

CSC263 Week 8

Announcements

Problem Set 4 is out!

Due Tuesday (Nov 17)



Other Announcements

→ Drop date Nov 8

→ Final exam schedule is posted

CSC263 exam Dec 11, 2-5pm

This week's outline

→Graphs

→BFS

Graph

A really, really important ADT that is used to model **relationships** between objects.

Get that job at Google

Whenever someone gives you a problem, *think graphs*. They are the most fundamental and flexible way of representing any kind of a relationship, so it's about a 50-50 shot that any interesting design problem has a **graph** involved in it. Make absolutely sure you can't think of a way to solve it using **graphs** before moving on to other solution types. This tip is important!

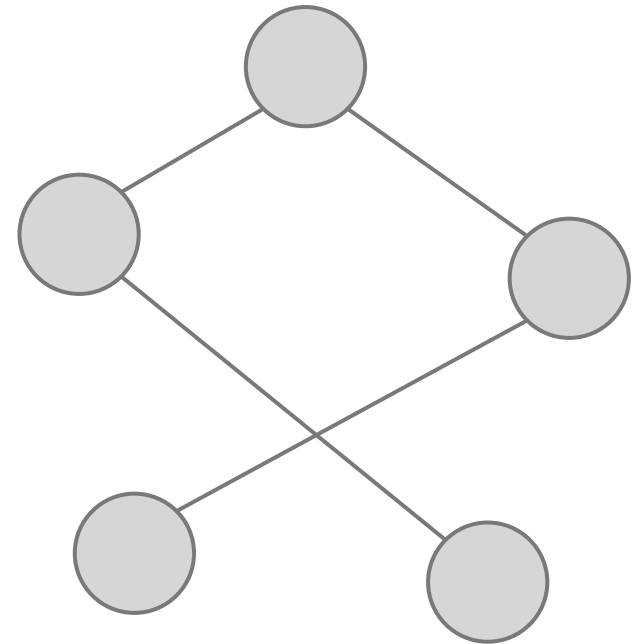
Reference: <http://steve-yegge.blogspot.ca/2008/03/get-that-job-at-google.html>

Things that can be modelled using graphs

- Web
- Facebook
- Task scheduling
- Maps & GPS
- Compiler (garbage collection)
- OCR (computer vision)
- Database
- Rubik's cube
- (many many other things)

Definition

$$G = (V, E)$$

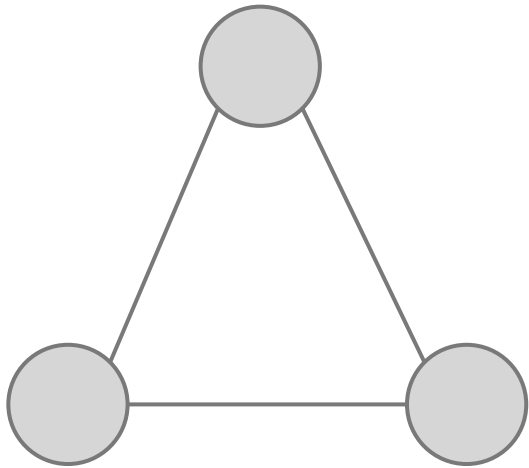


Set of **vertices**
e.g., {a, b, c}

Set of **edges**
e.g., { (a, b), (c, a) }

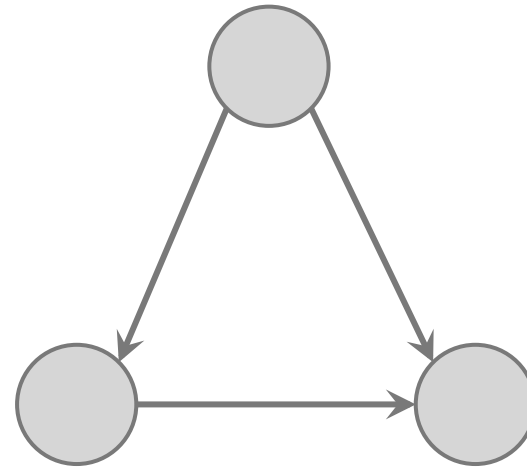
Flavours of graphs

each edge is an
unordered pair
 $(u, v) = (v, u)$

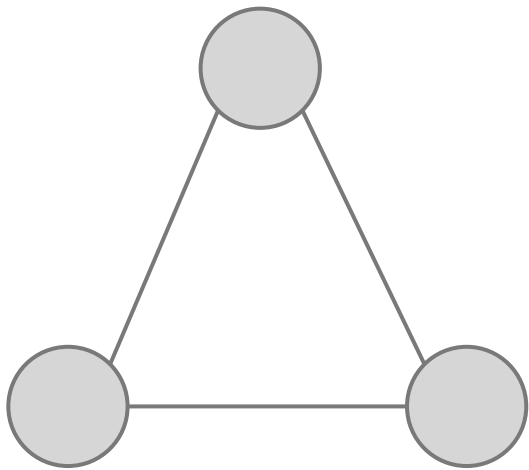


Undirected

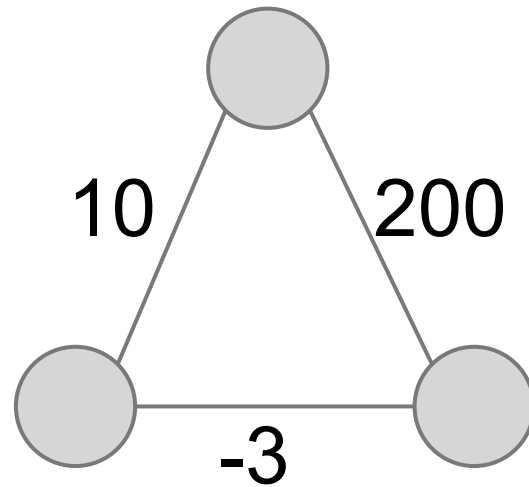
each edge is an
ordered pair
 $(u, v) \neq (v, u)$



