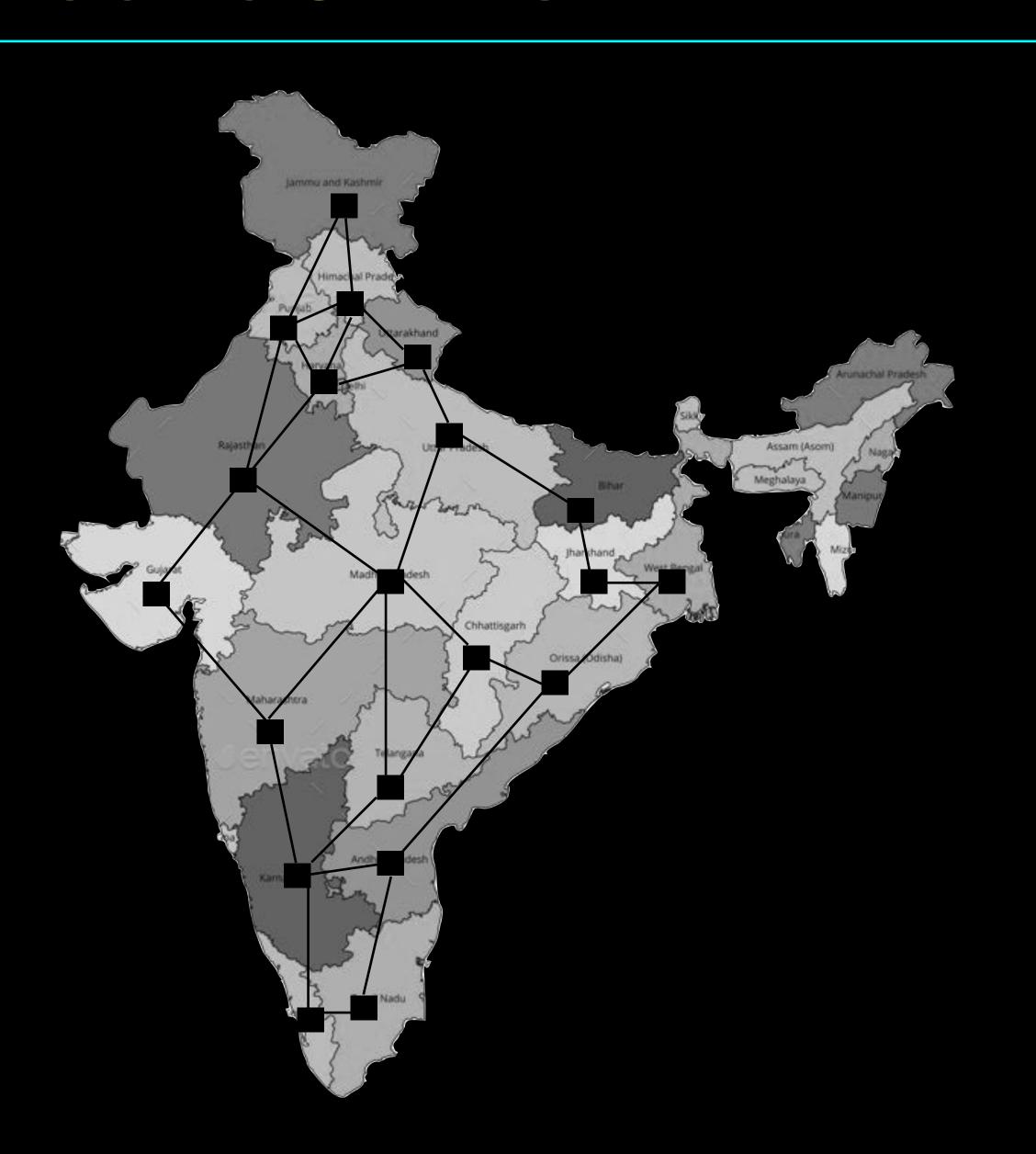
# ADVANCED PROGRAMMING

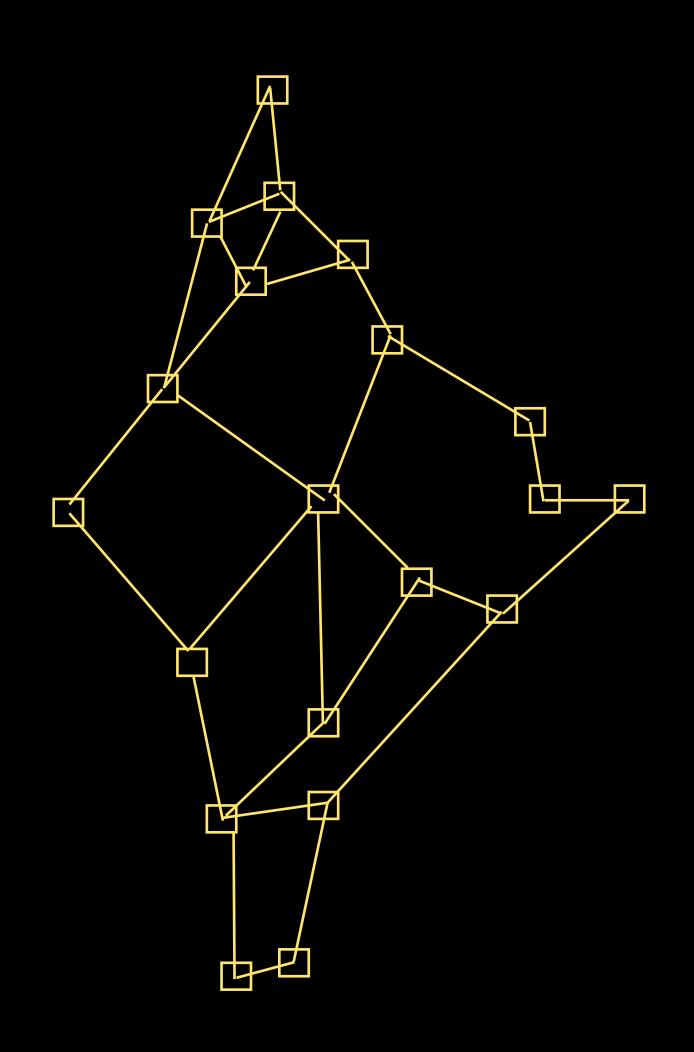
Graphs

Ramaseshan Ramachandran

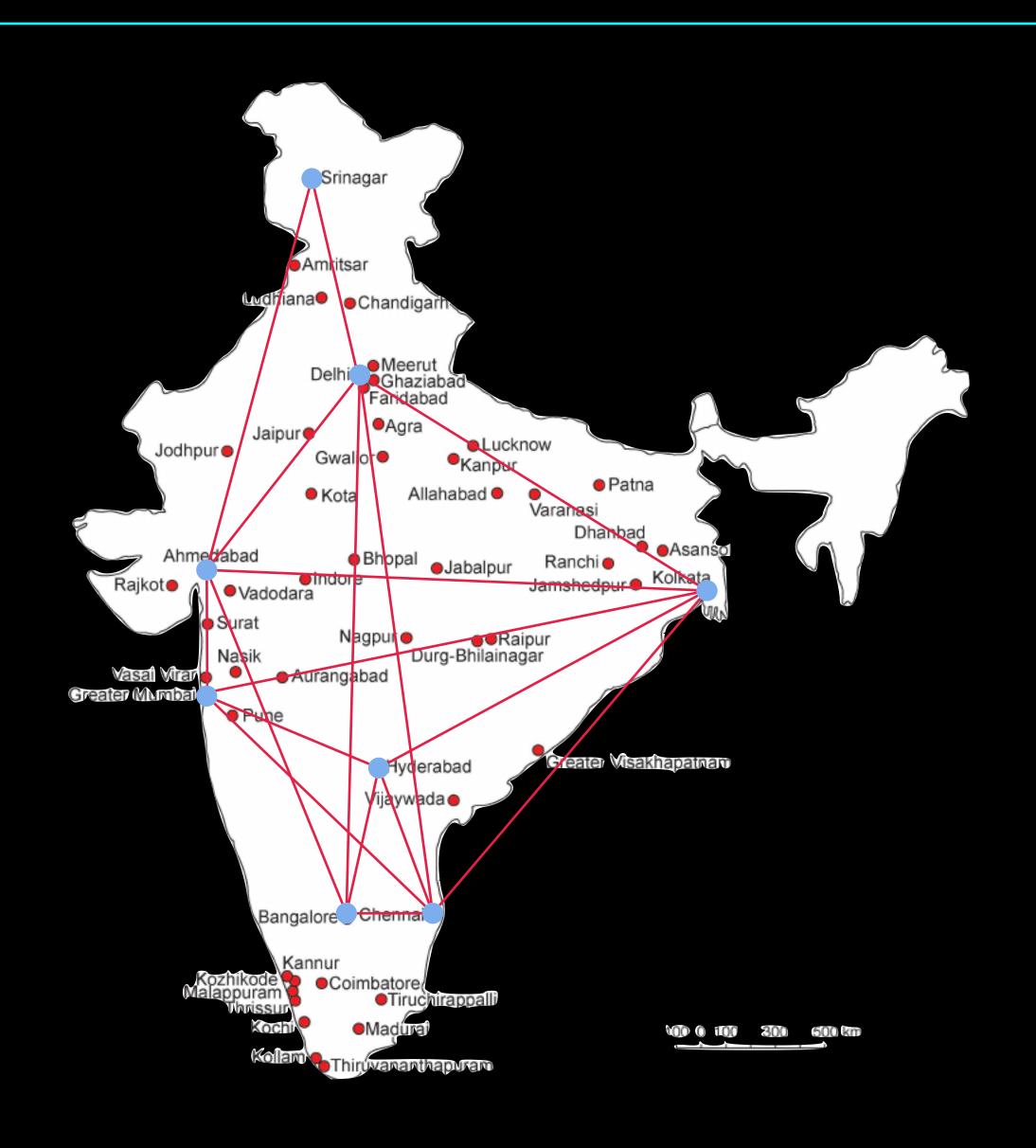
# COLOURING A MAP



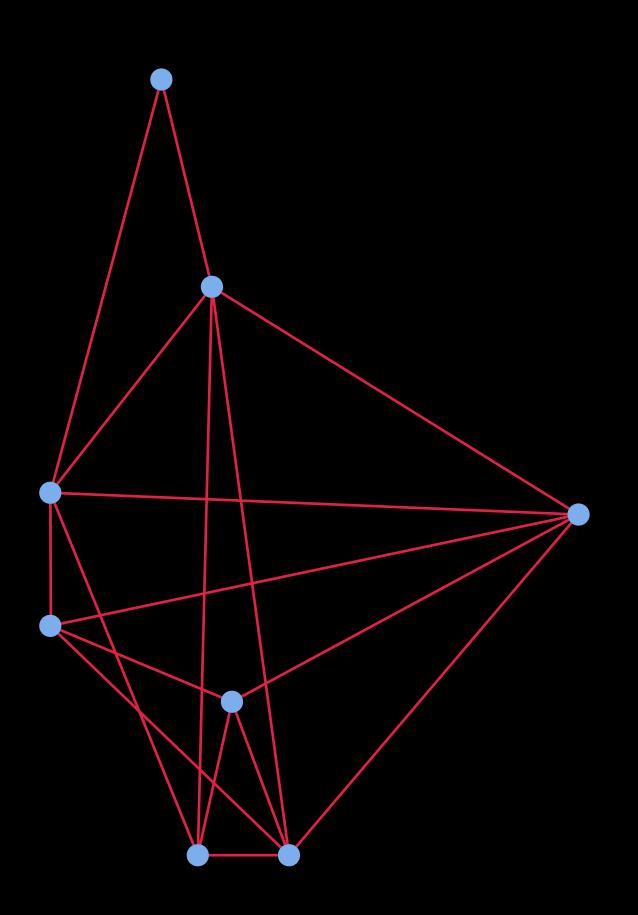
# COLOURING A MAP



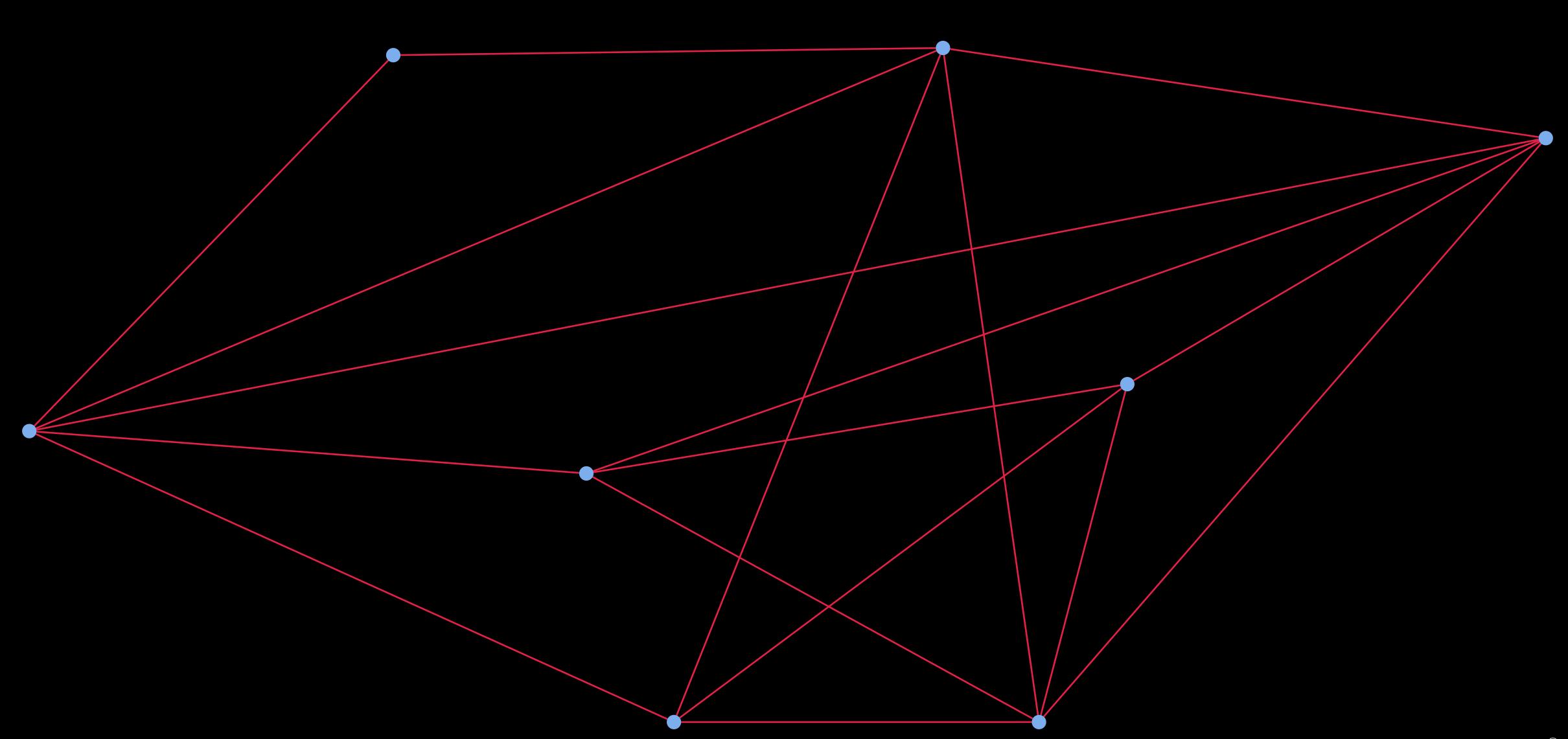
# FLIGHT CONNECTIONS



# CITIES AS A GRAPH

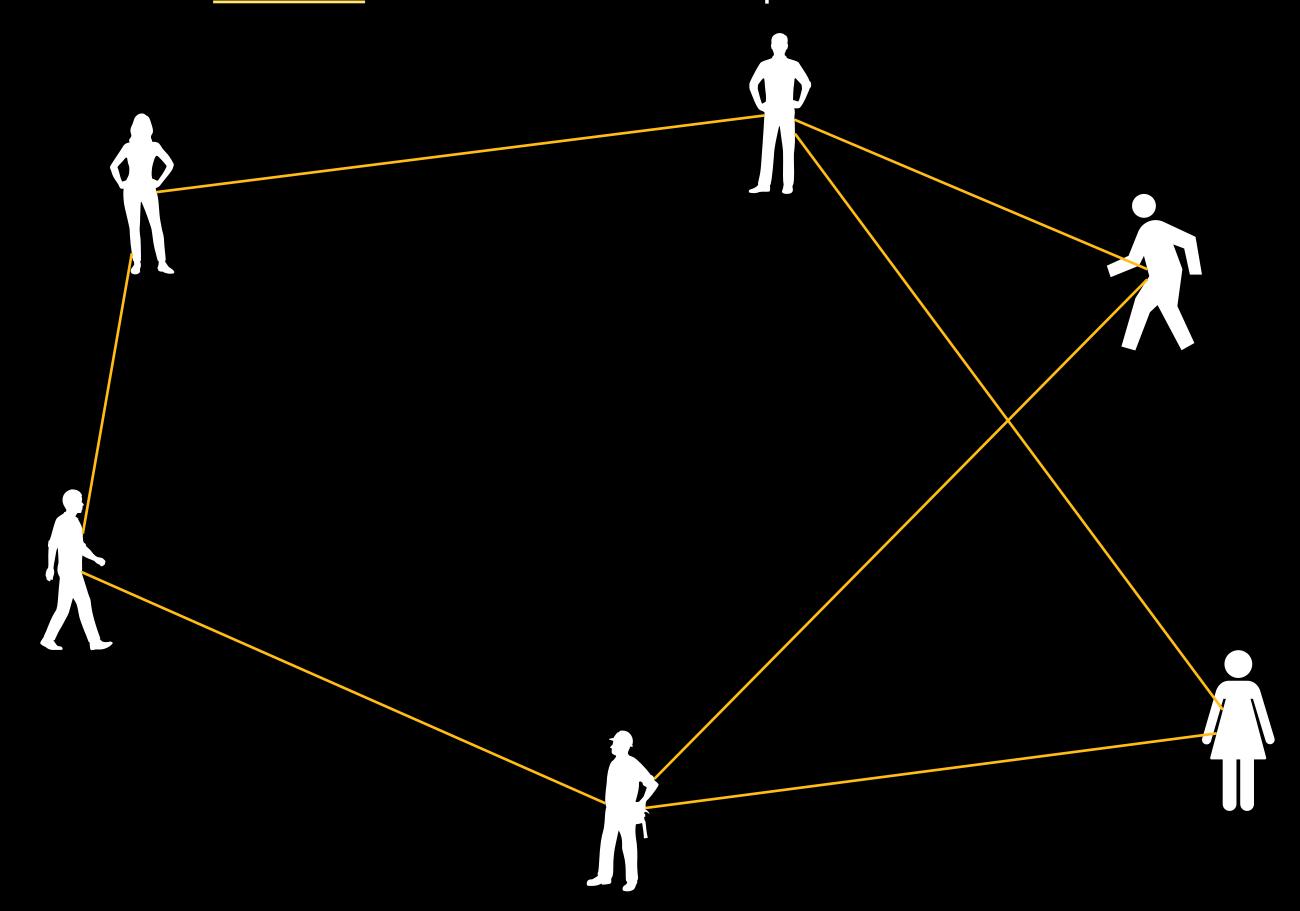


