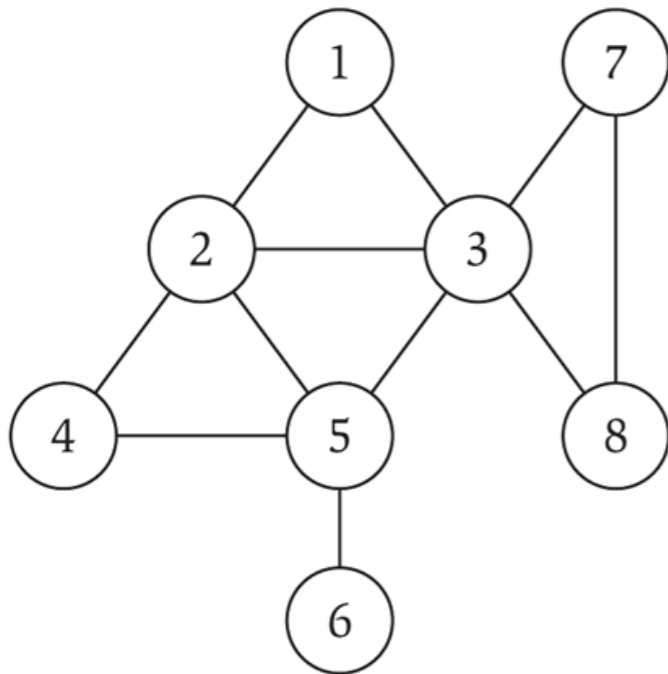
neighbor id      edge weight

Question:

Let A be an adjacency list. What is the total number of keys stored in A?

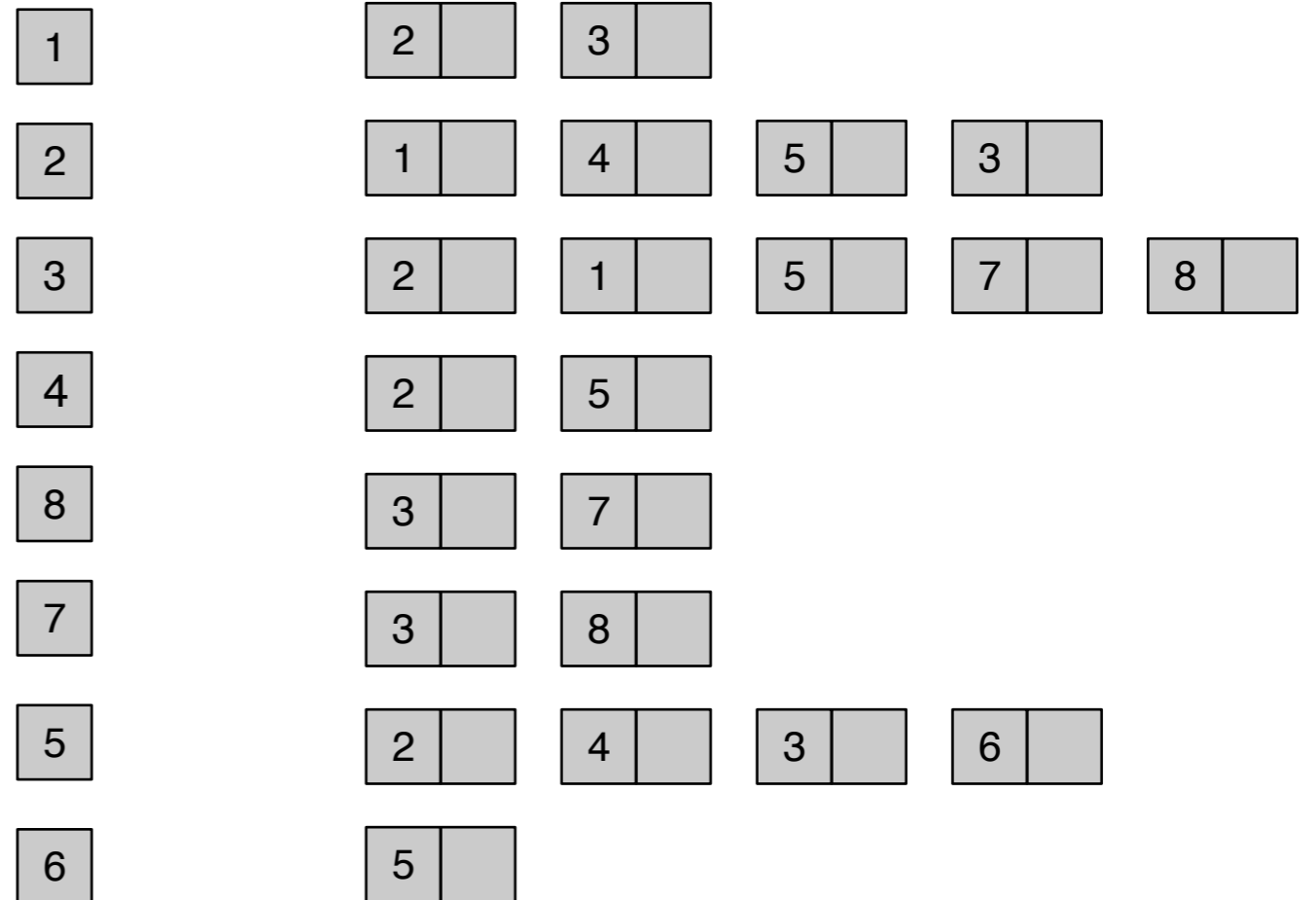
- A.  $\Theta(|V|)$
- B.  $\Theta(|V|^2)$
- C.  $\Theta(|E|)$
- D.  $\Theta(\max\{degree(v)\})$
- E.  $\Theta(|V| + |E|)$