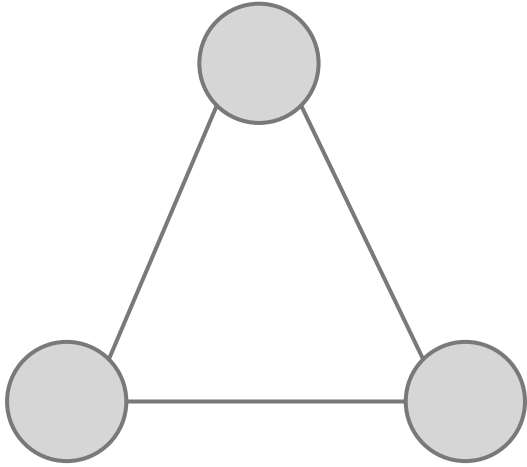
Directed



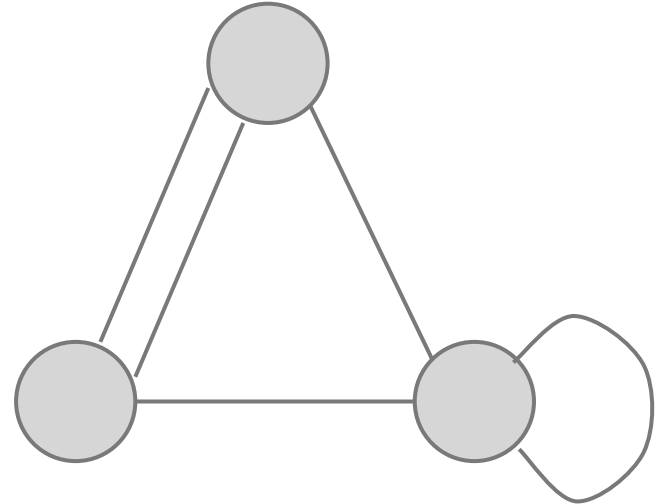
Unweighted



Weighted

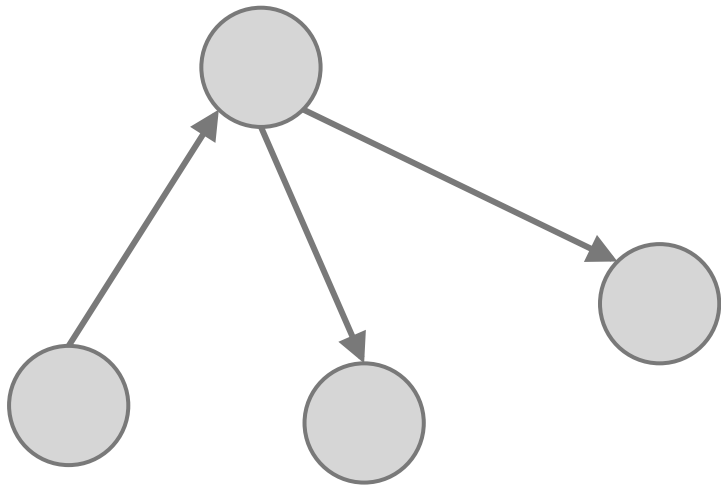


Simple

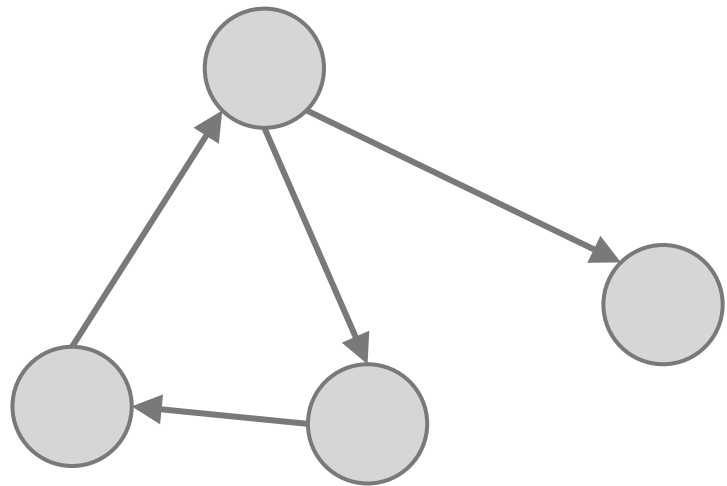


Non-simple

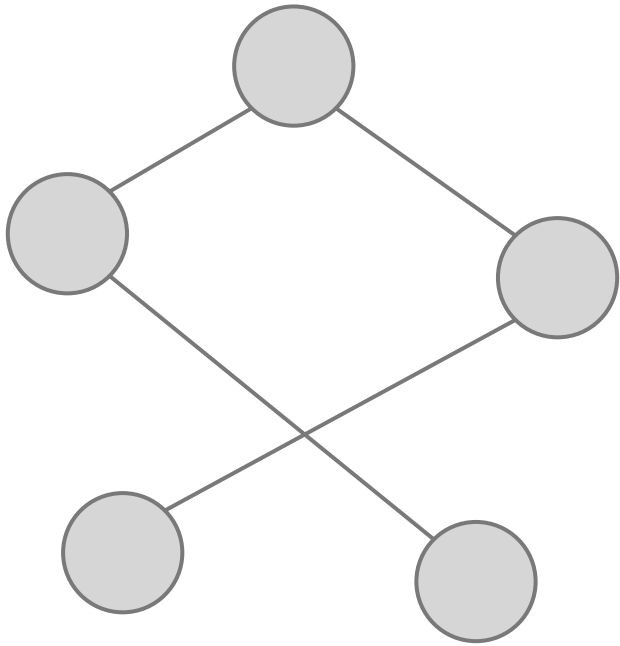
No multiple edge, no self-loop



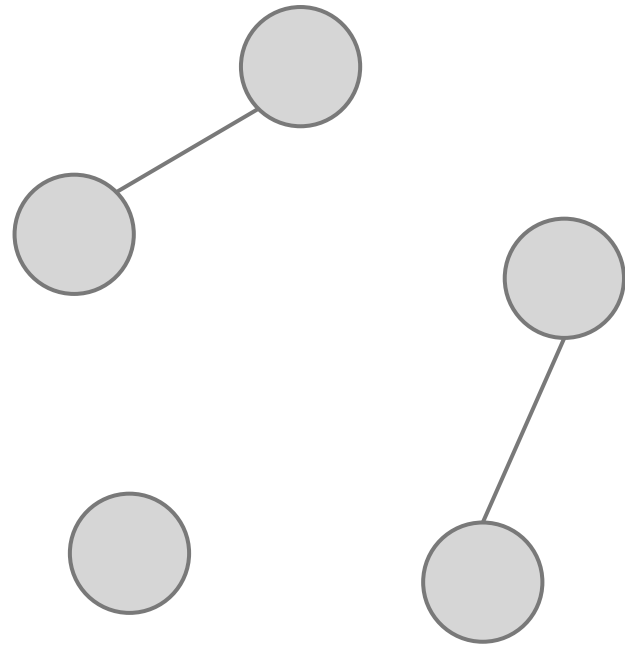
Acyclic



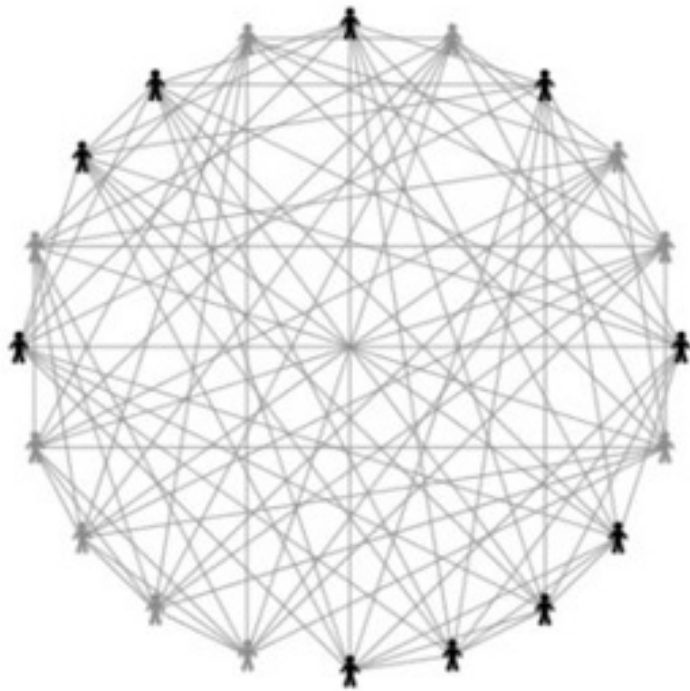
Cyclic



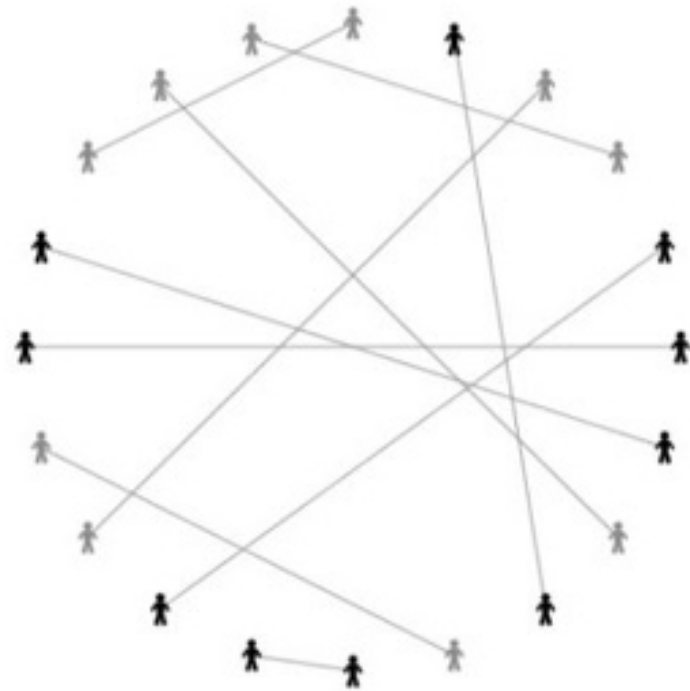
Connected



Disconnected

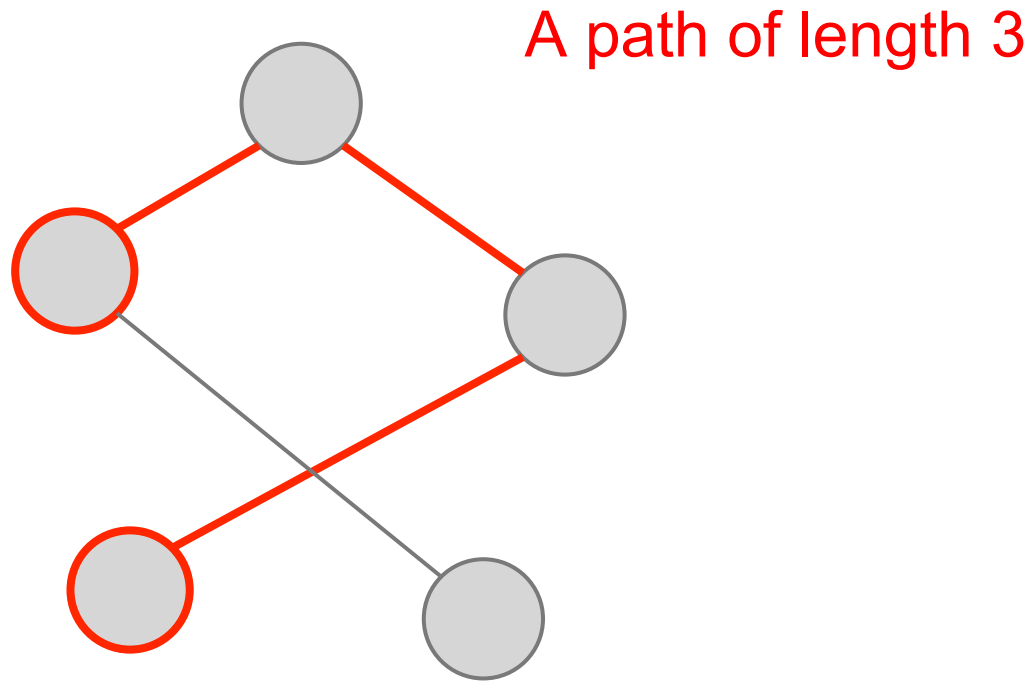


Dense



Sparse

Path



Length of path = number of **edges**

Read Appendix B.4 for more background on graphs.

Operations on a graph

→ **Add** a vertex; **remove** a vertex

→ **Add** an edge; **remove** an edge

→ **Get neighbours** (undirected graph)

◆ Neighbourhood(u): all $v \in V$ such that $(u, v) \in E$

→ **Get in-neighbours / out-neighbours**
(directed graph)

→ **Traversal**: visit every vertex in the graph

Many other operations:

→ Traversal:

BFS (breadth first search)

DFS (depth first search)

→ Given s, t find a (minimum length) path from s to t

→ Given a connected graph G , output a spanning tree of G

→ Is G connected?

Data structures for the graph ADT

- Adjacency **matrix**
- Adjacency **list**

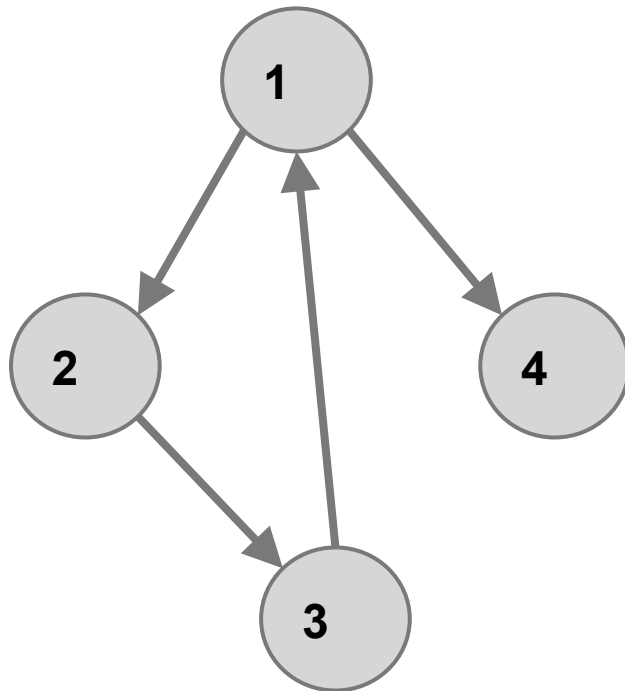
Adjacency matrix

A $|V| \times |V|$ matrix A

Let $V = \{v_1, v_2, \dots, v_n\}$

$$A[i, j] = \begin{cases} 1 & \text{if } (v_i, v_j) \in E \\ 0 & \text{otherwise} \end{cases}$$

Adjacency matrix



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

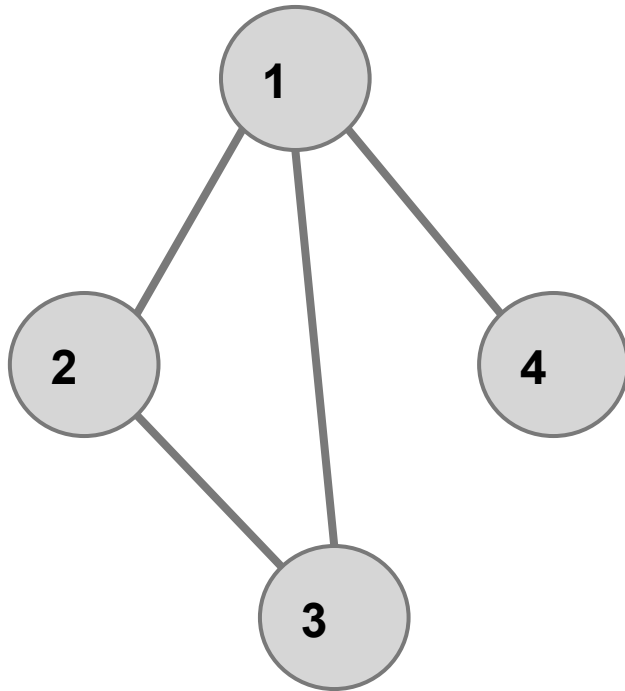
Adjacency matrix

How much space
does it take?

$$|V|^2$$

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

Adjacency matrix (undirected graph)



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

The adjacency matrix of an **undirected** graph is **symmetric**.

Adjacency matrix (undirected graph)

How much space
does it take?