# CITIES AS A GRAPH



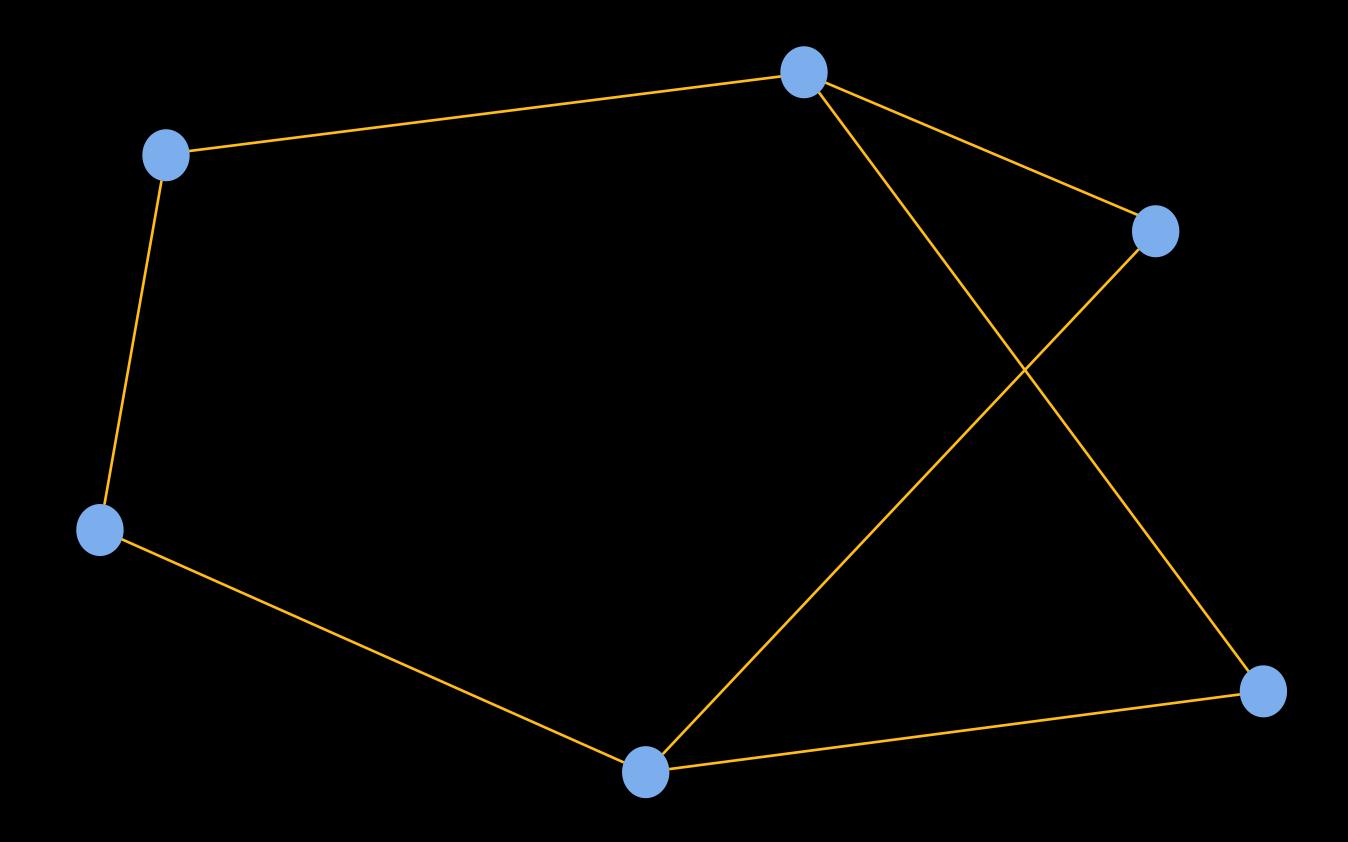
# FRIENDSHIP

A relation is defined as a <u>subset</u> of the Cartesian product of two sets



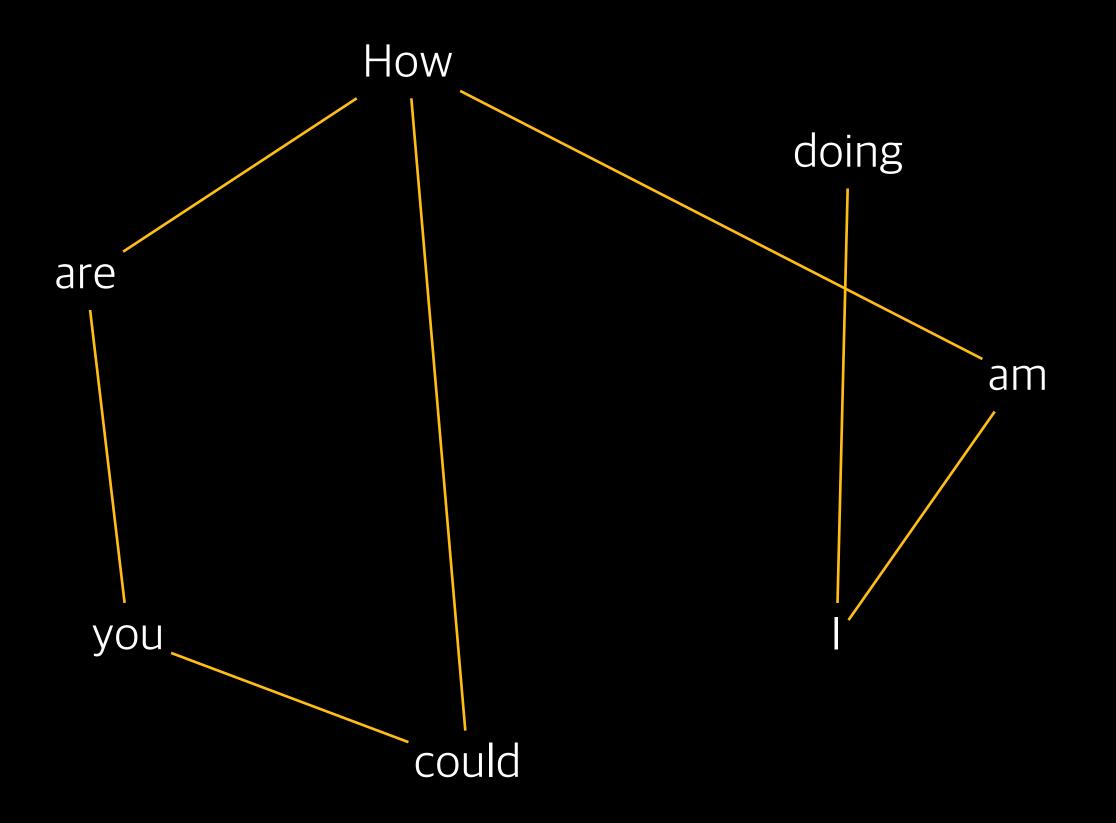
# RELATIONSHIP AS A GRAPH

A relation is defined as a <u>subset</u> of the Cartesian product of two sets



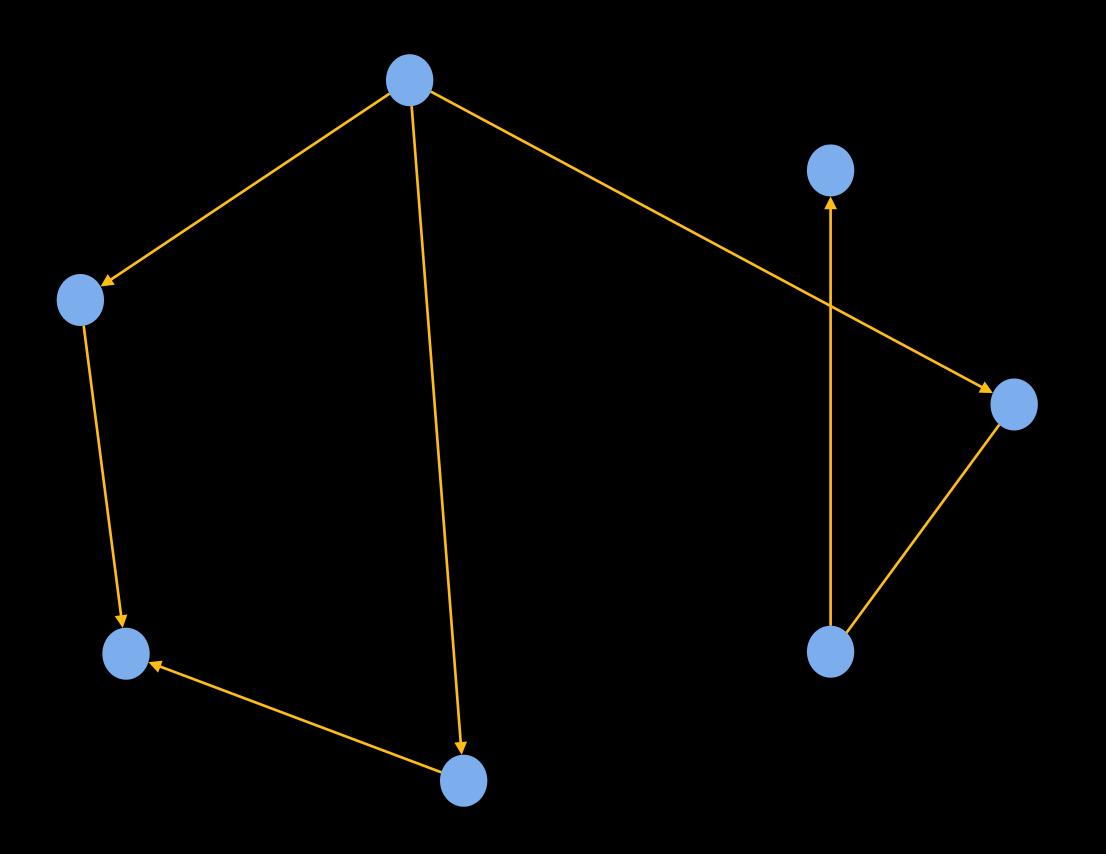
# SENTENCES AS A GRAPH

How are you? How could you? How am I doing?