# Graph representation: adjacency lists



NodeId

Neighbors



space complexity	$\Theta( V  +  E )$
Operation	time complexity
check edge (u,v)	$\Theta(\delta(u))$
find all neighbors of u	$\Theta(\delta(u))$
find all edges	$\Theta( V  +  E )$

## Adjacency list – space complexity analysis

# of nodes  $\rightarrow$  # of edges

$$|V| = n$$

$$|E| = m$$

Claim. The size of the adjacency list is  $\Theta(n + m)$

Note. using  $n+m$  is the preferred notation.  $\Theta(n + m) = \Theta(\max\{n, m\})$

$\Theta(n)$  = the size of list containing nodes id  
 $\Theta(m)$  = each edge in this data structure is represented as a node on the right side of the list of nodes. (counted twice)

We need at least  $\Theta(n)$  to represent the nodes

We need at least  $\Theta(m)$  to represent the edges

Thus, we need  $\Theta(\max\{n, m\})$  to represent

the adjacency list  $\Rightarrow \Theta(\max\{n, m\}) = \Theta(n + m)$

# Adjacency list – space complexity analysis

**Claim.** The size of the adjacency list is  $\Theta(n + m)$

**Note.** using  $n+m$  is the preferred notation.  $\Theta(n + m) = \Theta(\max\{n, m\})$

proof 1:

$\Theta(n)$  : there is a slot in the main list for every vertex  $v$  even if their degree is 0

$\Theta(m)$  : *there is a list of size  $\delta(v)$  for every  $v$  in the main list*

- For each  $v$  the size of the list containing  $v$ 's neighbors is  $\text{degree}(v)$
- We know that each edge  $(v,w)$  contributes once to the degree of  $v$  and once to  $w$

$$\sum_{v \in V} \text{degree}(v) = 2m = \Theta(m)$$

proof 2:

$\Theta(n)$  : there is a slot in the main list for every vertex  $v$  even if their degree is 0

$\Theta(m)$  : *every neighbor of  $v$  is represented on the right side of,*

- Each edge  $(v,w)$  is represented in the table twice;  $w$  is in the neighbor list of  $v$  and vice versa.