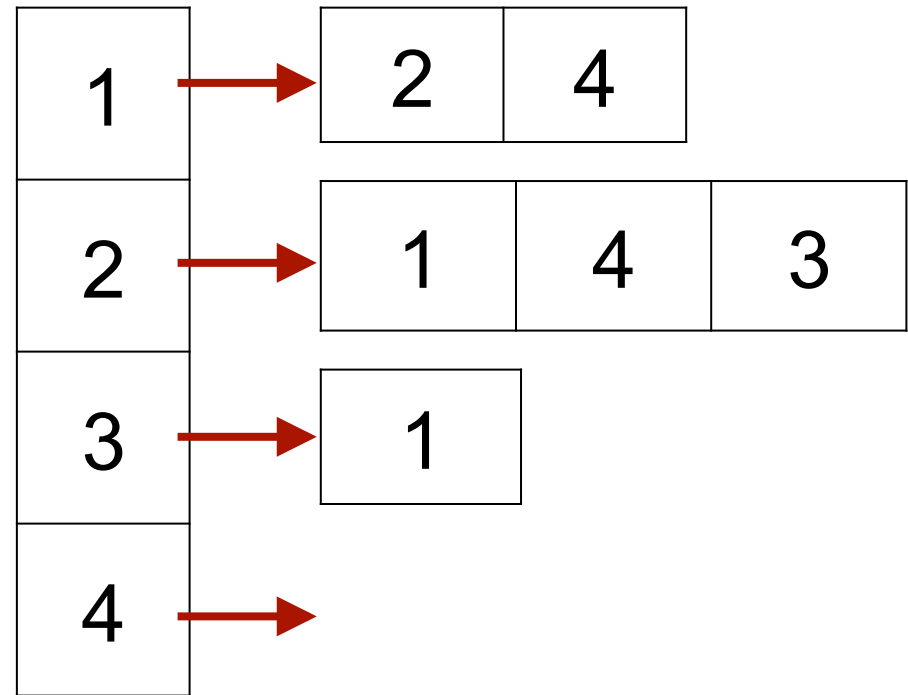
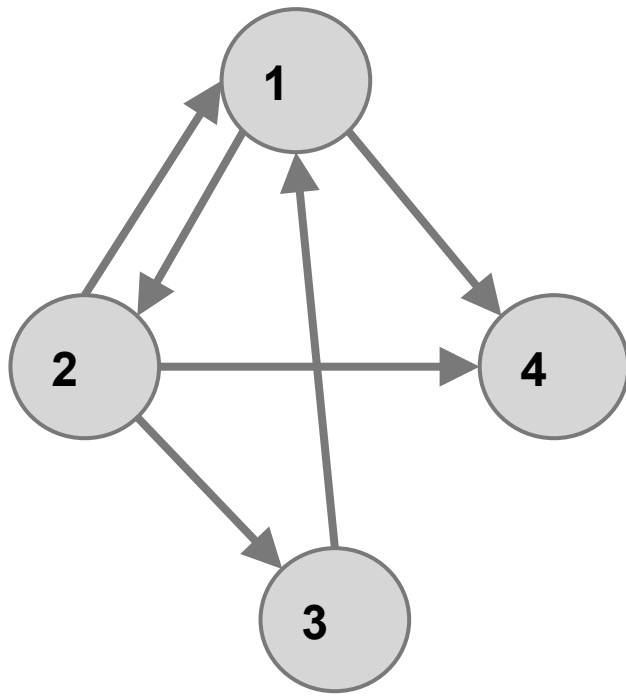
$$|V|^2$$

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

Adjacency list

Adjacency list (directed graph)

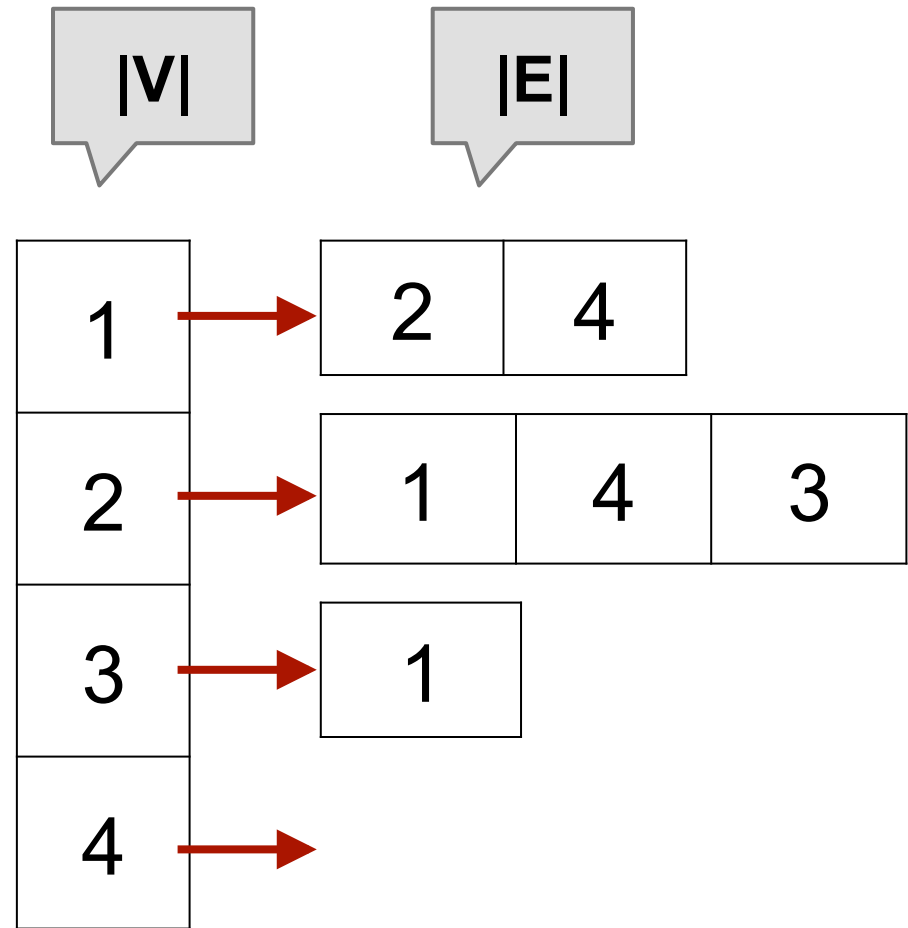
Each vertex v_i stores a list $A[i]$ of v_j that satisfies $(v_i, v_j) \in E$



Adjacency list (directed graph)

How much space does it take?

$|V| + |E|$

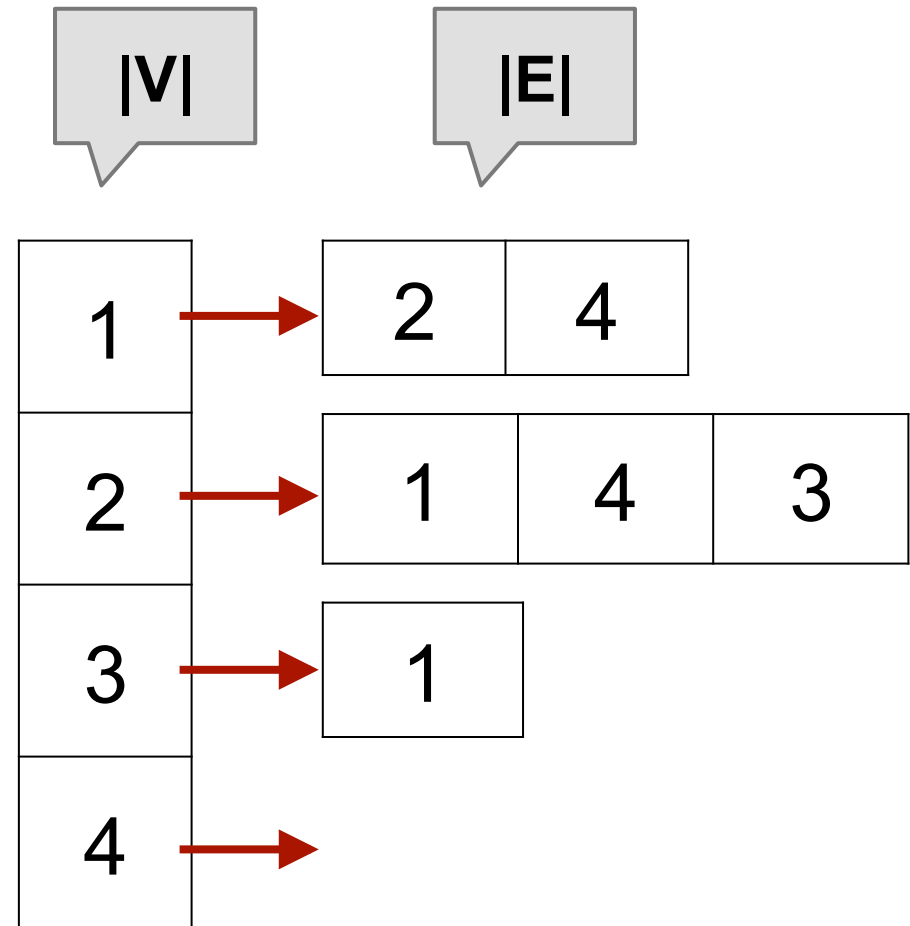


Adjacency list (directed graph)

How much space does it take?

$$|V| + |E|$$

This assumes we can store the name of a vertex in one cell of the linked list. In terms of bits, the size would be more like $|V| + |E| (\log |V|)$

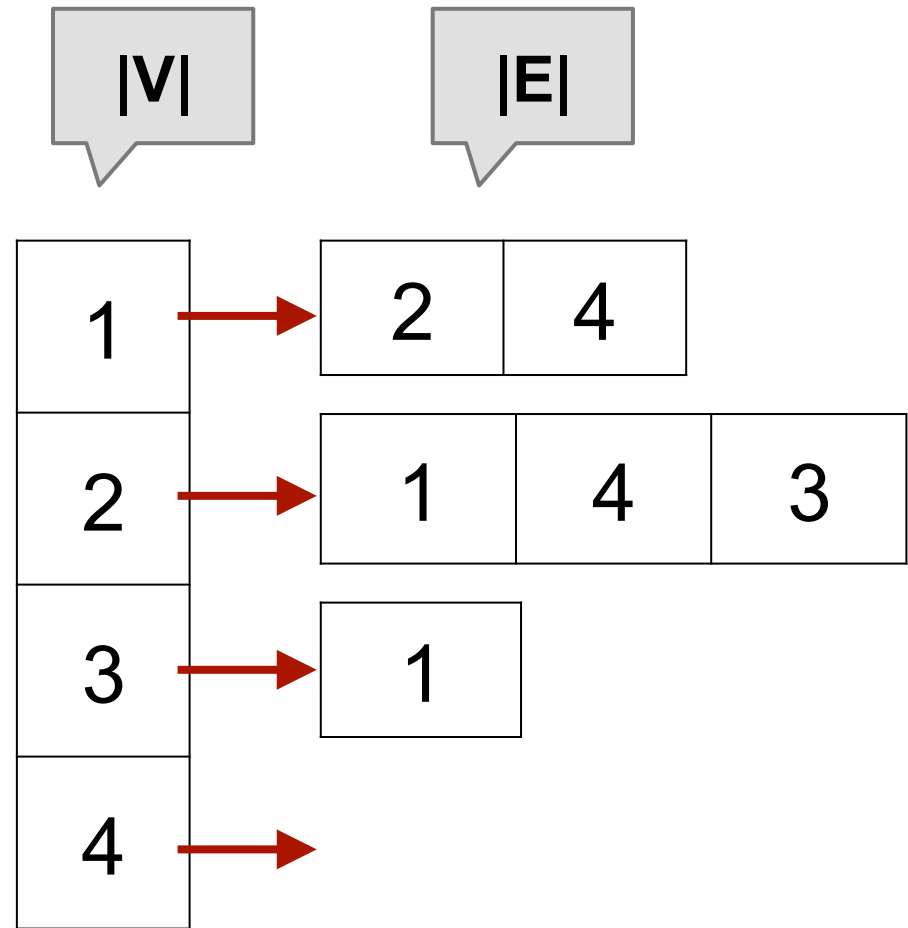


Adjacency list (directed graph)