# SENTENCES

How are you? How could you? How am I doing?



#### GRAPHS

Most impressive and immersive representation of combinatorics is Graph

A graph can have cycles, and multiple paths between two nodes.

Non-linear data structure

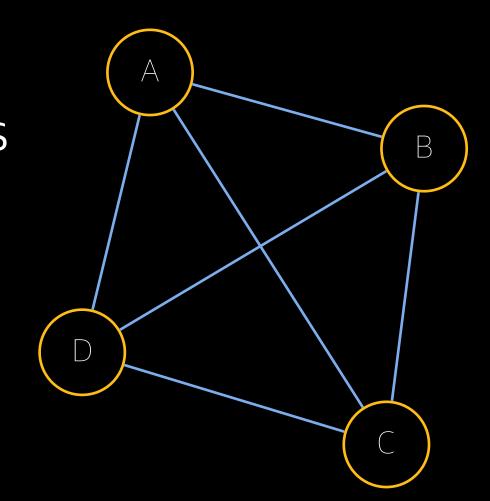
#### MODELLING A PROBLEM

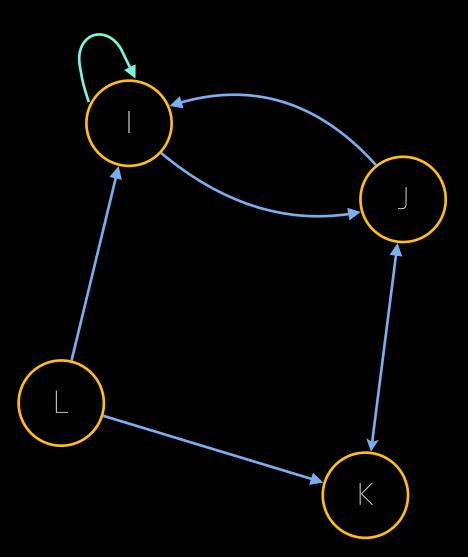
- Representing the information of the problem
- → How to use the information?
- How do I manipulate the information?
- We need some abstract notation and structures to model the problem
- Identify key elements and relationships
- What data structure?
- Any mathematical representation?

Modelling a problem is the process of creating a simplified representation of a real-world situation. This representation, called a model, helps us understand, analyze, and potentially solve the problem more effectively

## FORMAL DEFINITION

- A graph is a pair of vertices (or nodes) connected by edges
- Represented as G = (V, E)
- $V = \{v_1, v_2, \dots, v_n\} \text{ or } v \in V$
- and  $E=\{e_1,e_2,\cdots,e_m\}$  or  $e\in E$
- $E \subseteq \{(u,v) \mid u,v \in V\}$
- n = |V| number of vertices
- m = |E| number of edges

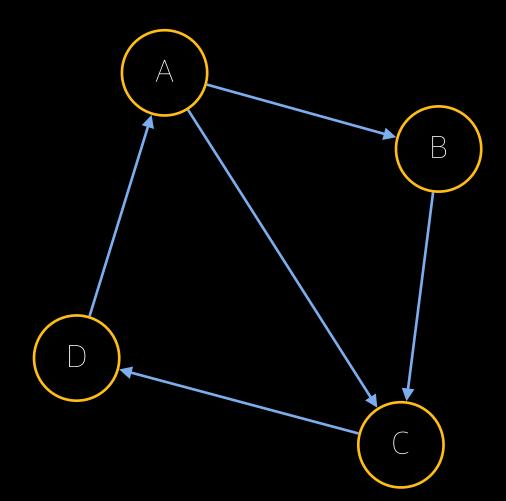




# FORMAL DEFINITION

• A directed graph or (digraph) is a pair G = (V, E) where V is a set of vertices (or nodes), and  $E \subseteq V \times V$  is a set of directed edges.



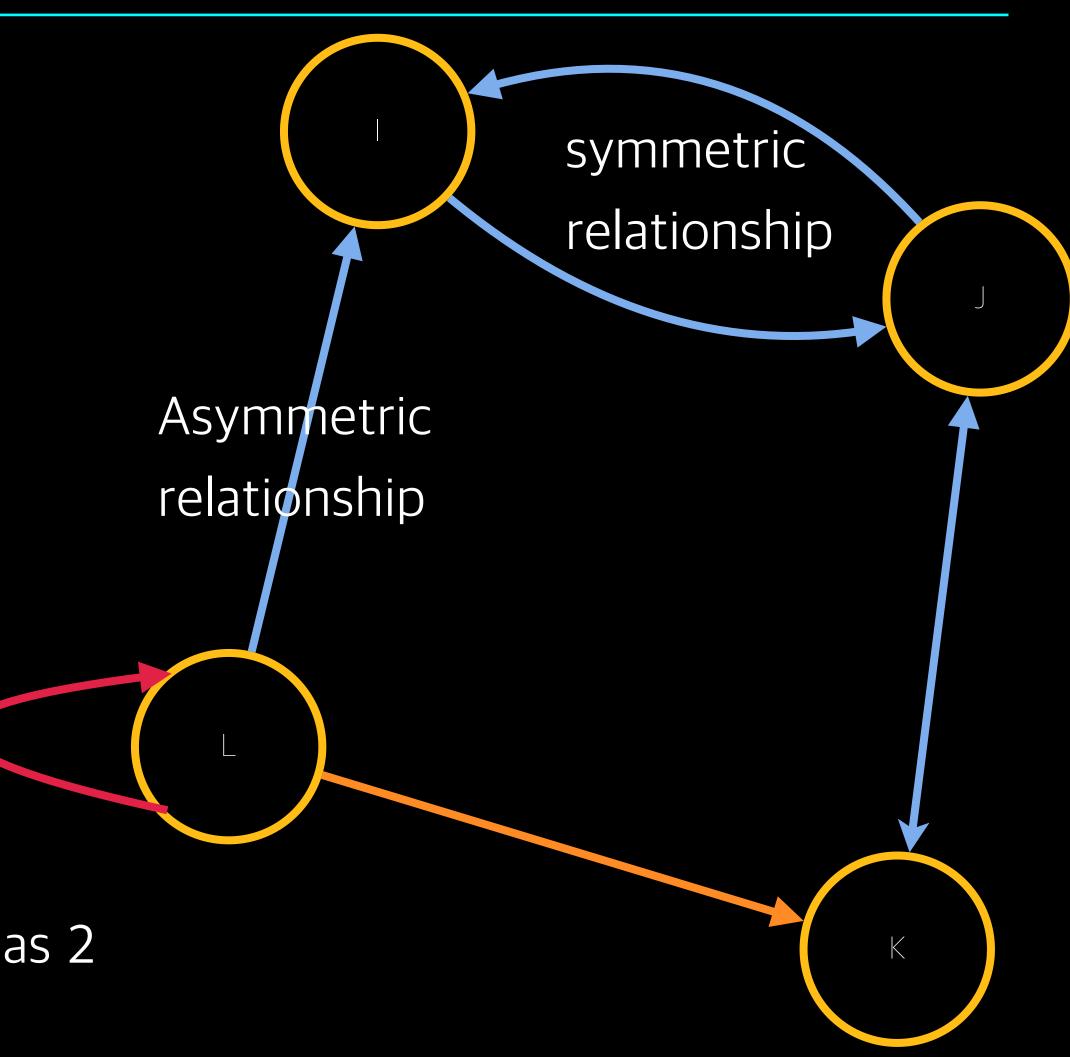


# DIRECTED GRAPH(DIGRAPH)

A directed graph of digraph is a pair G = (V, E) where

- lacktriangledow V is a set of vertices and
- **♦** E is a set of directed edges  $E \subseteq V \times V$  and  $0 \le E \le V(V-1)$
- Edges are ordered pairs of vertices
  - If vertices u and v are connected, then there is a directed edge from u to v

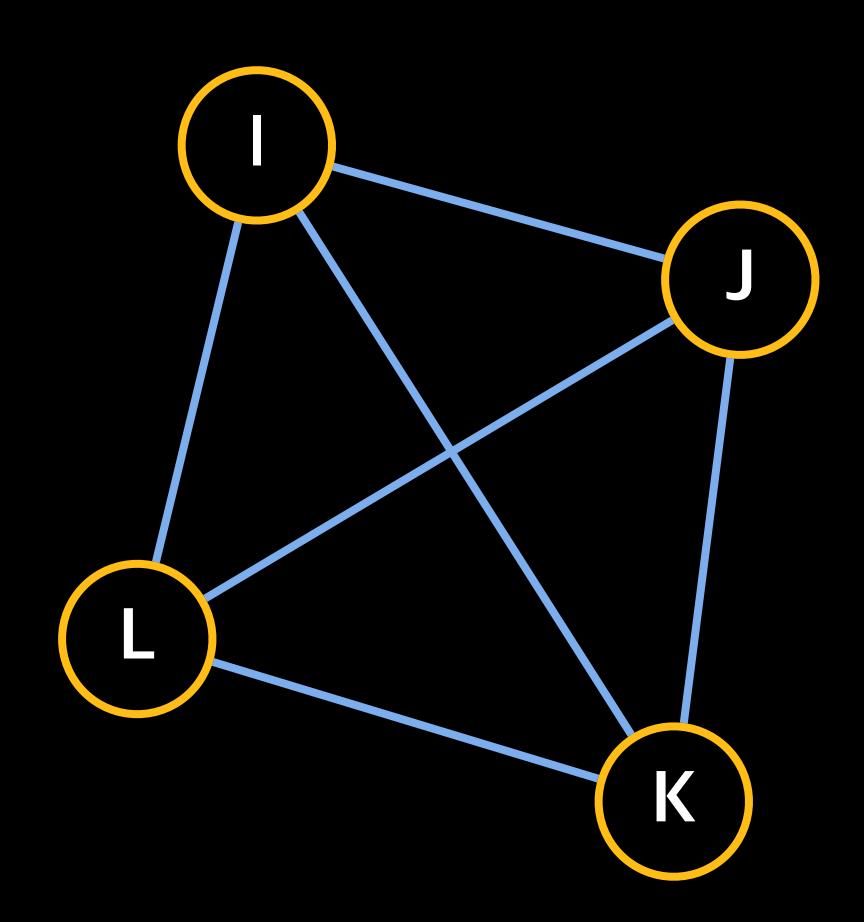
A digraph can have self loops (u, u) and it is counted as 2 It can have at the most  $n^2$  edges



## UNDIRECTED GRAPH

A undirected graph is a pair G = (V, E) where

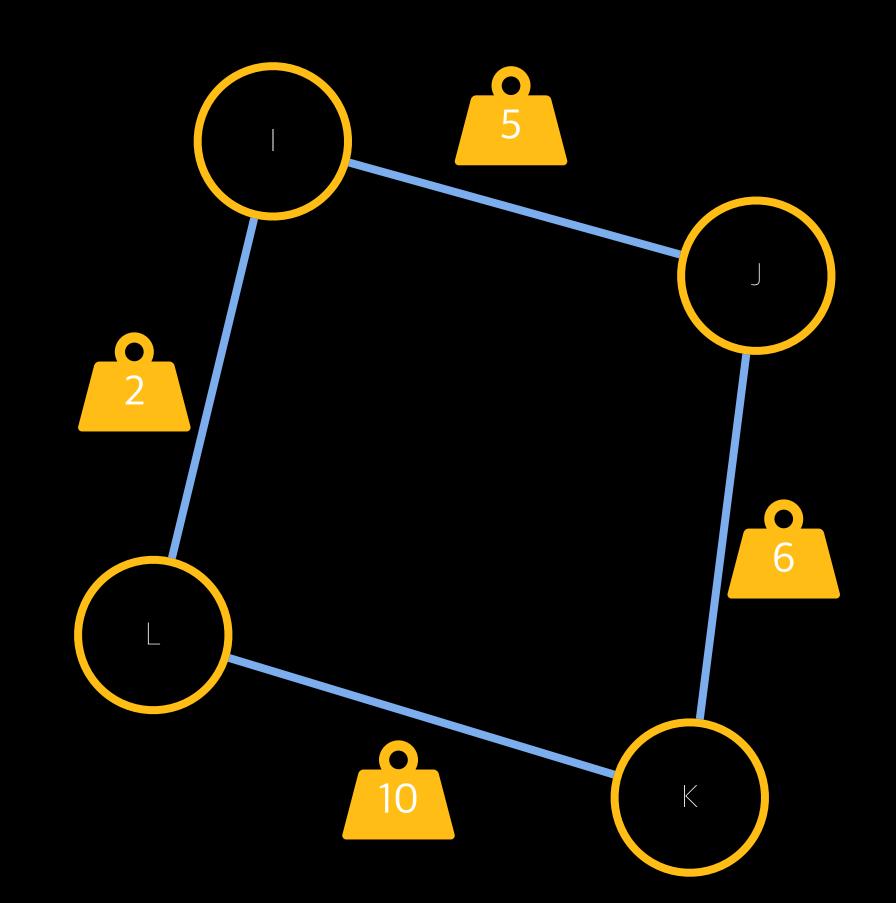
- lacktriangledow V is a set of vertices and
- lacktriangle Undirected graph can have at most  $\frac{n(n-1)}{2}$  edges
- $\diamond$  A undirected graph cannot have self loops (u, u).