*↳ the main list*

**Exercise.** Repeat the same proof on directed graphs and the directed adj list.

## $n^2$ vs. $m$ in graphs

Why is the use of adjacency lists, with  $\Theta(n + m)$  complexity considered better than the  $\Theta(n^2)$  of the adjacency matrix?

Intuitively most of the time a graph contains more edges than vertices. So, why use  $\Theta(n + m)$  instead  $\Theta(n^2)$ ?

$$\Theta(n^2) = \frac{n(n-1)}{2} \leq c \cdot n^2 \quad (\text{Because every edge is counted twice})$$

What is the minimum amount of edges in a connected graph?

$$n-1 \leq m \leq n^2$$

Exercise. Compute the minimum number of edges in a connected graph on  $n$  vertices.

# Computing the running time in loops

A is the adjacency matrix of a graph, G is the adjacency list of the same graph.

---

## Algorithm 1: MatrixCount( $A$ )

---

```
1  $count \leftarrow 0$ ;  
2 for  $i = 1$  to  $n$  do  
3   | for  $j = 1$  to  $n$  do  
4   | | if  $A[i, j] == 1$  then  
5   | | |  $count+ = 1$ ;  
6 return  $count/2$ 
```

---

---

## Algorithm 2: ListCount( $G$ )

---

```
1  $count \leftarrow 0$ ;  
2 for  $v$  in  $G$  do  
3   | for  $w$  in  $G[v]$  do  
4   | |  $count+ = 1$ ;  
5 return  $count/2$ 
```

---

Same algorithm, but using different data structures.

Both return the number of edges

# Computing the running time in loops - TopHat

A is the adjacency matrix of a graph, G is the adjacency list of the same graph.

---

**Algorithm 1: MatrixCount(A)**

---

```
1 count ← 0;
2 for i = 1 to n do → n times
3   | for j = 1 to n do → n times
4   | | if A[i, j] == 1 then
5   | | | count+ = 1; } Θ(1)
6 return count/2
```

---

(multiple choice!)

What is the running time of MatrixCount?

A.  $\Theta(n)$

**B.  $\Theta(n^2)$**

What is the running time of ListCount?

**C.  $\Theta(n + m)$**

D.  $\Theta(nm)$

E.  $\Theta(n \cdot deg_{max})$

---

**Algorithm 2: ListCount(G)**

---

```
1 count ← 0;
2 for v in G do → n times
3   | for w in G[v] do
4   | | count+ = 1;
5 return count/2
```

---

the length of the loop depends on the amount of neighbors of v

# Computing the running time in loops

A is the adjacency matrix of a graph, G is the adjacency list of the same graph.

---

## Algorithm 1: MatrixCount( $A$ )

---

```
1  $count \leftarrow 0$ ;  
2 for  $i = 1$  to  $n$  do  
3   |   for  $j = 1$  to  $n$  do  
4   |   |   if  $A[i, j] == 1$  then  
5   |   |   |    $count+ = 1$ ;  
6 return  $count/2$ 
```

---

If the number of iterations in the inner loop is *fixed*, then we can multiply to get the number of iterations.:

$n \times n$  iterations,  $O(1)$  each results in  $O(n^2)$

---

## Algorithm 2: ListCount( $G$ )

---

```
1  $count \leftarrow 0$ ;  
2 for  $v$  in  $G$  do  
3   |   for  $w$  in  $G[v]$  do  
4   |   |    $count+ = 1$ ;  
5 return  $count/2$ 
```

---

If the number of iterations is *different* each time then compute the total number of operations across the iterations.:

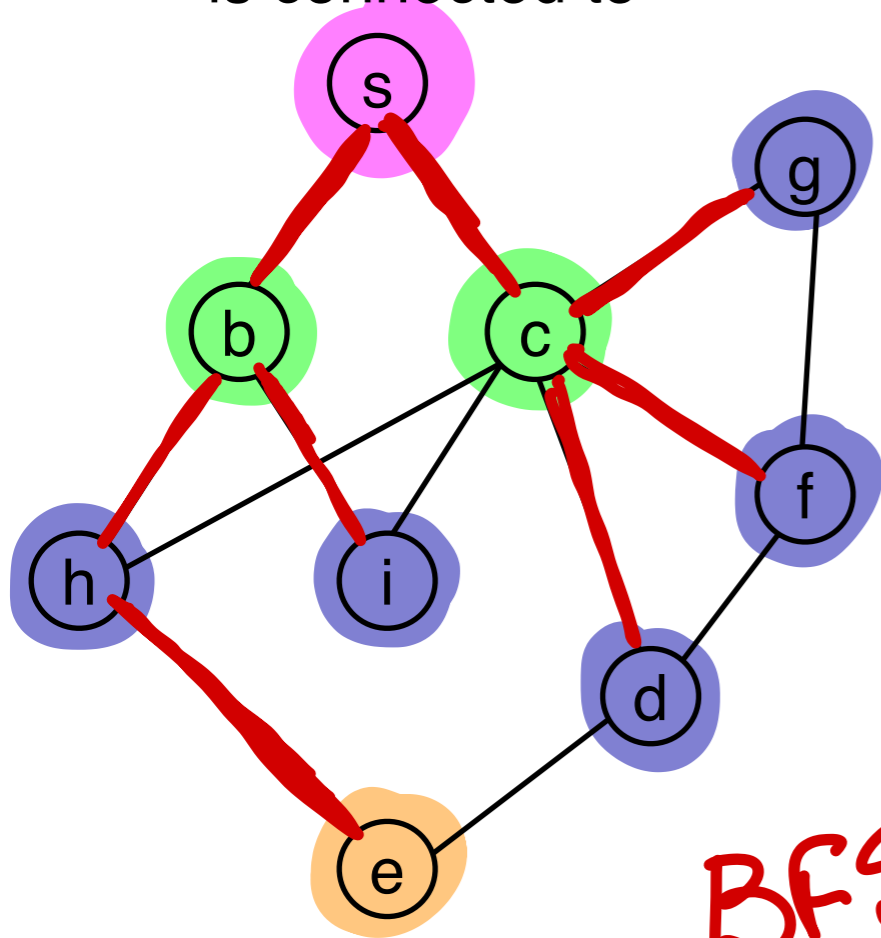
- the outer loop visits each node ones:  $O(n)$
- the inner loop iterates over each edge in  $G$ , the counter is increased once for each edge:  $O(m)$
- total:  $O(n+m)$

# Breadth First Search (BFS)



Breadth first search.

- Task: Given a source node **s**, find the *shortest* path from **s** to each node **v** that it is connected to



- 1 - explore directed neighbors of **s**
- 2 - explore all neighbors of **b** and **c**
- 3 - explore all neighbors of **h, i, d, f, and g**.

**BFS Tree** (represented by red edges)

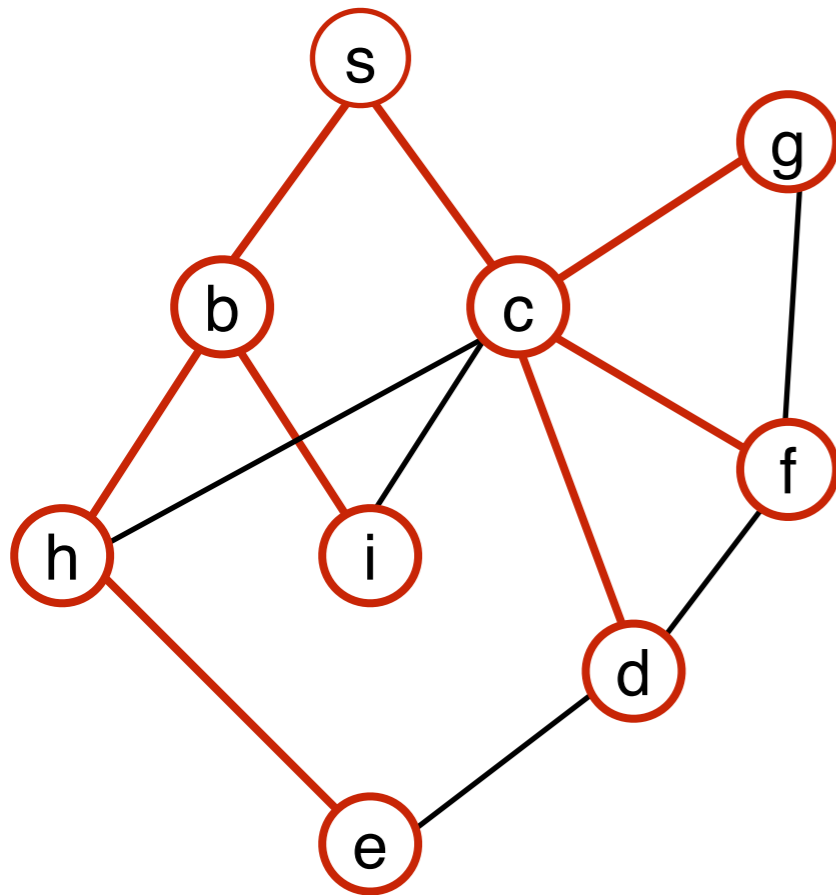
- Contains all nodes of the original graph
- has the minimum amount of edges