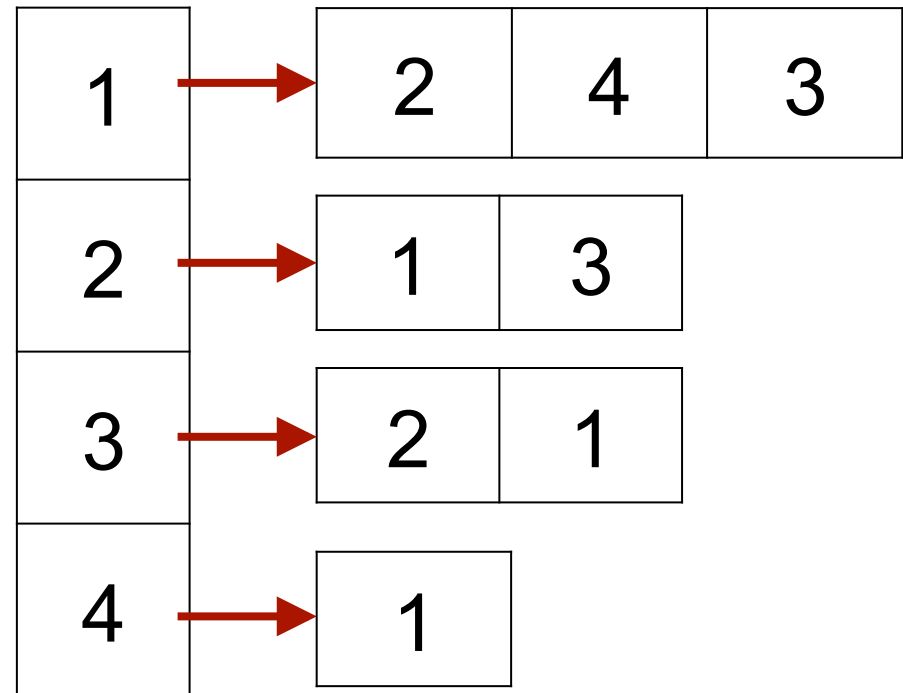
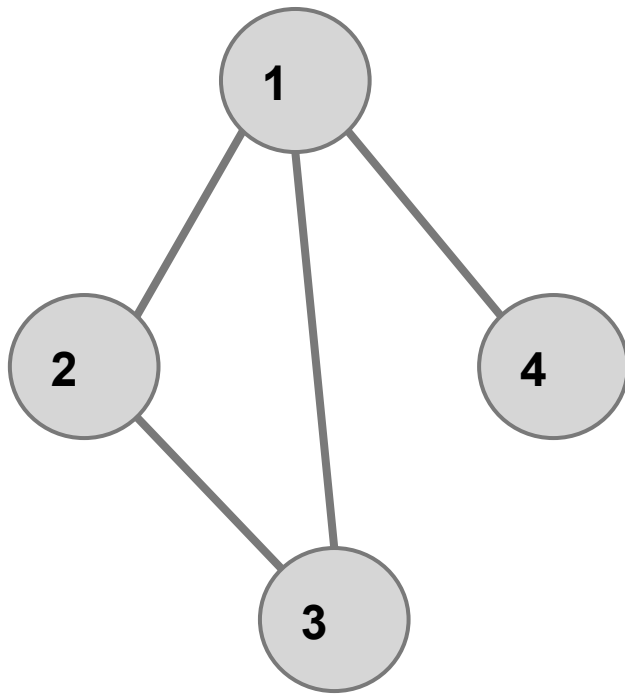
How much space does it take?

$$|V| + |E|$$

One often ignores these lower order factors of $\log n$.
(Recall for hashing, we assume $h(x)$ in constant time.)



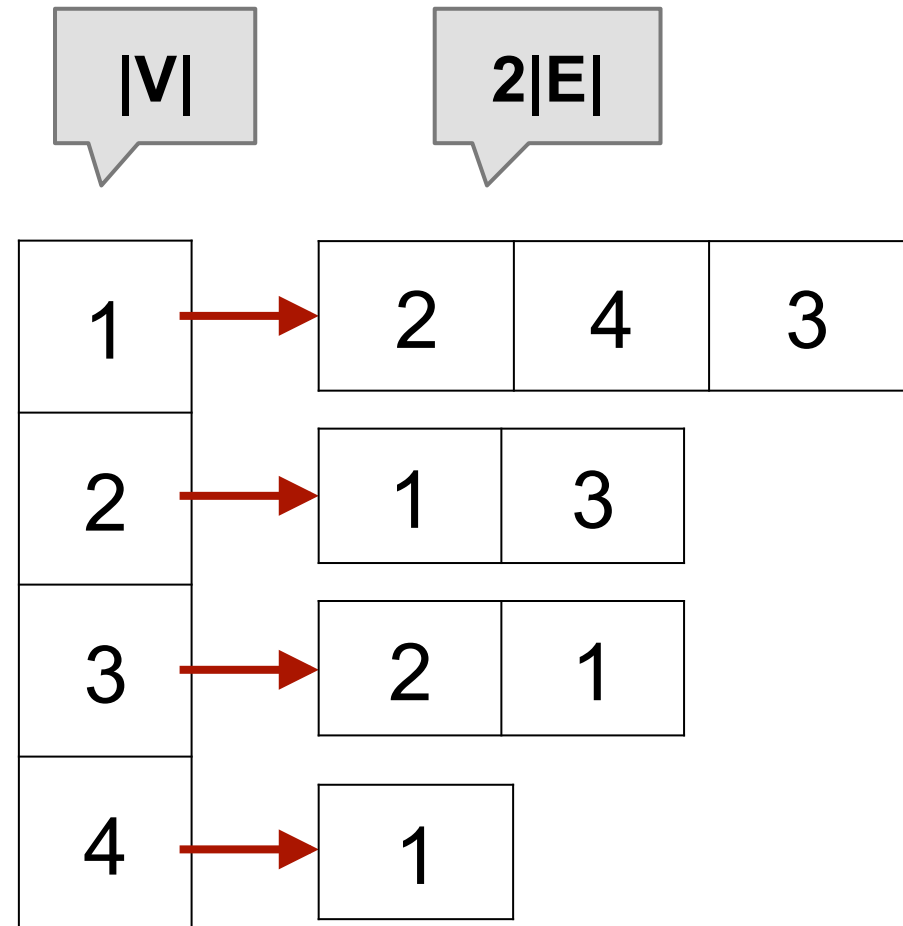
Adjacency list (undirected graph)



Adjacency list (undirected graph)

How much space does it take?

$|V| + 2|E|$



Matrix *versus* List

In term of space complexity

→ adjacency matrix is $\Theta(|V|^2)$

→ adjacency list is $\Theta(|V|+|E|)$

Which one is more space-efficient?

Adjacency list, if $|E| \ll |V|^2$, i.e., the graph is not very **dense**.

Matrix *versus* List

Anything that **Matrix** does better than **List**?



Check whether edge (v_i, v_j) is in **E**

→ **Matrix**: just check if $A[i, j] = 1$, **$O(1)$**

→ **List**: go through list $A[i]$ see if j is in there,
 $O(\text{length of list})$

Takeaway

Adjacency **matrix** or adjacency **list**?

Choose the more appropriate one depending on the problem.

Graph Traversals

BFS and **DFS**



Graph traversals

Visiting **every** vertex **once**, starting from a given vertex.

The visits can follow different **orders**, we will learn about the following two ways

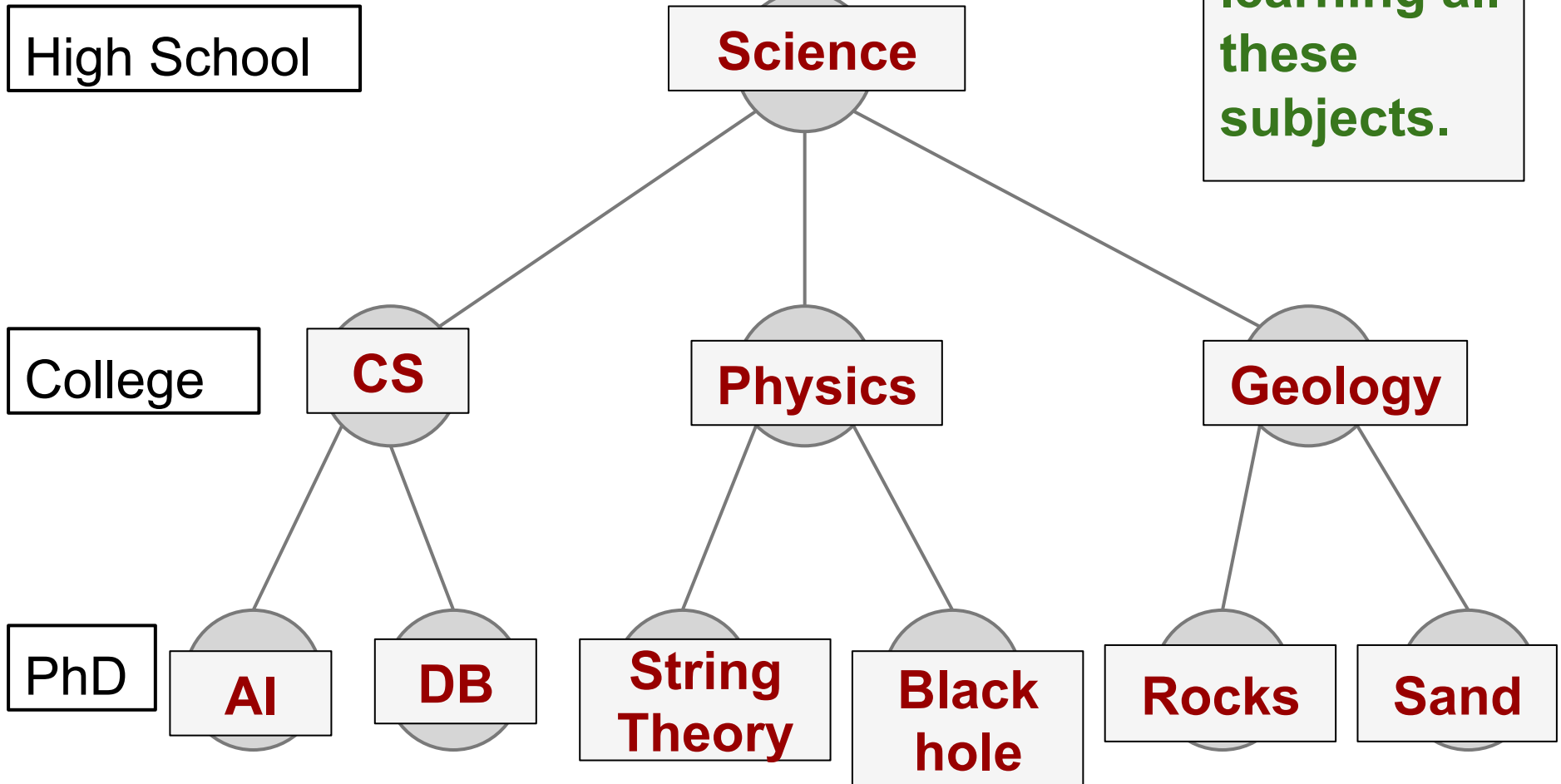
→ **Breadth** First Search (**BFS**)

→ **Depth** First Search (**DFS**)