# WEIGHTED GRAPH

A weighted graph is a pair G = (V, E) where

- lacktriangledow V is a set of vertices and
- lacktriangledown E is a set of edges  $E \subseteq V \times V$
- lacktriangle Every  $e \in E$  will be associated with a number



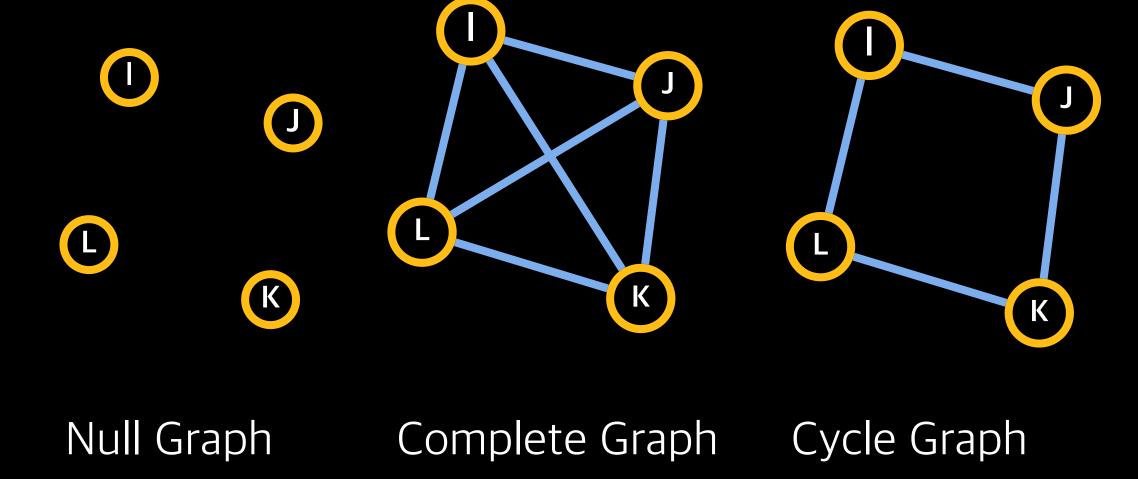
#### TERMINOLOGY

- ♦ A vertex u is a neighbour of (adjacent to) a vertex v in a graph G = (V, E), if there is an edge  $\{u, v\} \in E$ . Here u is a neighbour of v and vice versa. It is written as  $u \sim v$
- ♦ In directed graphs, for any directed edge  $u \to v$ , we call u a **predecessor** of v, and we call v a **successor** of u.
- **↑** The <u>neighbourhood</u> of a vertex  $v \in V$  is its set of all neighbours of v:  $v_{nh} = \{u \mid \{u, v\} \in E\}$
- lacktriangle The <u>degree</u> of a vertex v is the size of the neighbourhood  $v_{nh}$
- ◆ Edge is **incident** on a vertex if the vertex is one of its endpoints. A vertex is incident on an edge if it is one of the endpoints of the edge
- ♦ A directed graph is strongly connected if all vertices are reachable from all other vertices
- lacktriangle In an undirected graph, edge (u, v)always implies(v, u)

#### TERMINOLOGY

- lacktriangle A <u>walk</u> is a sequence of vertices  $v_1, v_2, \cdots, v_k$  such that  $\forall i \in [1, 2, \cdots, k-1, (v_i \sim v_{i+1})]$
- lacktriangle A **path** is a walk where  $v_i \neq v_j$ ,  $\forall i \neq j$  A path is a walk that visits each vertex at most once
- $\bullet$  A <u>closed walk</u> is a walk where  $v_1 = v_k$ .
- lacktriangle A **cycle** is a **closed path**, i.e. a path combined with the edge  $(v_k, v_1)$ . An undirected graph is connected if all vertices are reachable from all other vertices
- $\bullet$  A vertex v is **reachable** from a vertex u, if there is a path starting at v and ending at u
- ◆ A graph <u>acyclic</u> if no subgraph is a cycle
- ◆ An undirected graph with no cycles is a forest and if it is connected it is called a tree
- ◆ A **tree** is a connected acyclic graph

# EXAMPLES



## DATA STRUCTURE OF A GRAPH

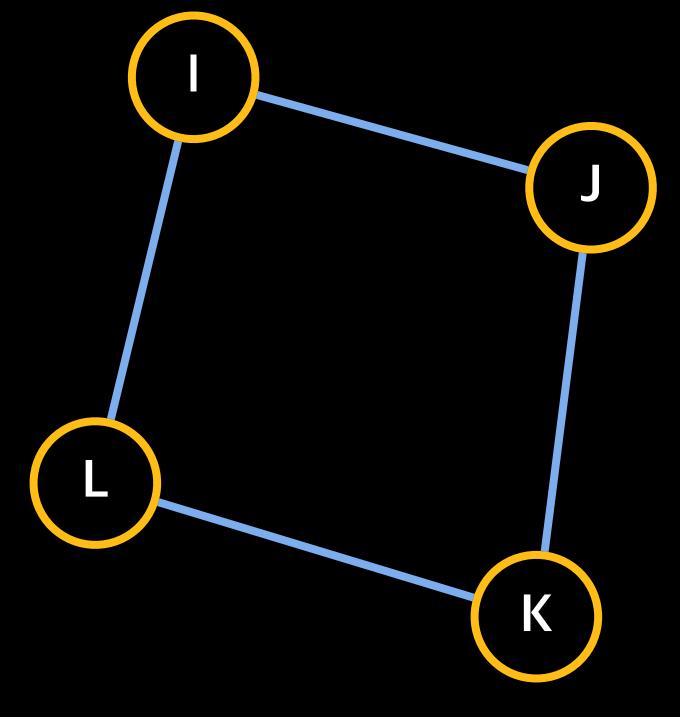
- ♦ A diagrammatic representation using points joined by and lines
- How do we store it in the computer?
- ◆ Can we list vertices adjacent to each vertex of the graph as a dictionary/hash map?

```
G = {

"I":["J", "L"],

"K":["J","K"],

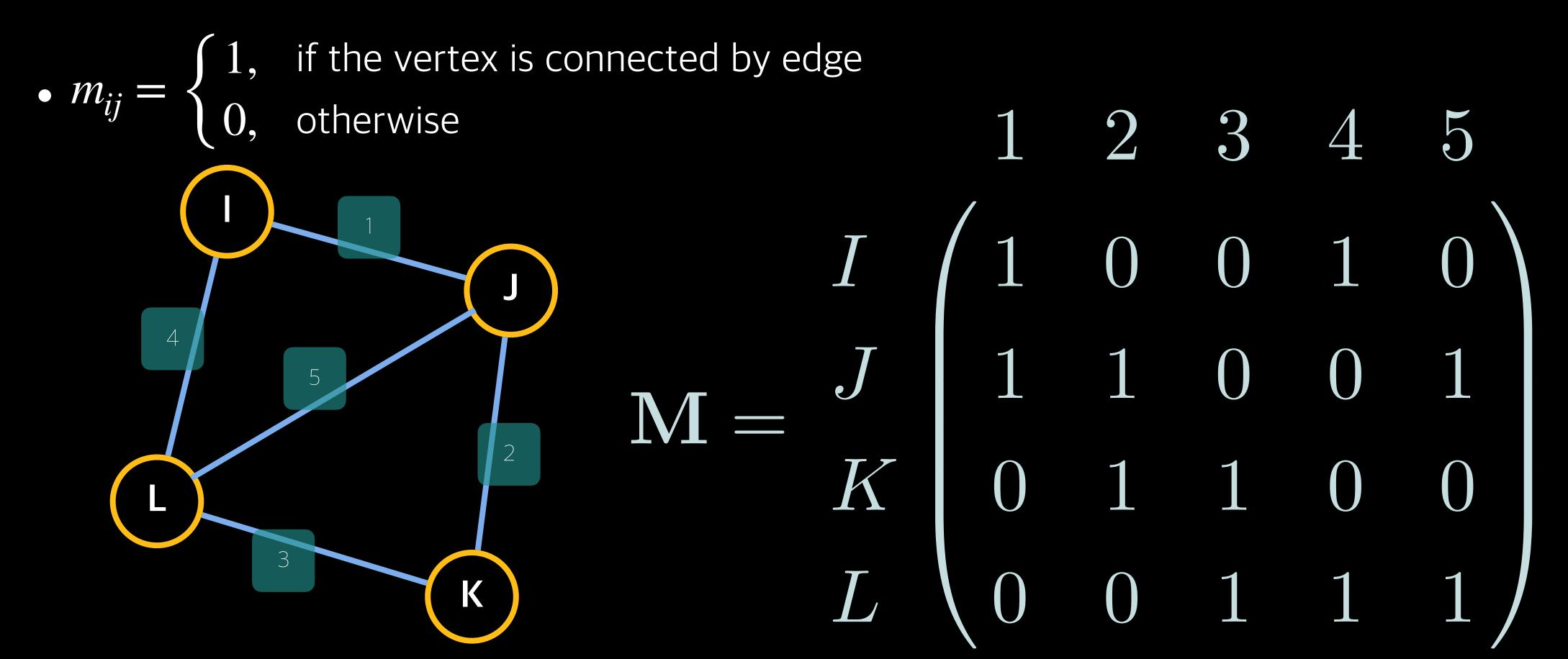
"L":["I","K"]
```



Any other representation?