shortest path between nodes **u** and **v** = the connecting path with the least edges

# BFS

## Breadth first search.

- Task: Given a source node **s**, find the *shortest* path from **s** to each node **v** that it is connected to



First iteration:

start from s.

Second iteration:

find all neighbors of s. first layer

Third iteration:

find all (so far undiscovered) neighbors of nodes in first layer. second layer

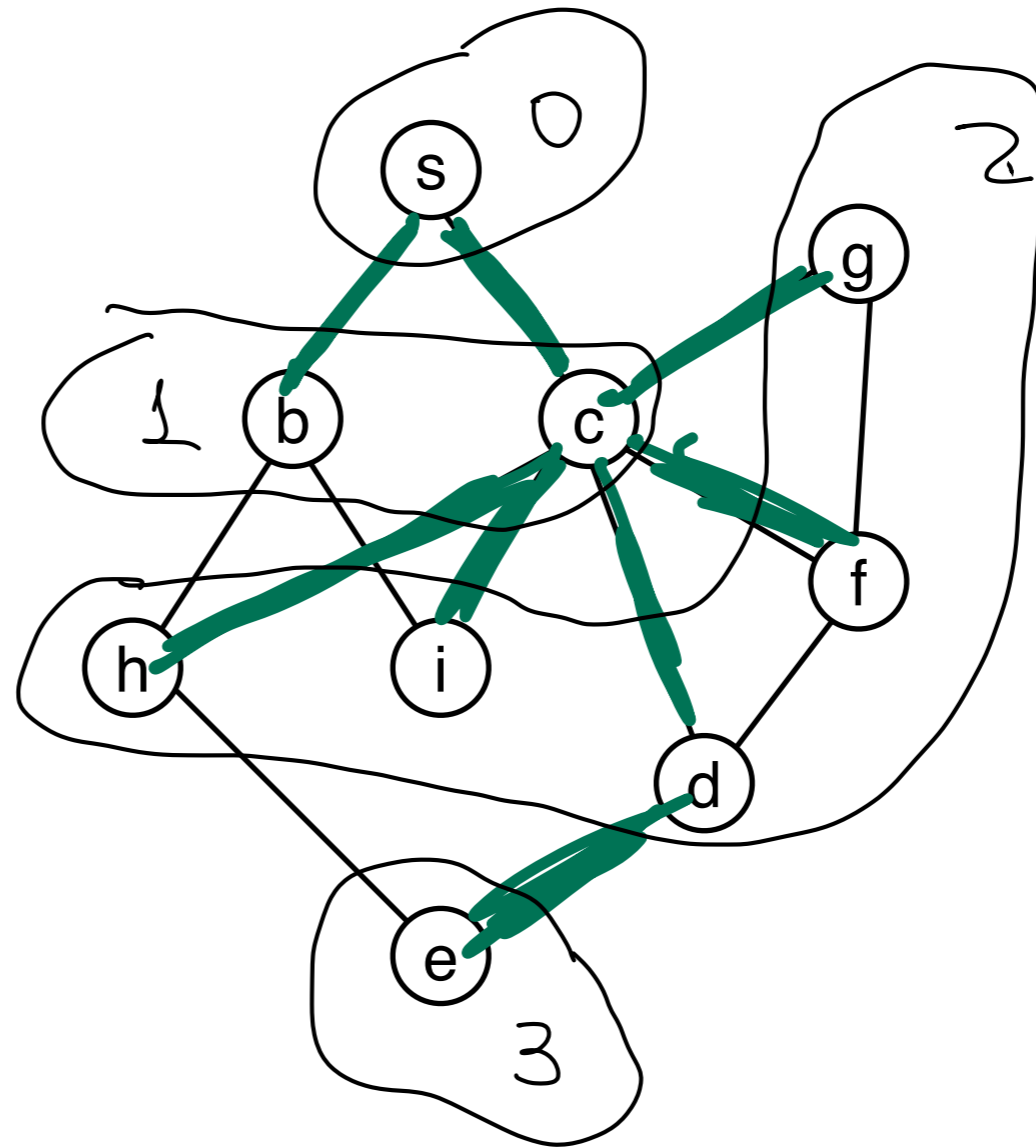
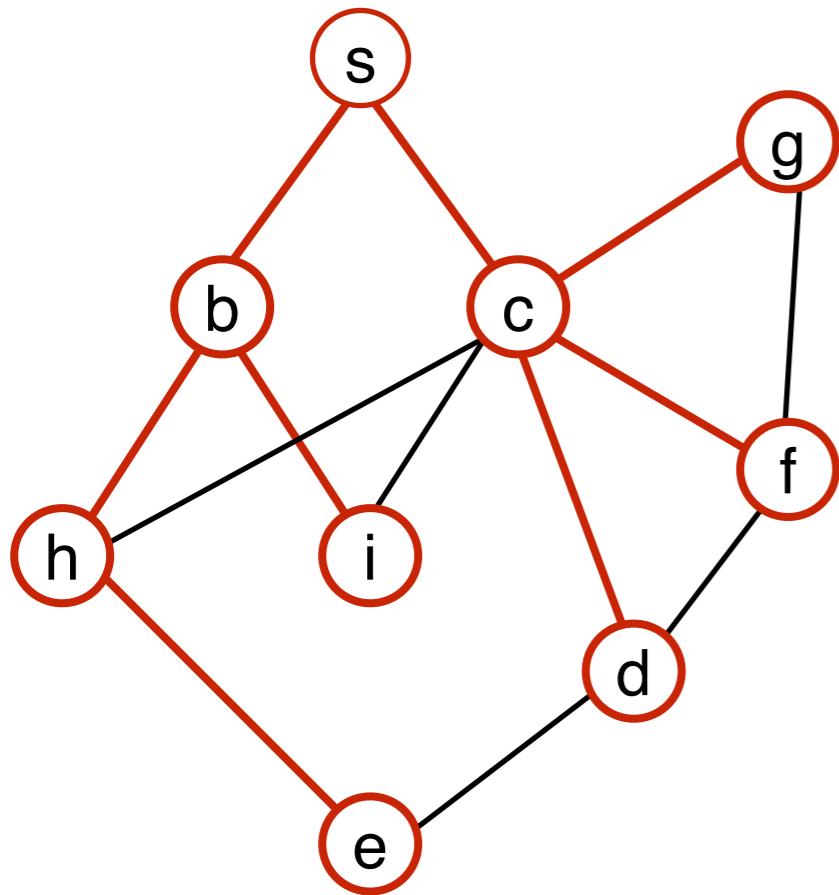
Fourth iteration:

find all neighbors of the second layer. third layer

BFS **search tree**: an edge connecting each node to the neighbor through which it was discovered.

# Breadth first search – order of node processing

Green: alternative order of node processing.



BFS search tree