Intuitions of BFS and DFS

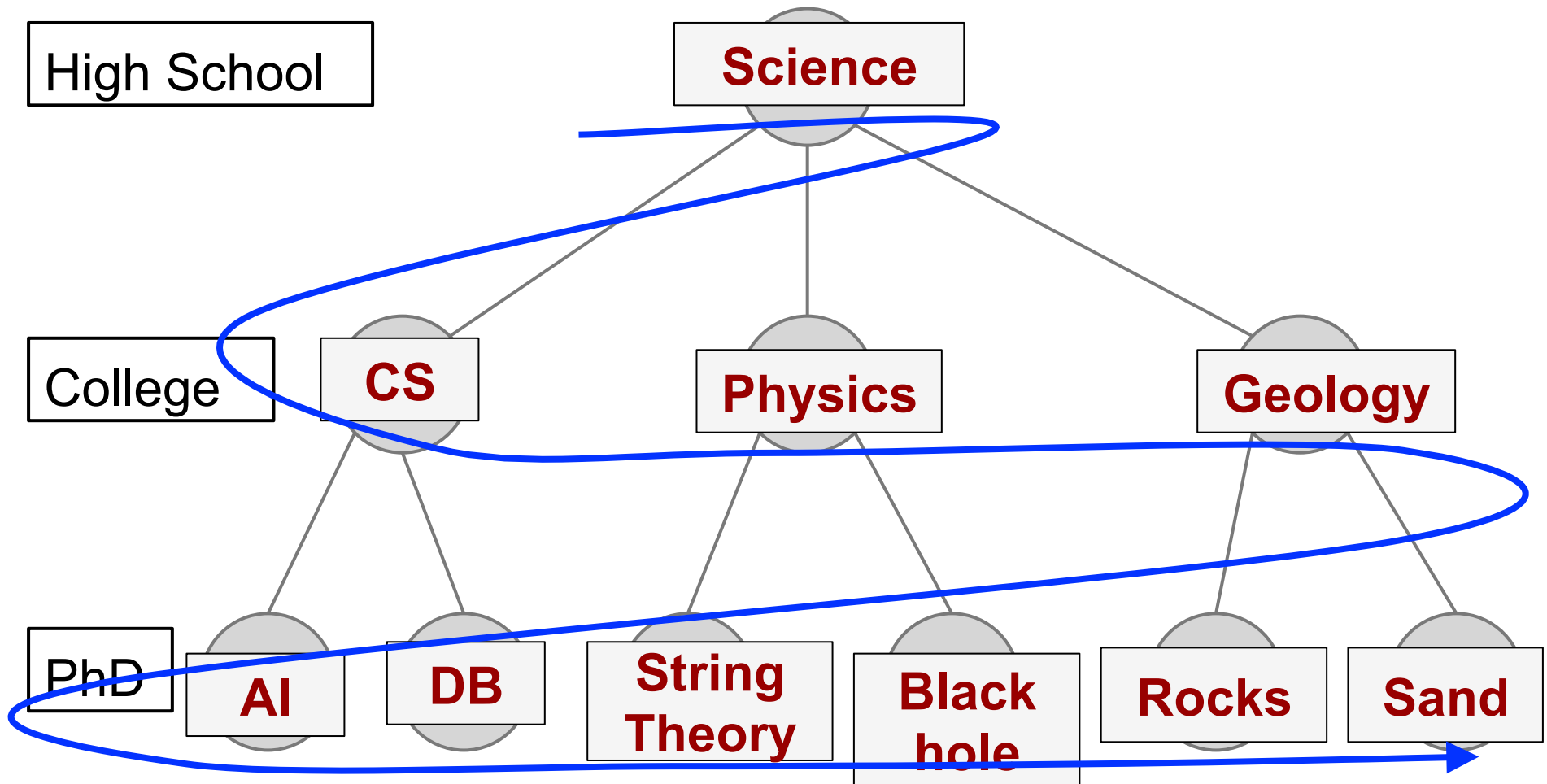
Consider a special graph -- a **tree**

“The knowledge learning tree”



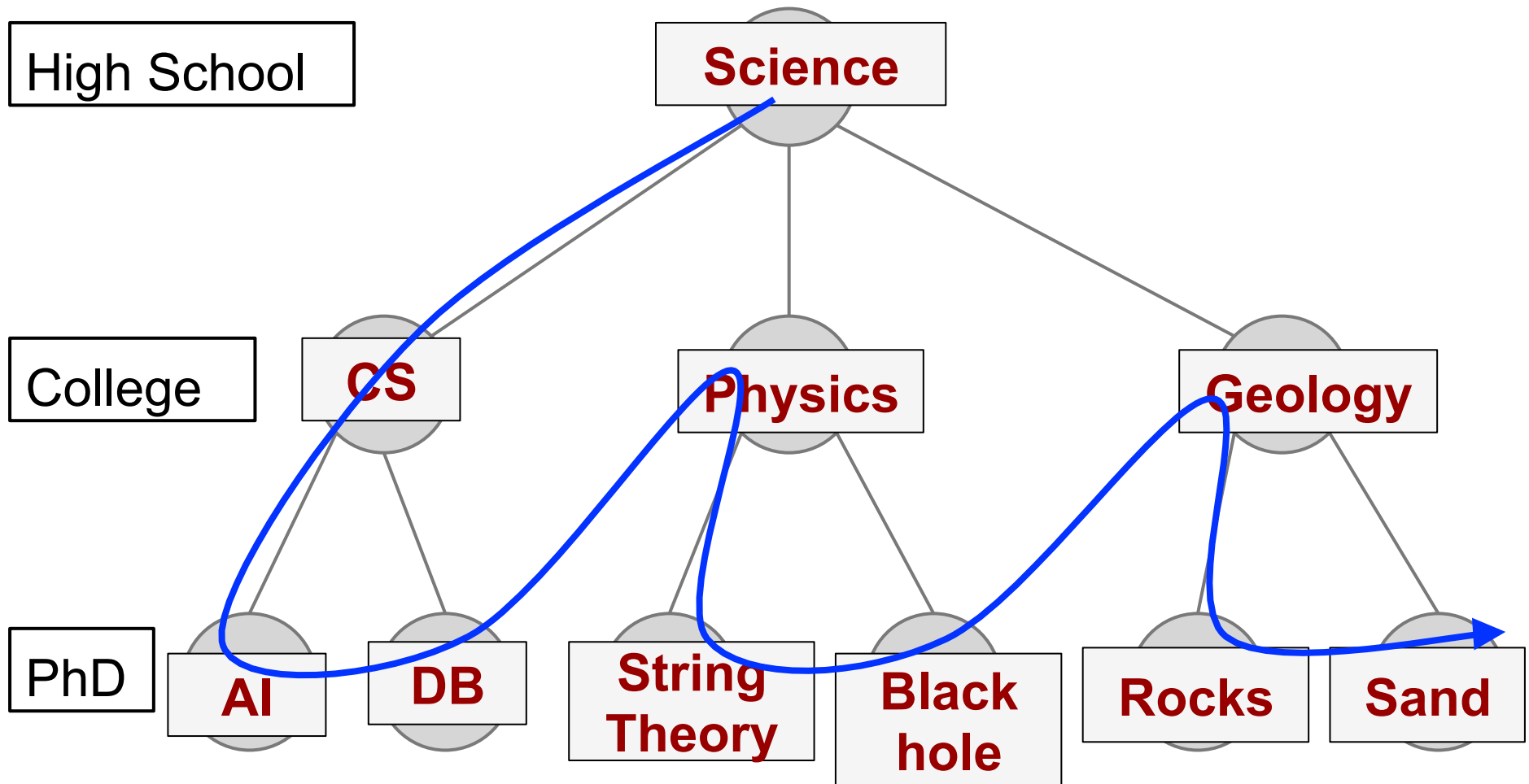
The **Breadth-First** ways of learning these subjects

→ Level by level, finish high school, then all subjects at College level, then finish all subjects in PhD level.



The **Depth-First** way of learning these subjects

→ Go towards PhD whenever possible; only start learning physics after finishing everything in CS.



Now let's seriously start
studying **BFS**



Special case: BFS in a **tree**

Review CSC148:

BFS in a tree (starting from root) is a
level-by-level traversal.

(NOT preorder!)

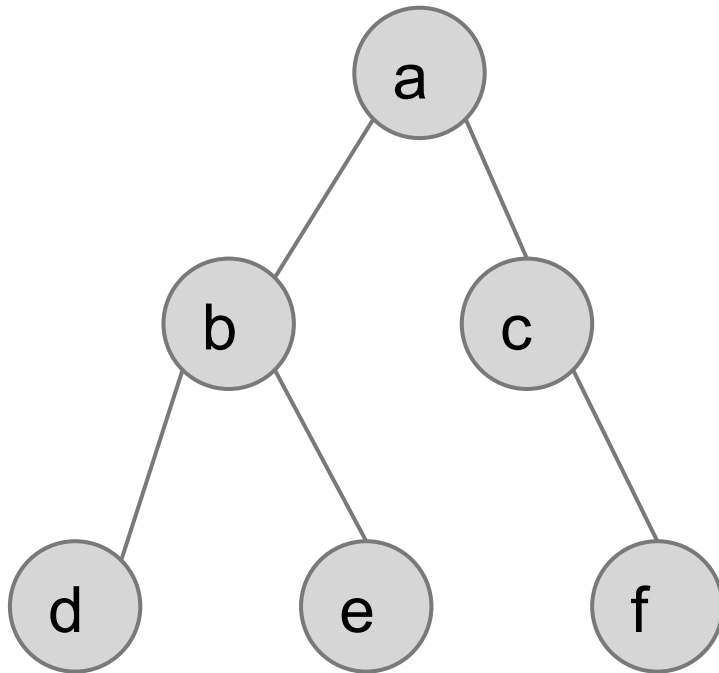
What ADT did we use for implementing the **level-by-level** traversal?

Queue!

Special case: BFS in a **tree**

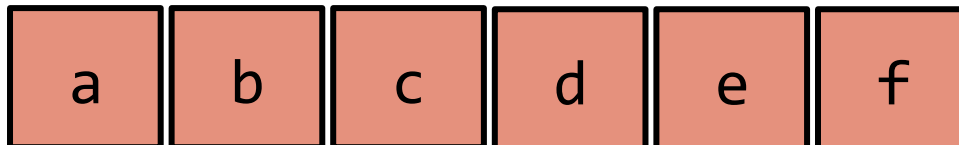
Output:

a
b
c
d
e
f



```
NOT_YET_BFS(root):  
  Q ← Queue()  
  Enqueue(Q, root)  
  while Q not empty:  
    x ← Dequeue(Q)  
    print x  
    for each child c of x:  
      Enqueue(Q, c)
```

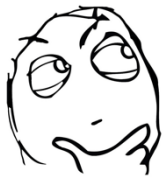
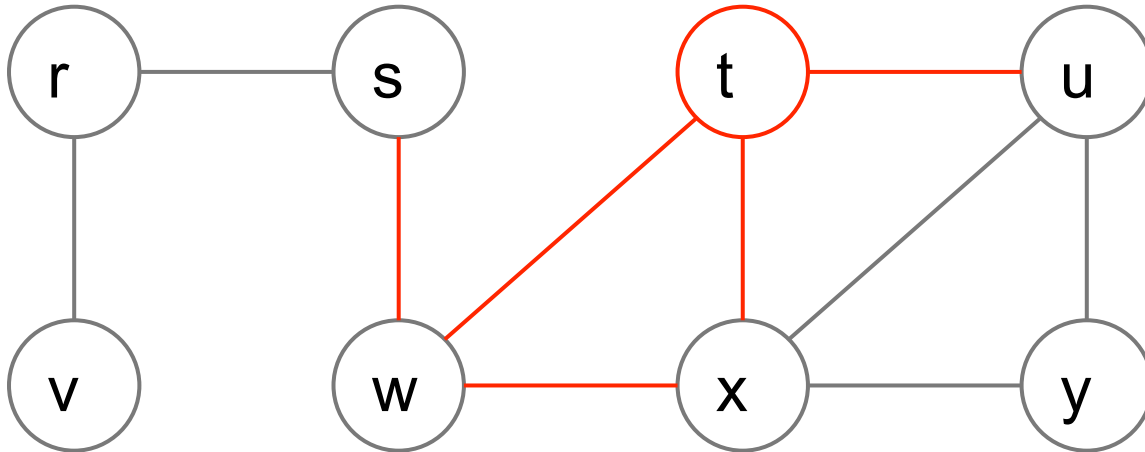
Queue:



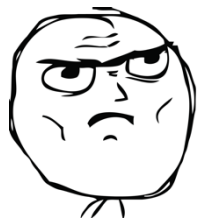
EMPTY!