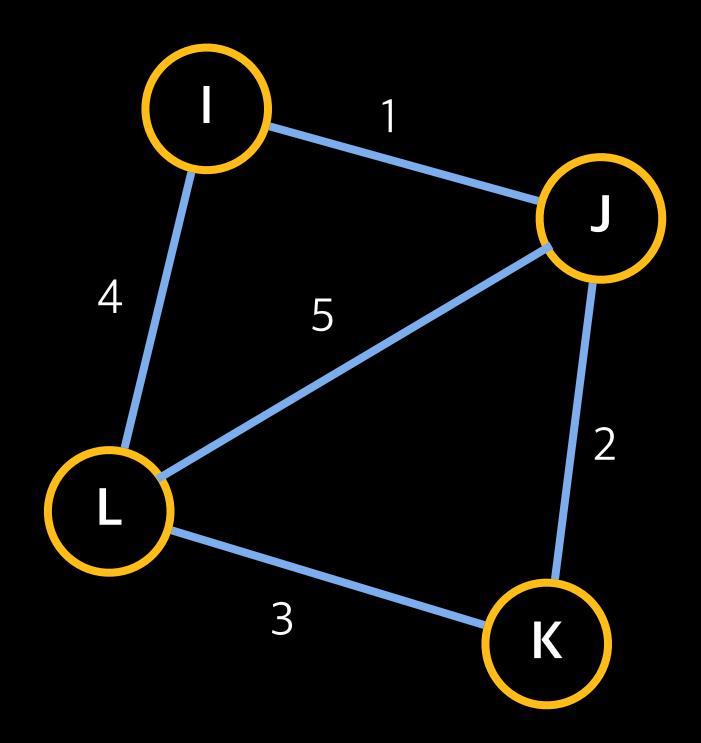
#### INCIDENCE MATRIX

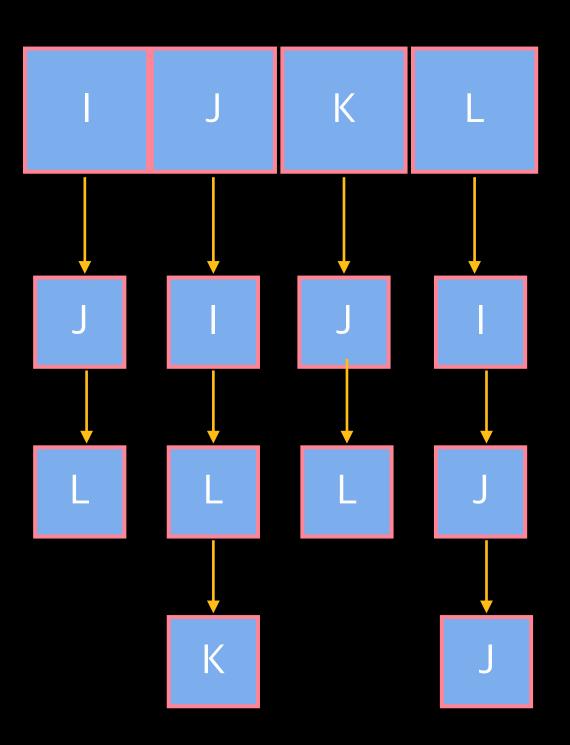
• Incidence Matrix M is a  $n \times m$ 



# ADJACENCY LIST

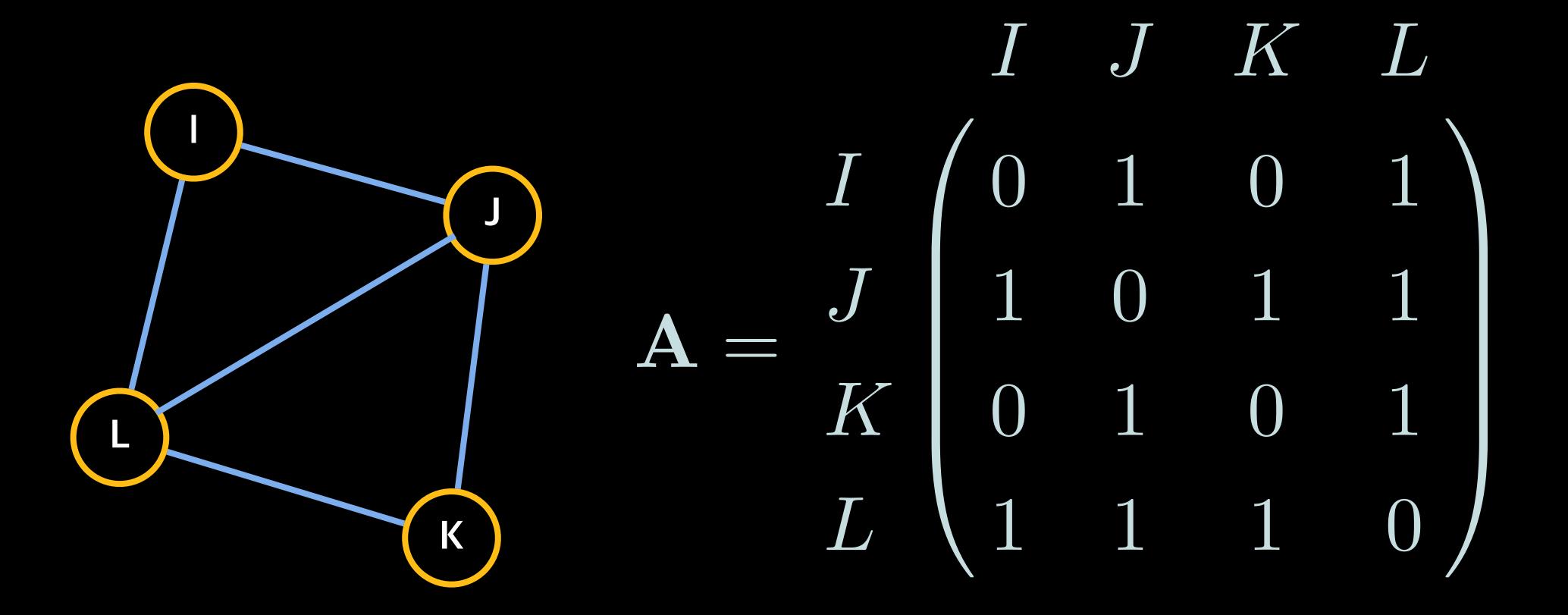
Represents pairwise connections between vertices





#### ADJACENCY MATRIX

- Represents pairwise connections between vertices
- ullet Square Matrix, A, whose  $ij^{th}$  entry is the number of edges joining vertex i and vertex j



## PROPERTY OF THE ADJ MATRIX

For all undirected graphs, the adjacency matrix is always symmetric  $a_{ij}=a_{ji} \forall i,j$  and

$$a_{ij} = 0$$
 when  $i = j$ 

For digraphs, the adjacency matrix may or may not be symmetric, and the diagonal entries may or may not be zero.

Time Complexity of finding connectivity of any two vertices is O(1)

Time Complexity of listing all the neighbours is O(n)

Space complexity is  $O(n^2)$ 

If the application required only to find the maximum number of edges in a graph, what data structure would you choose?

# SUMMARY OF COMPLEXITY

Adjacency Matrix	Faster to add or remove modes Faster to check if a node exists Better for dense graphs - $ E  \approxeq O( V ^2)$ Storage = $= O( V ^2)$
Edge List	Faster if the graph is sparse Better for sparse graphs - $ E  \approxeq O( V )$ Storage = $= O( V  +  E )$

	Standard adjacency list	Adjacency
	(linked lists)	matrix
Space	$\Theta(V+E)$	$\Theta(V^2)$
Test if $uv \in E$	$O(1 + \min\{\deg(u), \deg(v)\}) = O(V)$	O(1)
Test if $u \rightarrow v \in E$	$O(1 + \deg(u)) = O(V)$	O(1)
List $ u$ 's (out-)neighbors	$\Theta(1 + \deg(v)) = O(V)$	$\Theta(V)$
List all edges	$\Theta(V+E)$	$\Theta(V^2)$
Insert edge <i>uv</i>	O(1)	O(1)
Delete edge <i>uv</i>	$O(\deg(u) + \deg(v)) = O(V)$	O(1)

Times for basic operations on standard graph data structures

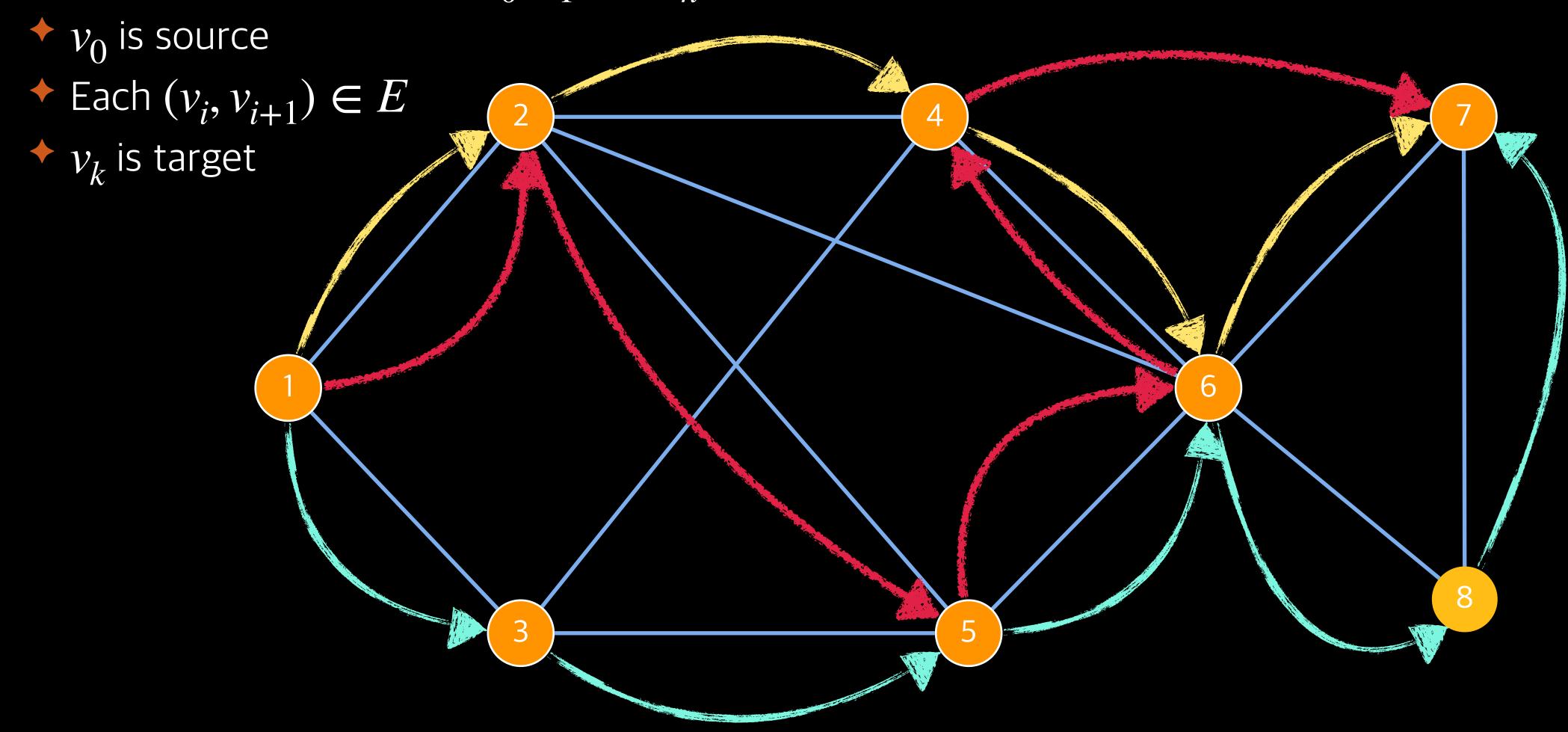
# OPERATIONS ON A GRAPH

#### Basic operations

- Search for a node
- ◆ Insert a node
- Delete a node
- Degree of a node
- Number of vertices
- Number of edges
- Most connected node

# FINDING A PATH

Find a sequence of vertices  $v_0, v_1, \dots, v_n$  such that



## FINDING A PATH

- ♦ In general, start at the start node and follow the edges to find its neighbours
- Mark vertices that have been visited
- ★ Keep track of vertices whose neighbours have already been explored
- Avoid going round indefinitely in circles
- ◆ Two approaches Breadth first and Depth First known as BFA and DFS
- ◆ BFS Start at the first node, visit all the direct neighbours recursively
- DFS Visit all the way down in one potential path, then backtrack to visit all potential paths

#### BREADTH-FIRST SEARCH

- Start from i, visit a neighbour j
   Suspend the exploration of i and explore j instead
- Continue till a vertex with no unexplored neighbours is reached
- ◆ Backtrack to nearest suspended vertex that still has an unexplored neighbour
- → Suspended vertices are stored in a stack
- ◆ Last in, first out: most recently suspended is checked first

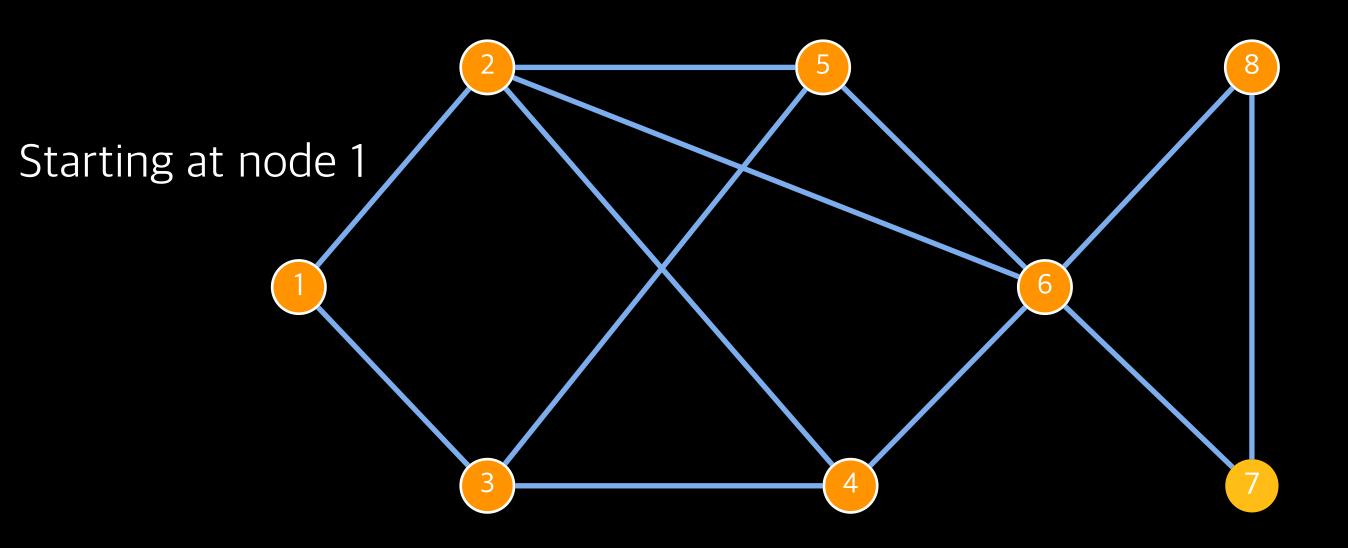
#### BREADTH-FIRST SEARCH

- Explore the graph level by level
- Visit all the neighbourhood that are one-step away or nearest neighbour
- Then second level two steps

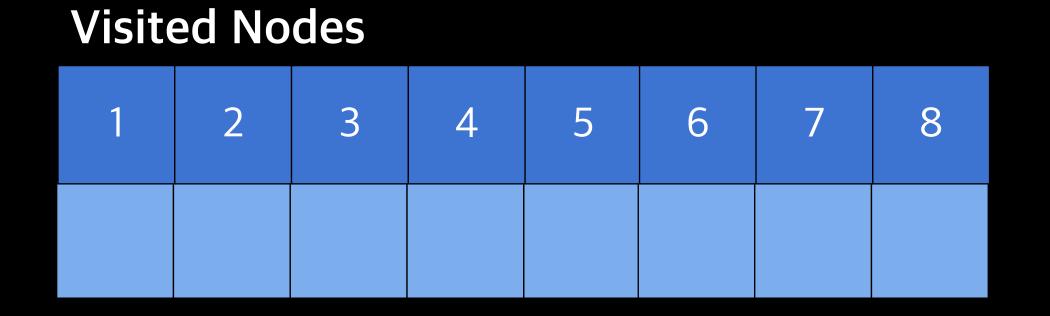
...