DQ DQ DQ DQ DQ DQ

The real deal: BFS in a **Graph**



If we just run **NOT_YET_BFS(t)** on the above graph. What problem would we have?



It would want to visit some vertices more than once (e.g., **x**)

```
NOT_YET_BFS(root):  
Q ← Queue()  
Enqueue(Q, root)  
while Q not empty:  
  x ← Dequeue(Q)  
  print x  
  for each neighbor c of x:  
    Enqueue(Q, c)
```

How avoid visiting a vertex twice

Remember you visited it by **labelling** it using **colours**.

- **White**: “unvisited”
- **Gray**: “encountered”
- **Black**: “explored”



- Initially all vertices are **white**
- Colour a vertex **gray** the **first** time visiting it
- Colour a vertex **black** when **all** its **neighbours** have been encountered
- Avoid visiting **gray** or **black** vertices
- In the end, all vertices are **black** (sort-of)

Some other values we want to remember during the traversal...

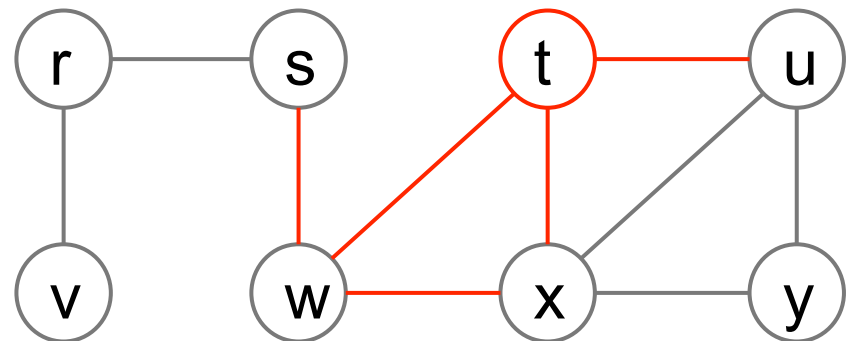
→ **pi[v]**: the vertex from which v is encountered

◆ “I was introduced as **whose** neighbour?”

→ **d[v]**: the distance value

◆ the distance from v to the source vertex of the BFS

This **d[v]** is going to be **really** useful!



Pseudocode: the **real** BFS

```
BFS(G=(V, E), s):  
1   for all v in V:  
2       colour[v] ← white  
3       d[v] ← ∞  
4       pi[v] ← NIL  
5   Q ← Queue()  
6   colour[s] ← gray  
7   d[s] ← 0  
8   Enqueue(Q, s)  
9   while Q not empty:  
10      u ← Dequeue(Q)  
11      for each neighbour v of u:  
12          if colour[v] = white  
13              colour[v] ← gray  
14              d[v] ← d[u] + 1  
15              pi[v] ← u  
16              Enqueue(Q, v)  
17      colour[u] ← black
```

Initialize vertices

source s is encountered

distance from s to s is 0

only visit unvisited vertices

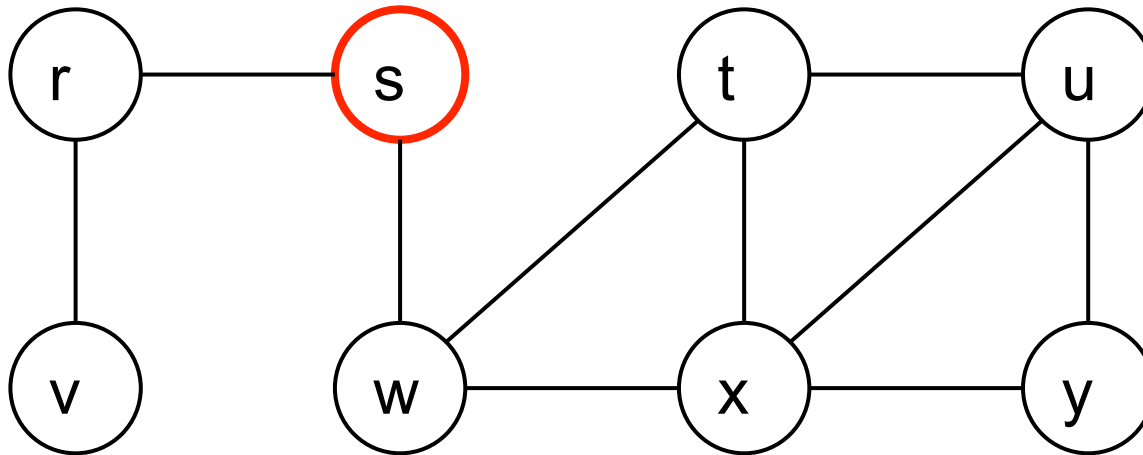
v is “1-level” farther from s than u

v is introduced as u’s neighbour

all neighbours of u have been encountered, therefore u is explored

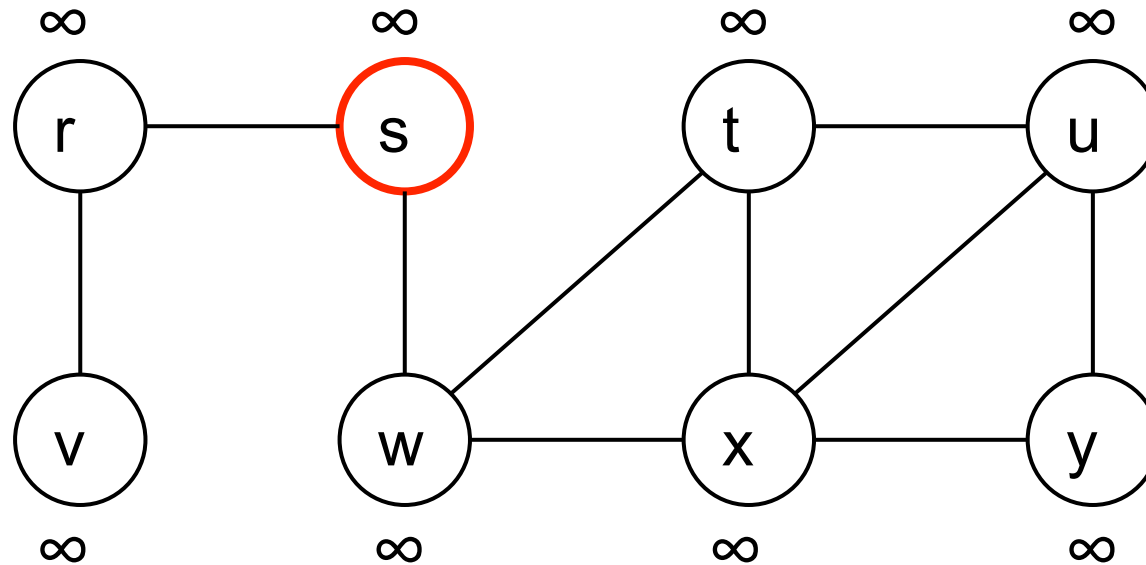
The blue lines are
the same as
NOT_YET_BFS

Let's run an example!



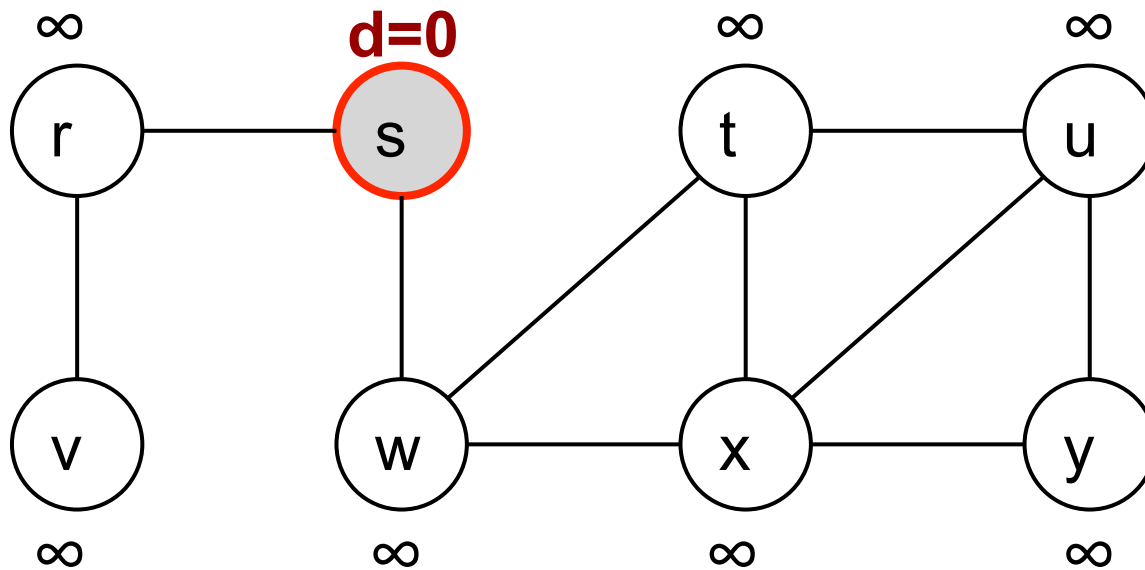
BFS(G, s)

After initialization

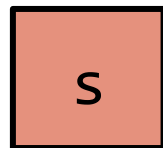


All vertices are **white** and have **d = ∞**

Start by “encountering” the source

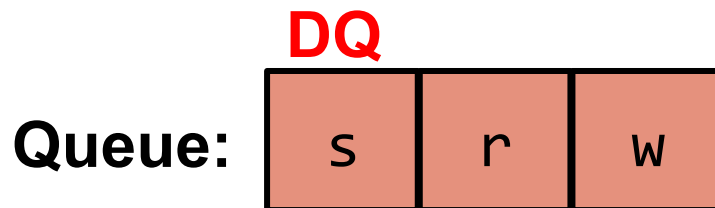
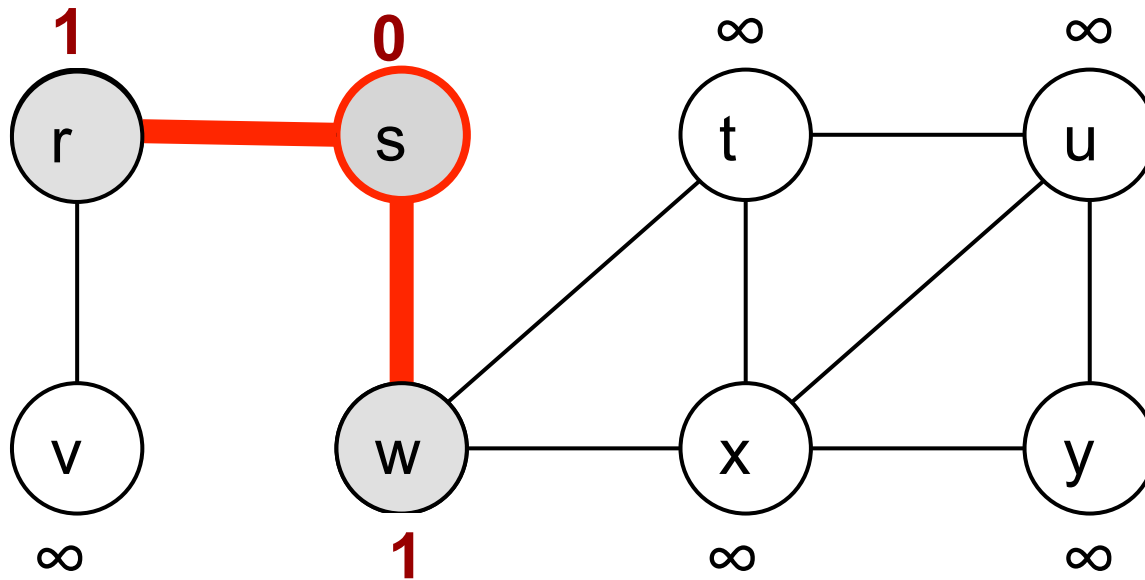


Queue:



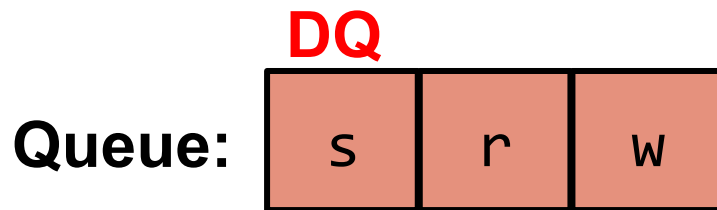
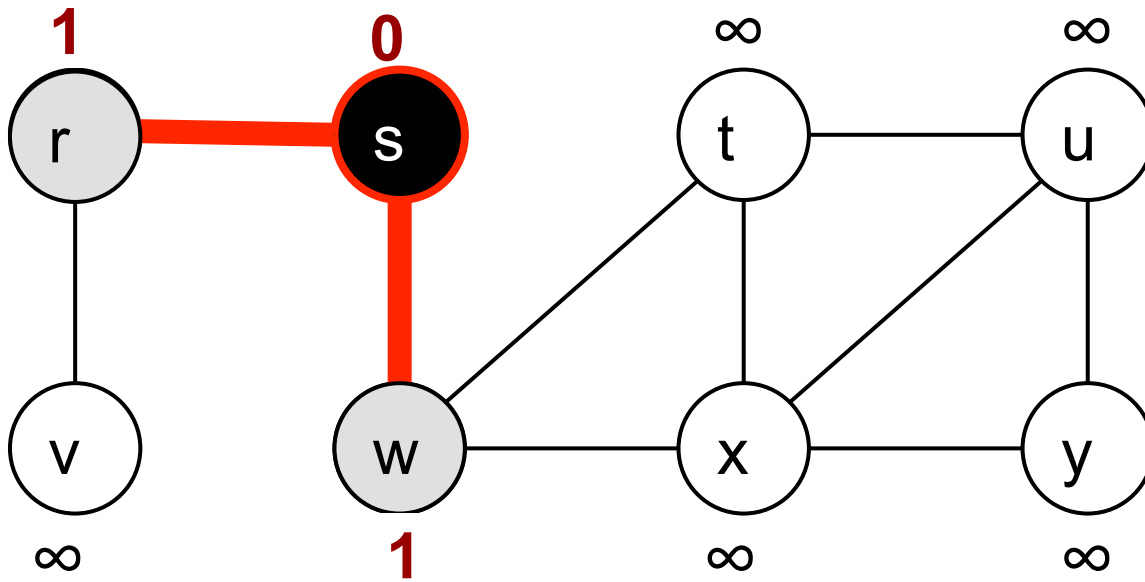
Colour the source **gray** and set its $d = 0$, and Enqueue it

Dequeue, explore neighbours

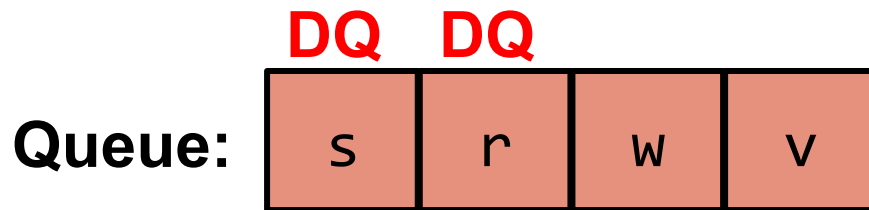
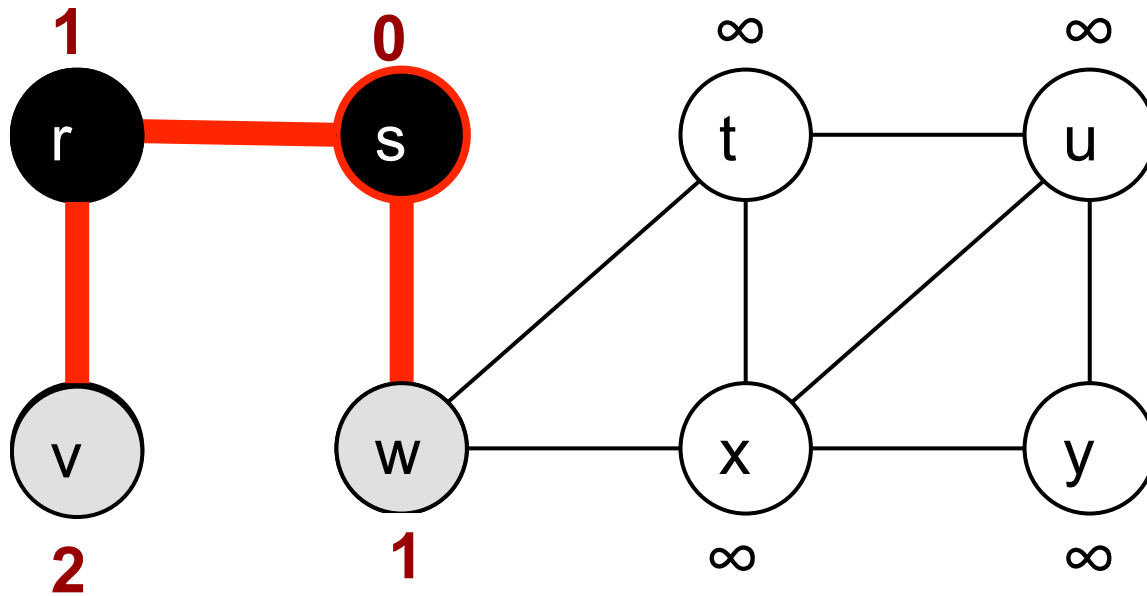


The red edge indicates the $\text{pi}[v]$ that got remembered

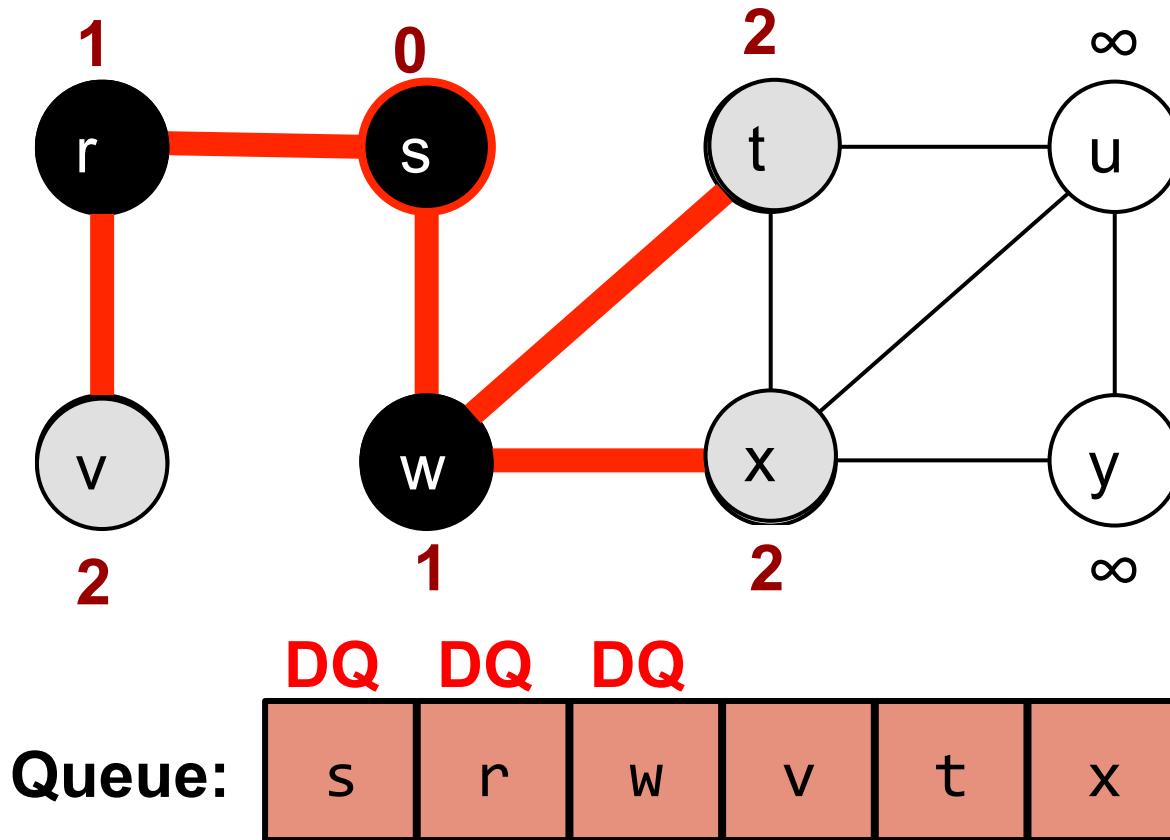
Colour black after exploring all neighbours



Dequeue, explore neighbours (2)



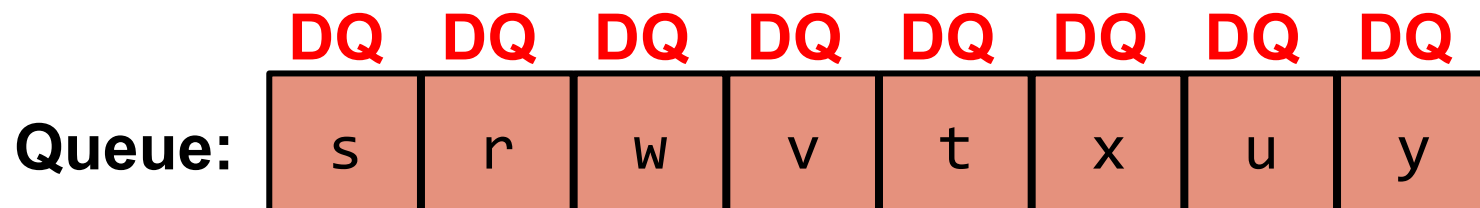
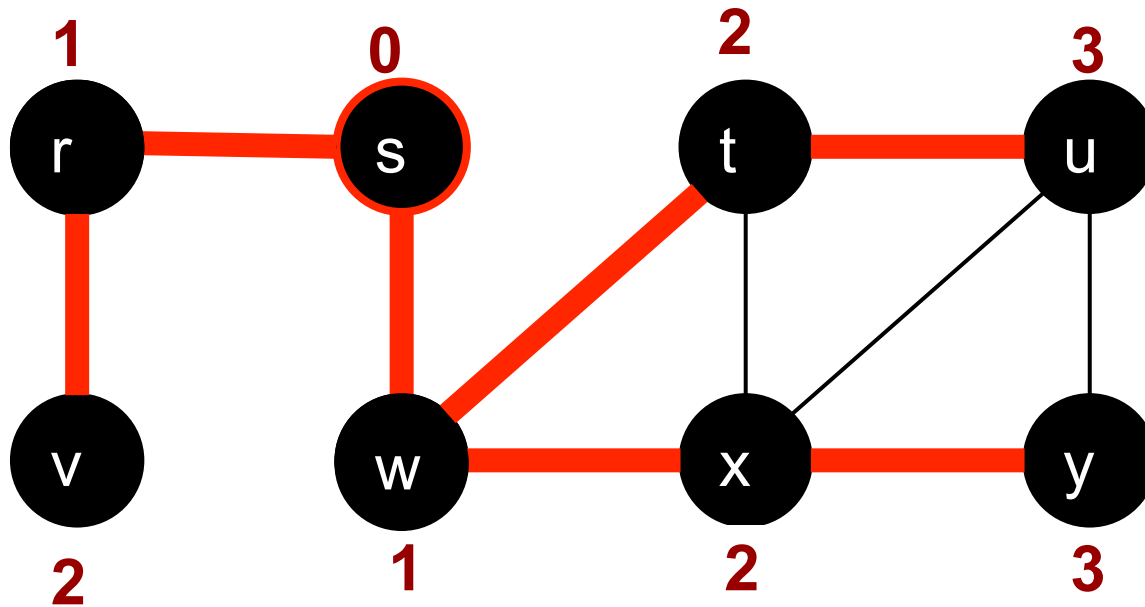
Dequeue, explore neighbours (3)