- Memorise all the visited vertices
- ★ Keep track of vertices visited, but whose neighbours are unexplored/investigared

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

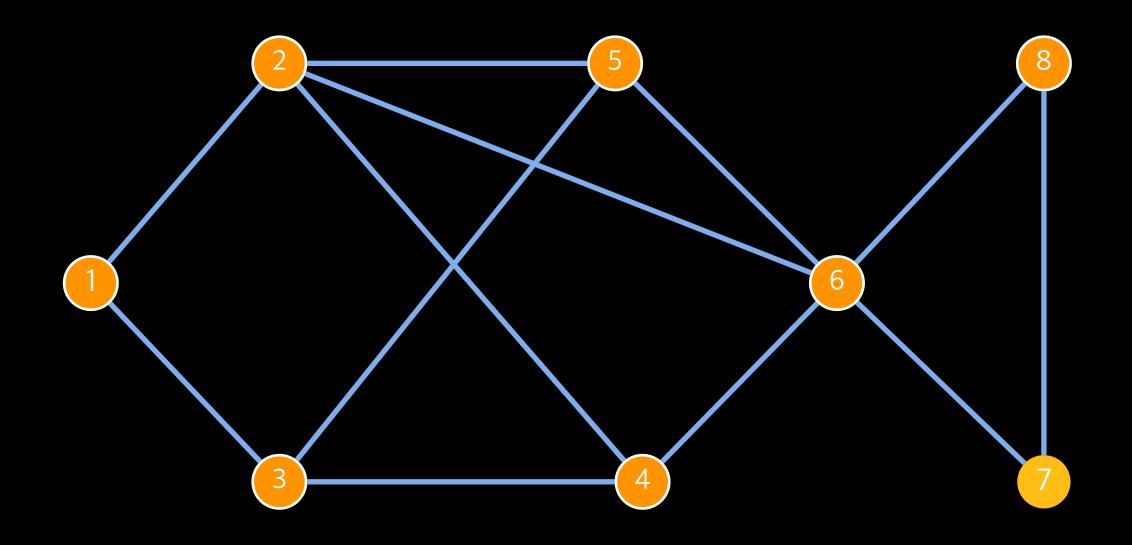


# Queue

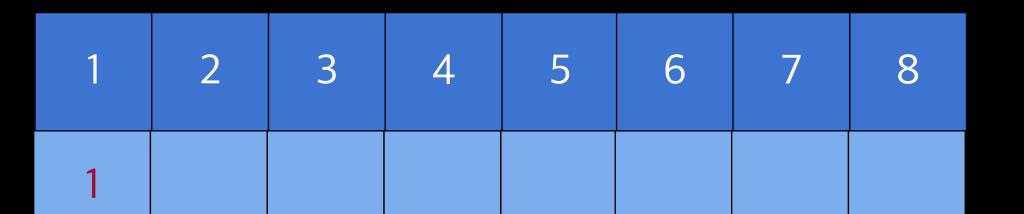


**Visited Nodes** 

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

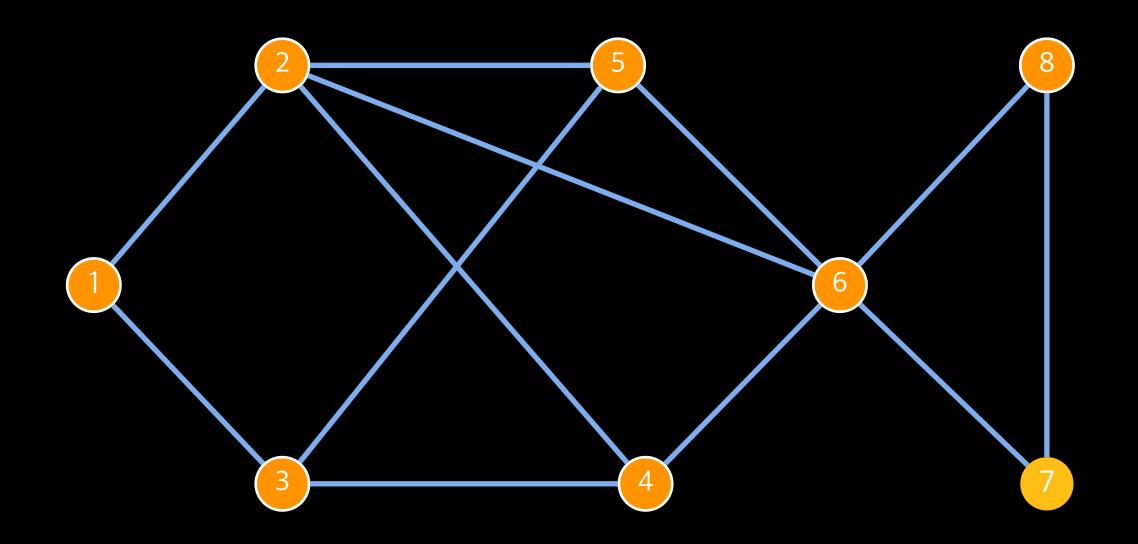


#### Queue



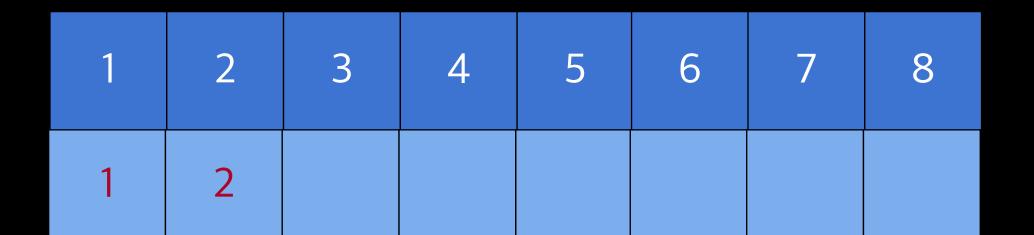
**Visited Nodes** 

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

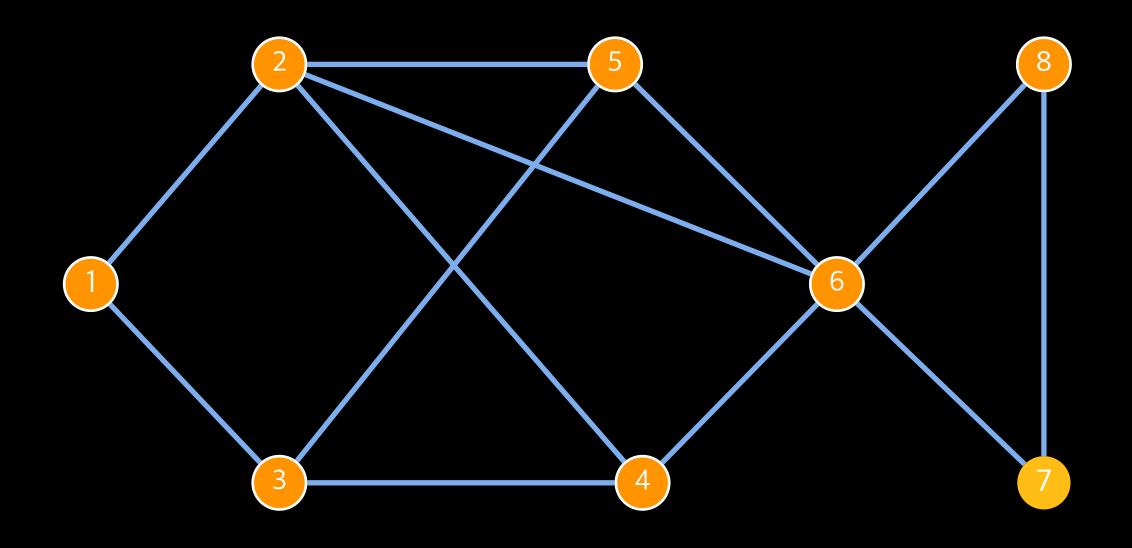


#### Queue

2



for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue



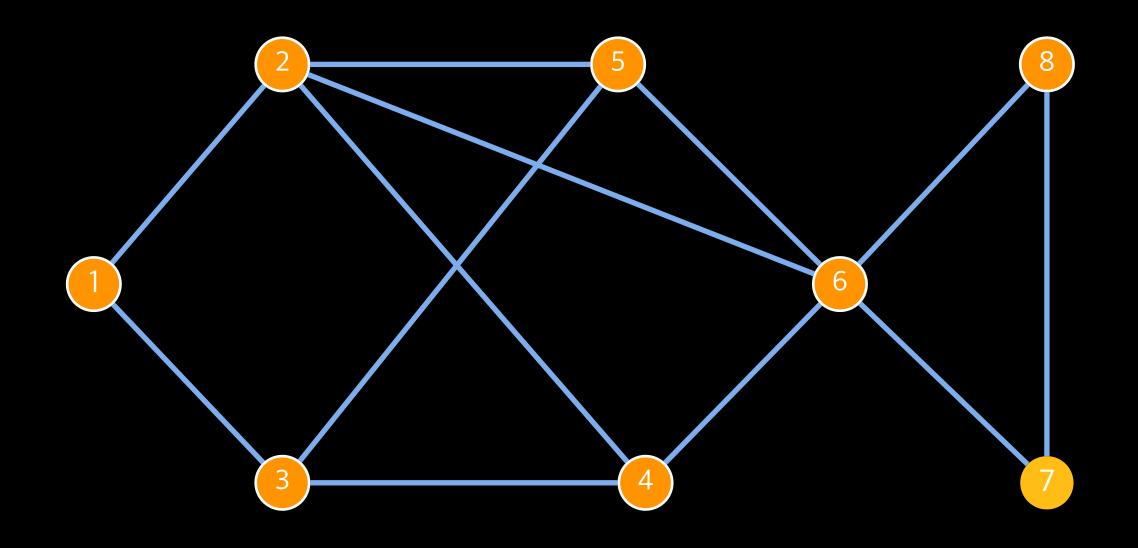
#### Queue

2	3			

#### **Visited Nodes**

1	2	3	4	5	6	7	8
1	2	3					

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue



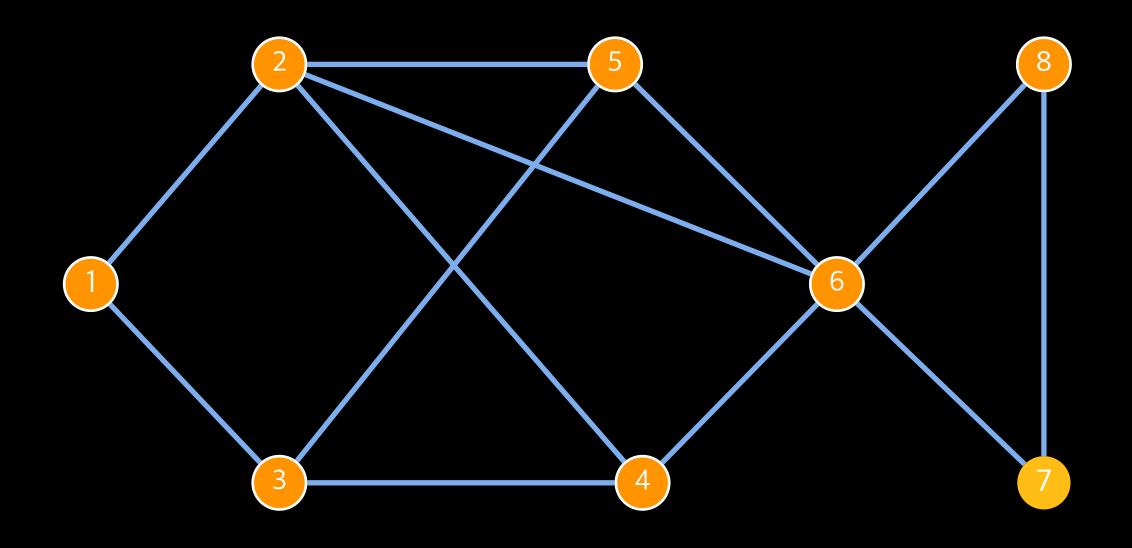
#### Queue

3 4



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

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue



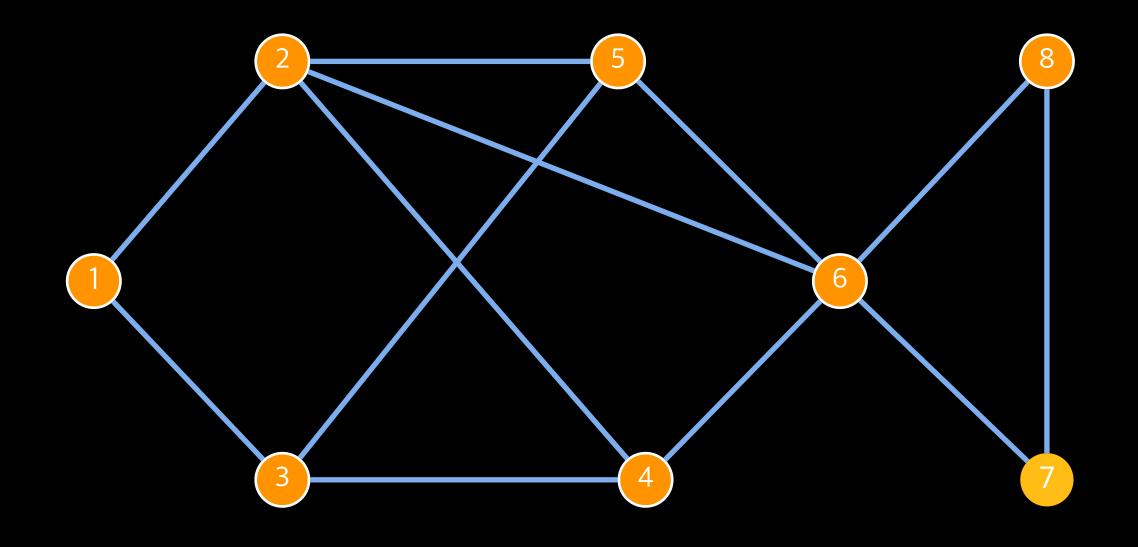
### Queue

3 4 5



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

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

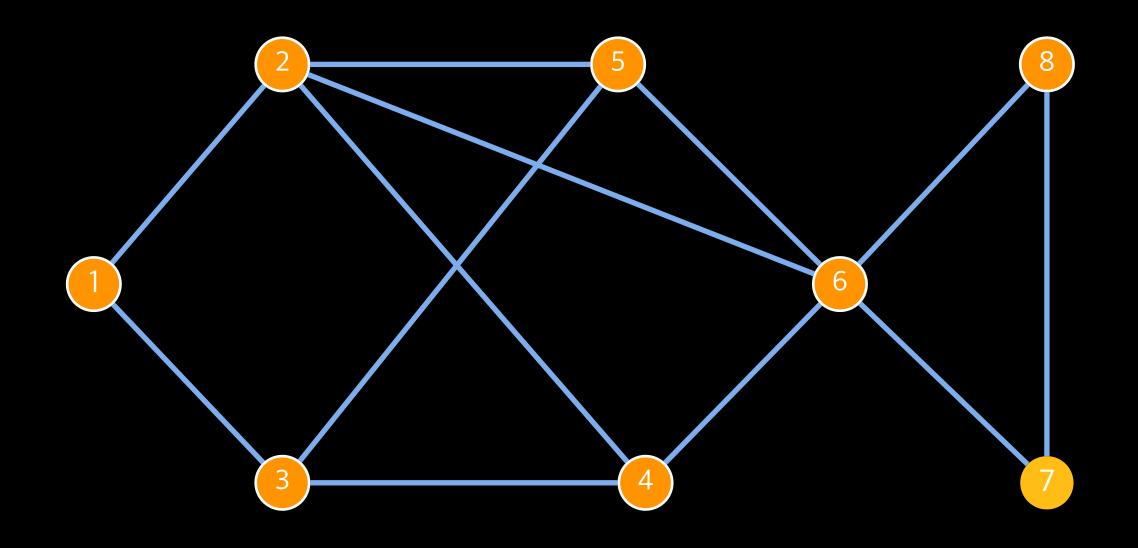


### Queue

3	4	5	6		

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

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue



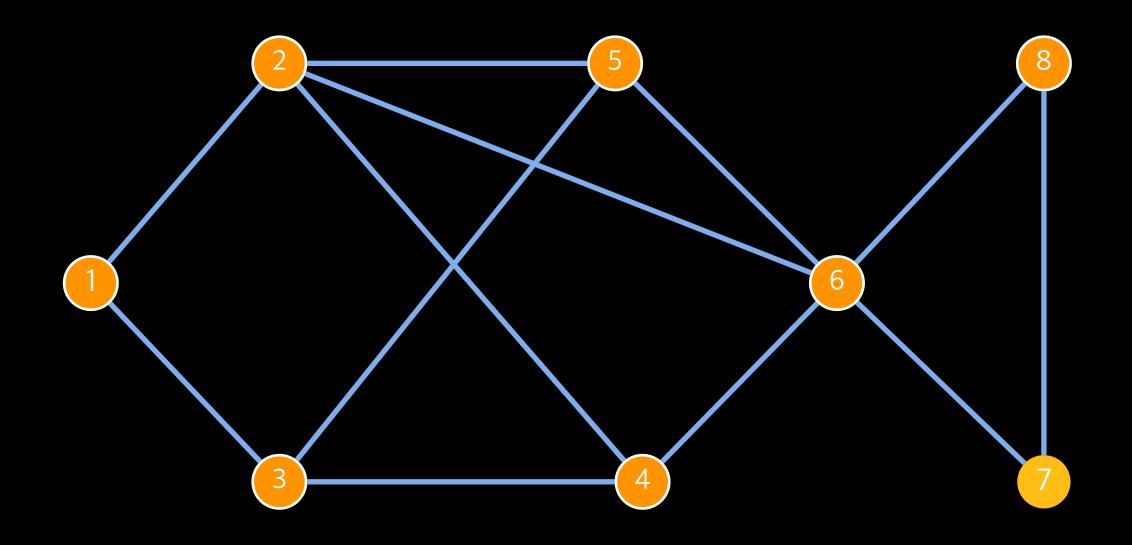
### Queue

4 5 6



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

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

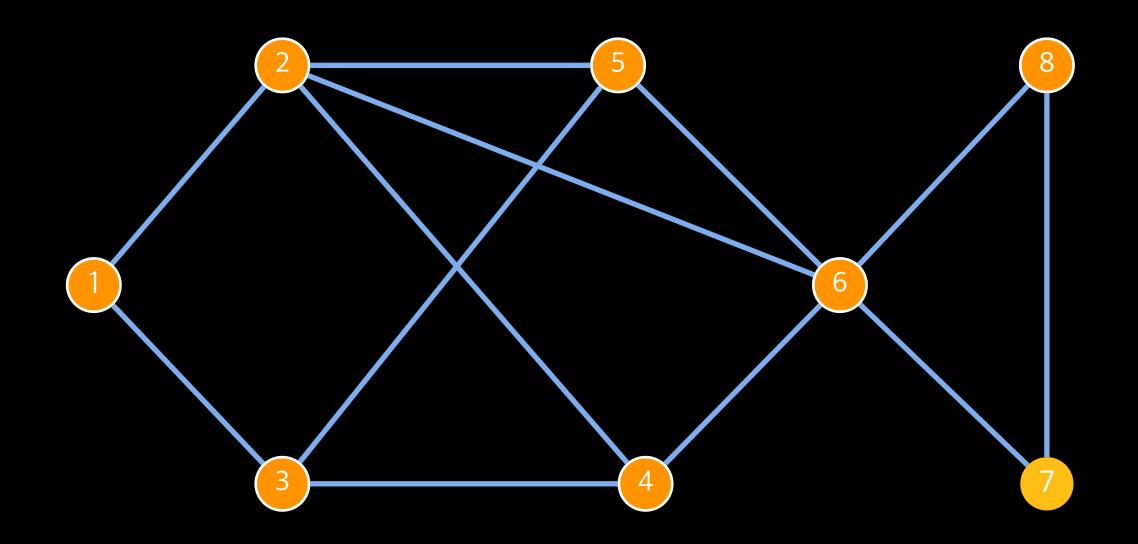


### Queue

5 6

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

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue



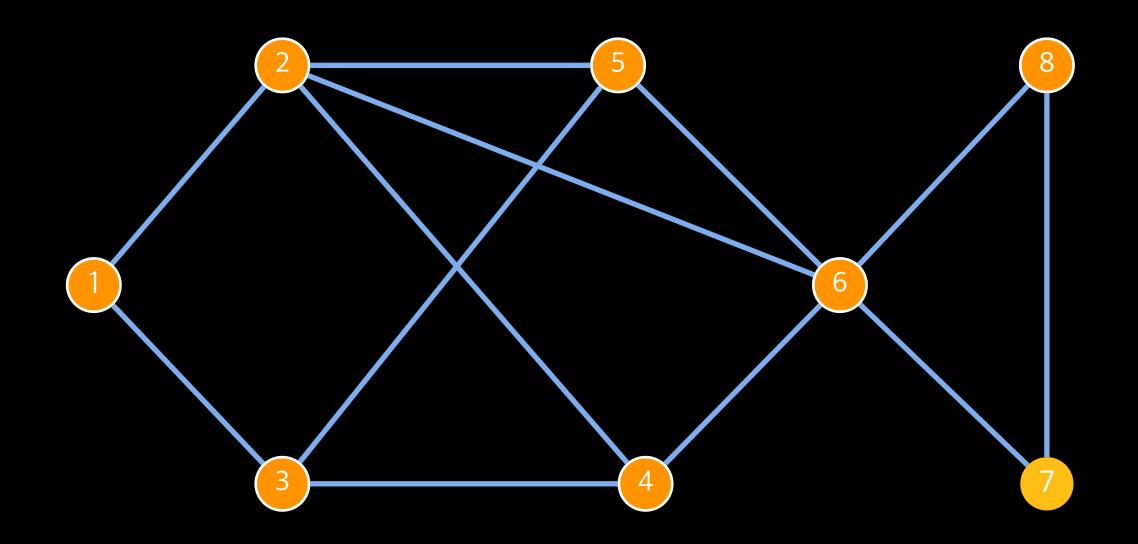
## Queue

6

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

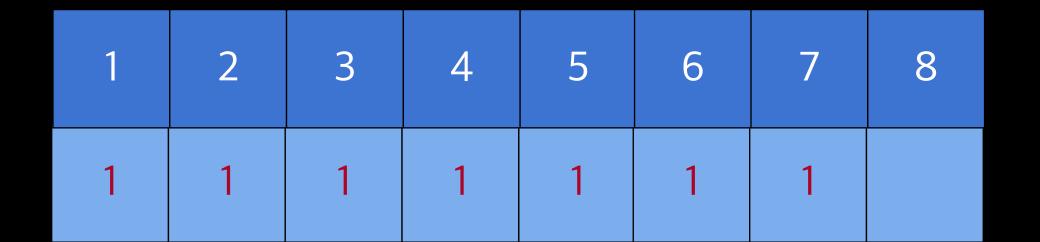
**Visited Nodes** 

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

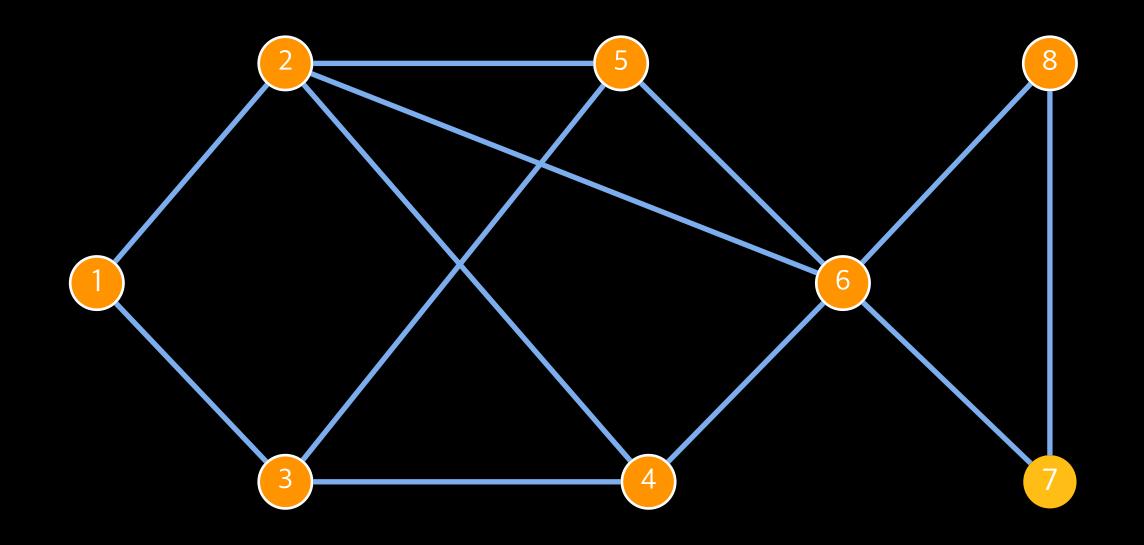


## Queue

7



for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

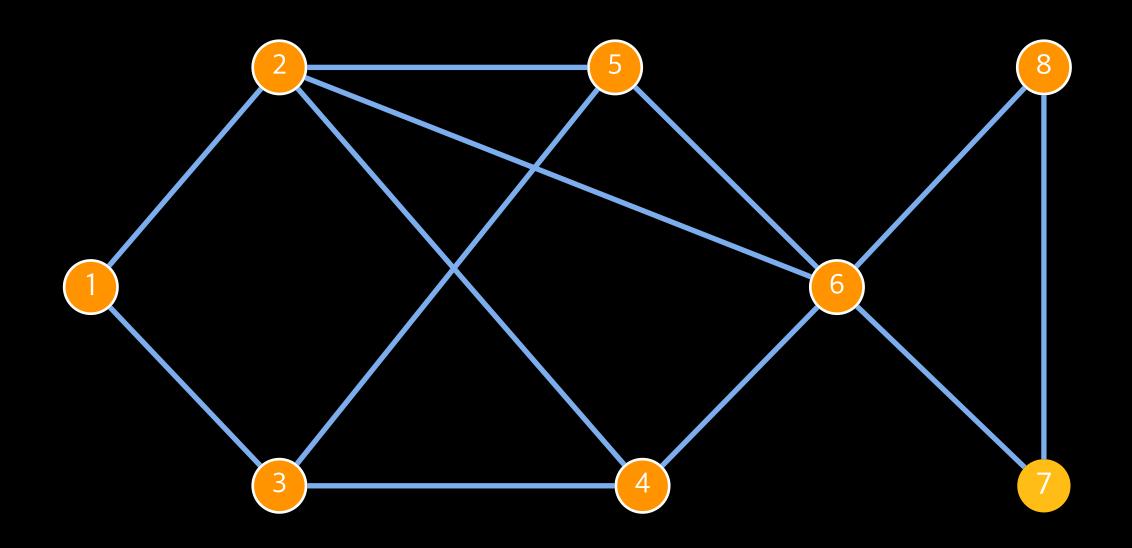


### Queue

7	8			

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

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

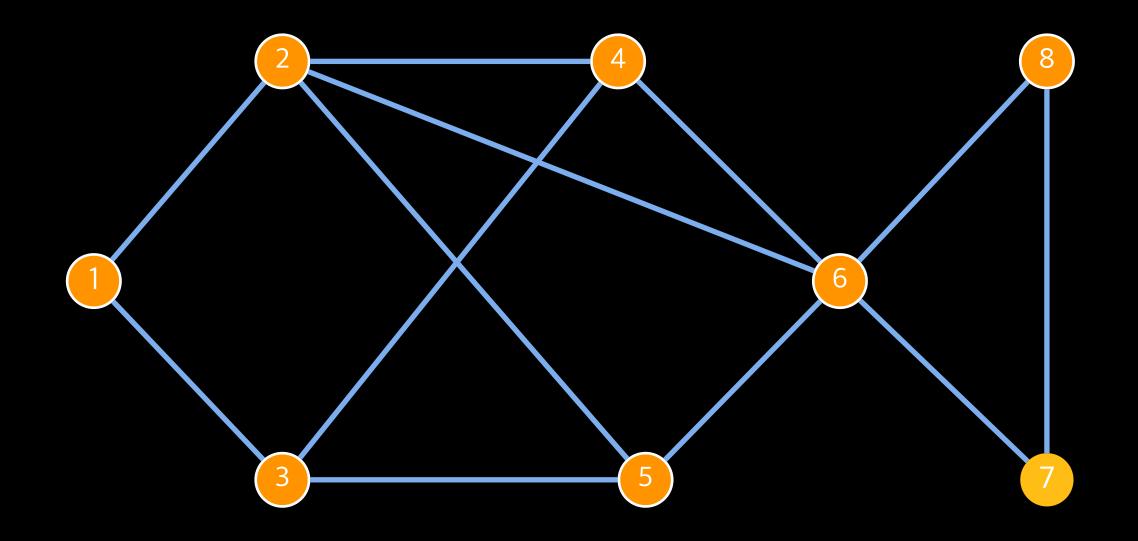


## Queue

8

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

for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue



### Queue

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

## DFS ALGORITHM

```
function breadth_first_search(start_node)
   visited \leftarrow a set to store references to all visited nodes
  queue \leftarrow a queue to store references to unexplored nodes
  queue.enqueue(start_node)
  visited.append(start_node)
   While queue \neq \phi
      current\_node \leftarrow queue.dequeue()
      process(current_node)
      For neighbour ∈ current node neighbours
          If neighbor \notin visited
            queue .enqueue(neighbour)
            visited.append(neighbour)
```