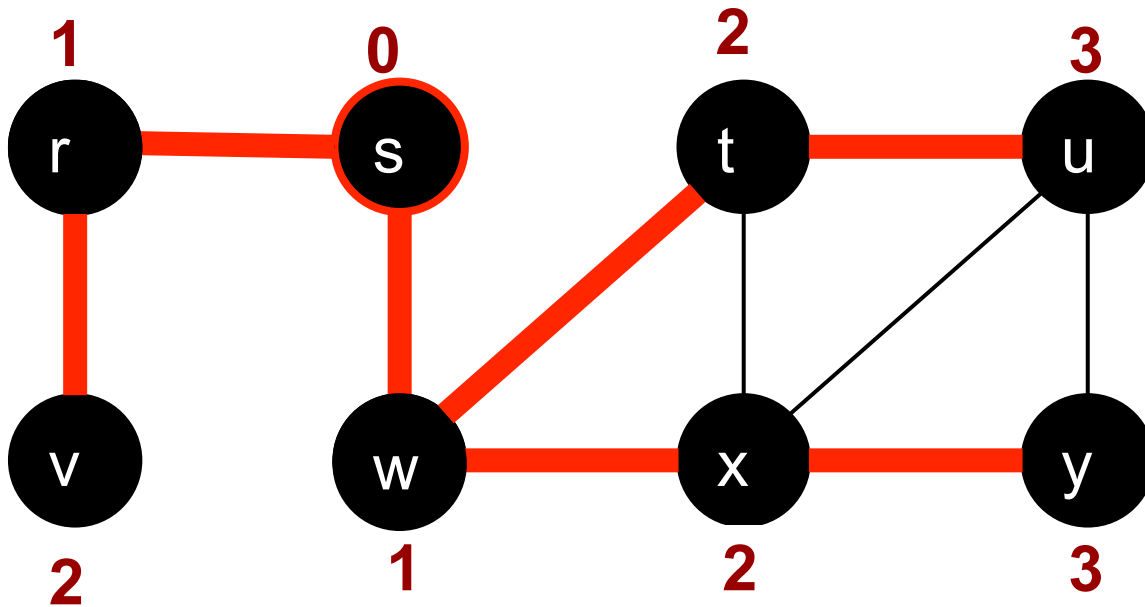
after a few more steps...



BFS done!

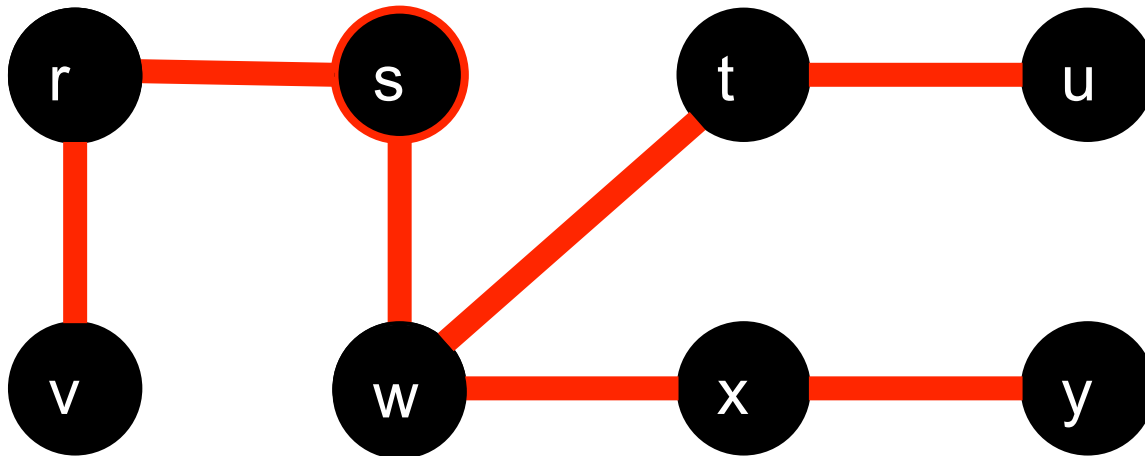


**What do we get after
doing all this?**



First of all, we get to visit **every** vertex **once**.

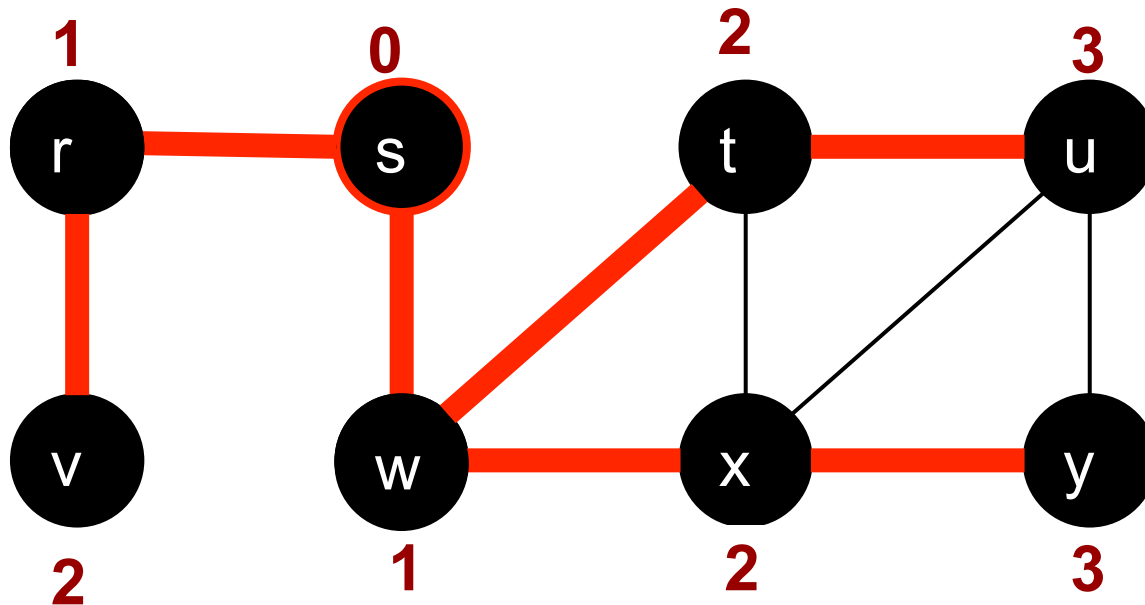
Did you know? The official name of the red edges are called “**tree edges**”.



This is called the **BFS-tree**, it's a **tree** that connects all vertices, if the graph is **connected**.

These $d[v]$ values, we said they were going to be really useful.

Short path from u to s :
 $u \rightarrow \text{pi}[u] \rightarrow \text{pi}[\text{pi}[u]] \rightarrow \text{pi}[\text{pi}[\text{pi}[u]]] \rightarrow \dots \rightarrow s$

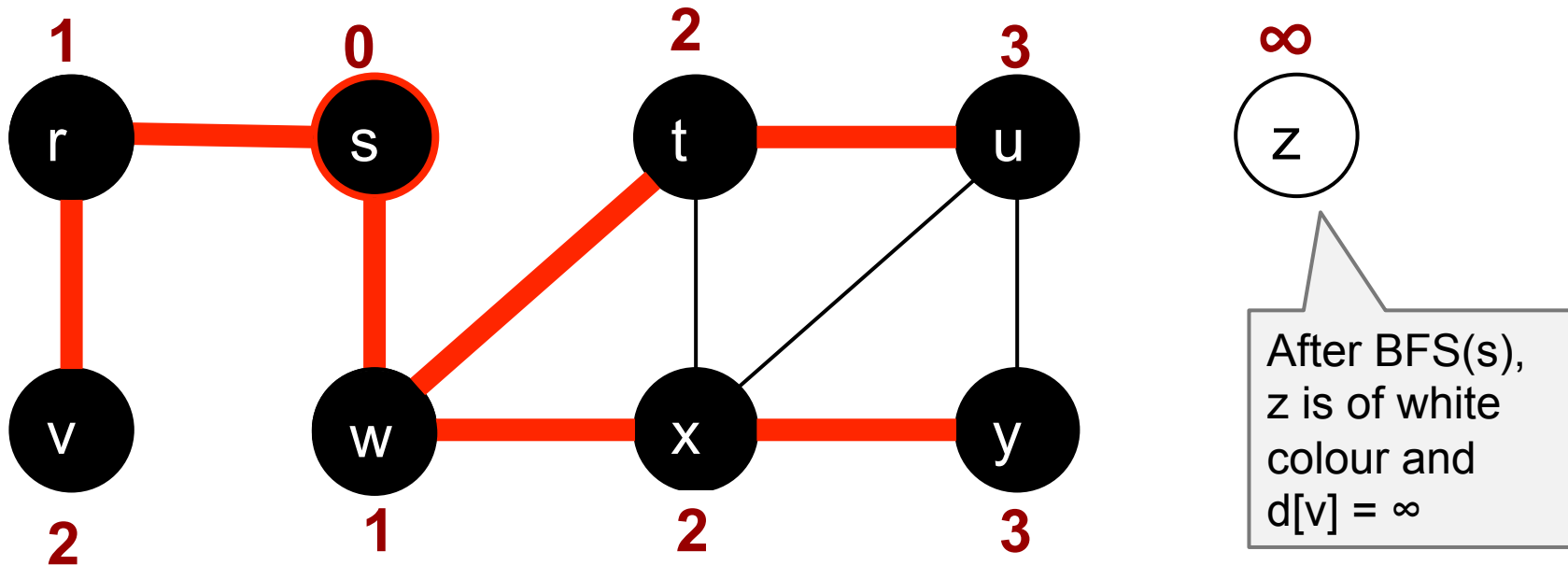


The value indicates the vertex's **distance** from the source vertex.

Actually more than that, it's the **shortest-path distance**, we can prove it.

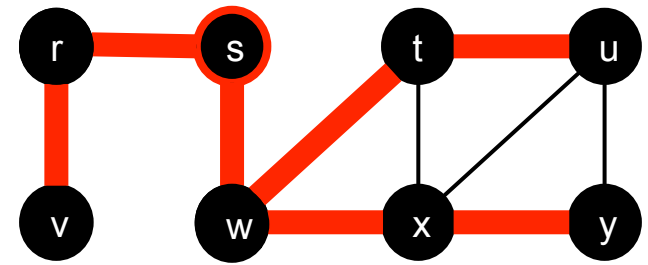
How about finding **short path** itself?
Follow the red edges, $\text{pi}[v]$ comes in handy for this.

What if G is **disconnected**?



The infinite distance value of z indicates that it is **unreachable** from the source vertex.

Runtime analysis!



The total amount of work (use **adjacency list**):

→ Visit each vertex once

- ◆ Enqueue, Dequeue, change colours, assign $d[v]$, ..., constant work per vertex
- ◆ in total: $O(|V|)$

→ At each vertex, check all its neighbours (all its **incident edges**)

- ◆ Each edge is checked **twice** (by the two end vertices)
- ◆ in total: $O(|E|)$

Total runtime:
 $O(|V|+|E|)$

Summary of BFS

- Explores **breadth** rather than **depth**
- Useful for getting **single-source shortest paths** on **unweighted** graphs
- Useful for testing reachability
- Runtime **$O(|V|+|E|)$** with adjacency **list** (with adjacency **matrix** it'll be different)

Next week

DFS



BFS