## COMPLEXITY OF BFS

### **Runtime Analysis**

- lacktriangle Let |V| and |E| be the number od vertices and edges, respectively
- Every  $v \in V$  is enqueued and processed exactly once, resulting in O(|V|) time
- $\buildrel$  Every edge is checked exactly once in the for loop resulting in  $O(\,|E\,|\,)$  time

The Complexity is O(|V| + |E|)

## COMPLEXITY OF BFS

### **Runtime Analysis**

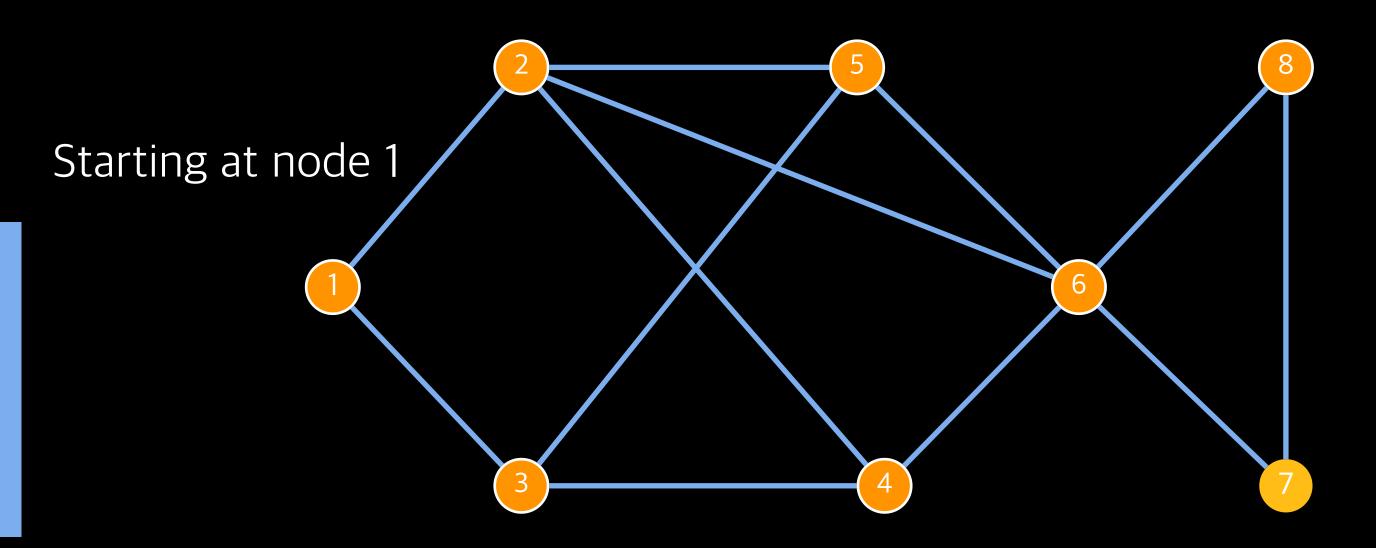
- lack Each vertex enters Q exactly once
- lack If the graph is connected, while loop to check Q iterated n times
- $lack {lack}$  For each j extracted from Q, need to examine all neighbours of j
- lacktriangle Rows are scanned n times in the adjacency matrix. The complexity is  $O(n^2)$

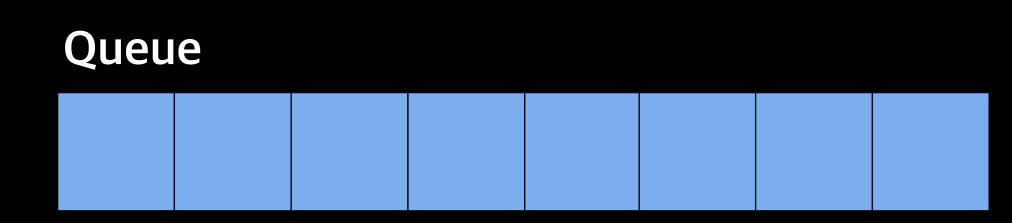
# BFS - FIND THE PATH

Can we identify the path traversed from the DFS Algorithm present in slide #

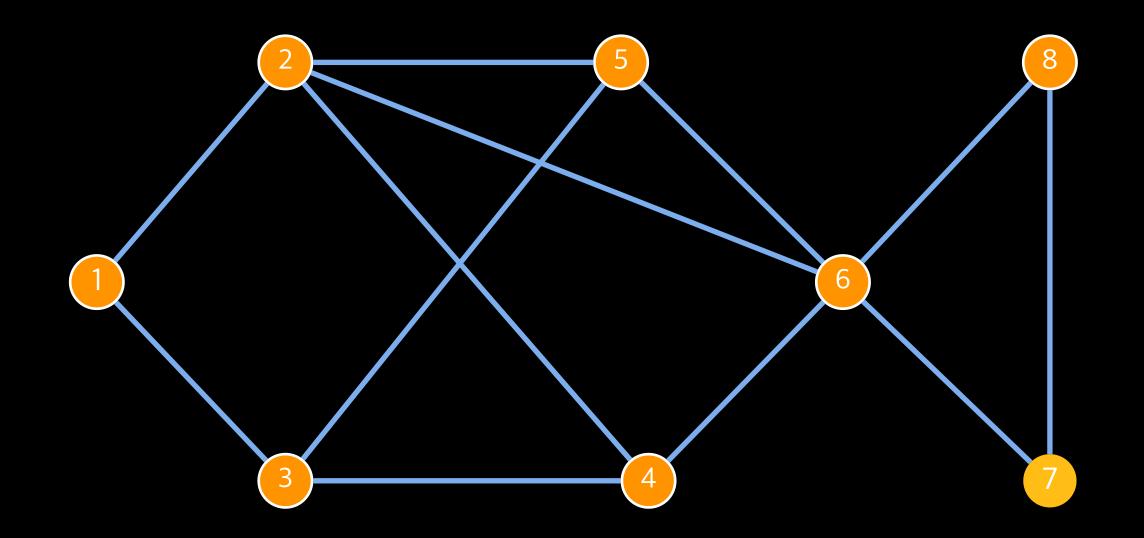
# BFS - FIND THE PATH

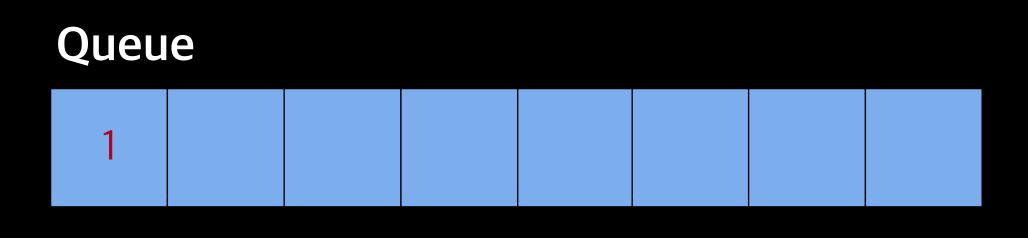
Can we identify the path traversed from the DFS Algorithm present in slide #

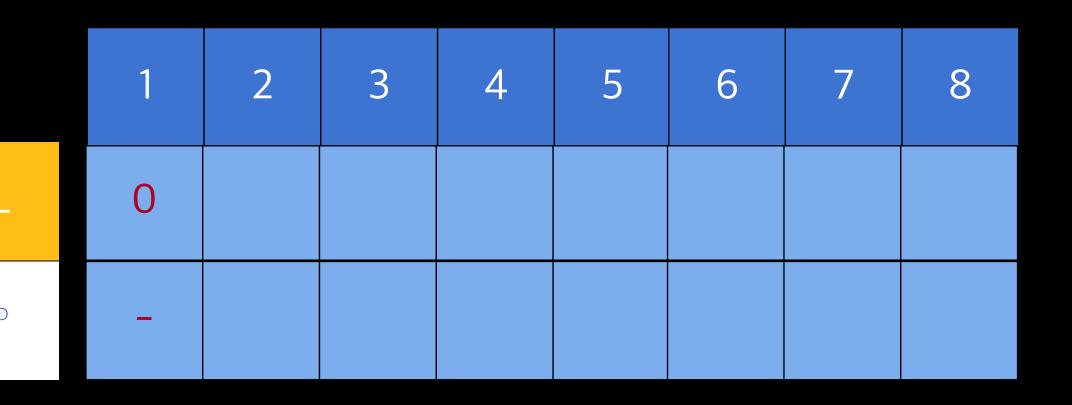


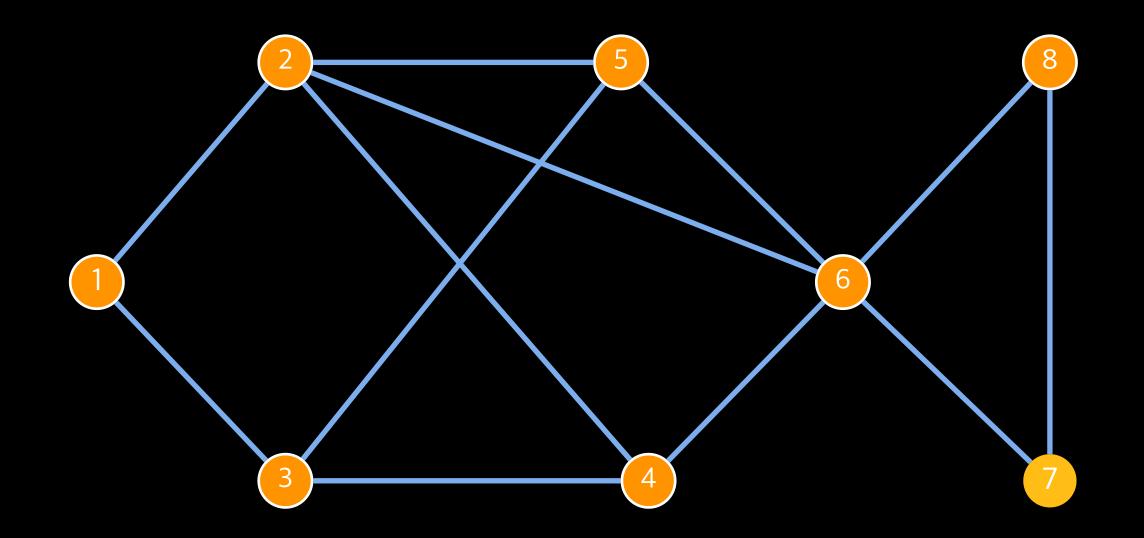






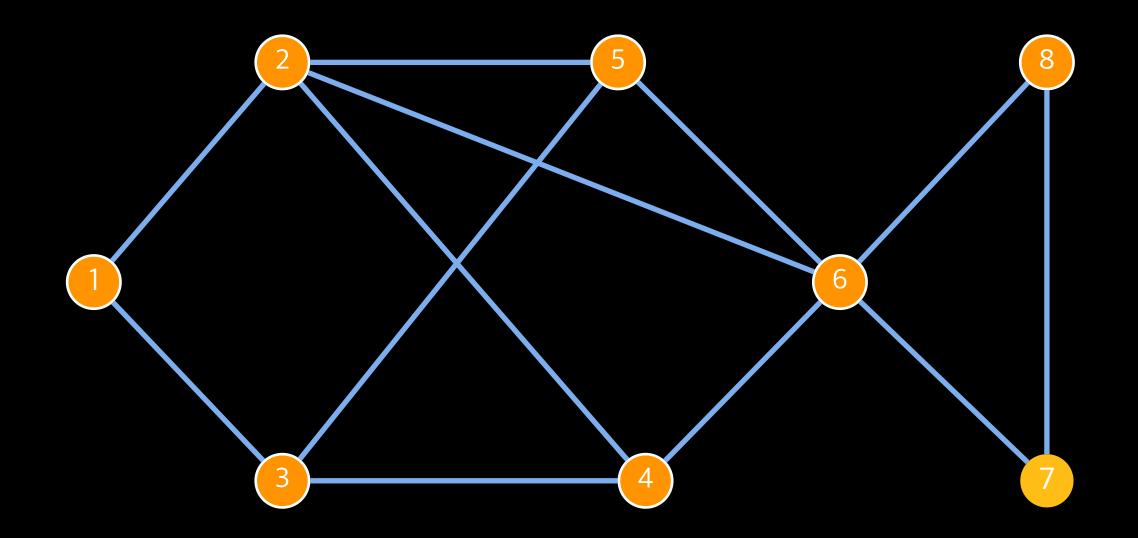




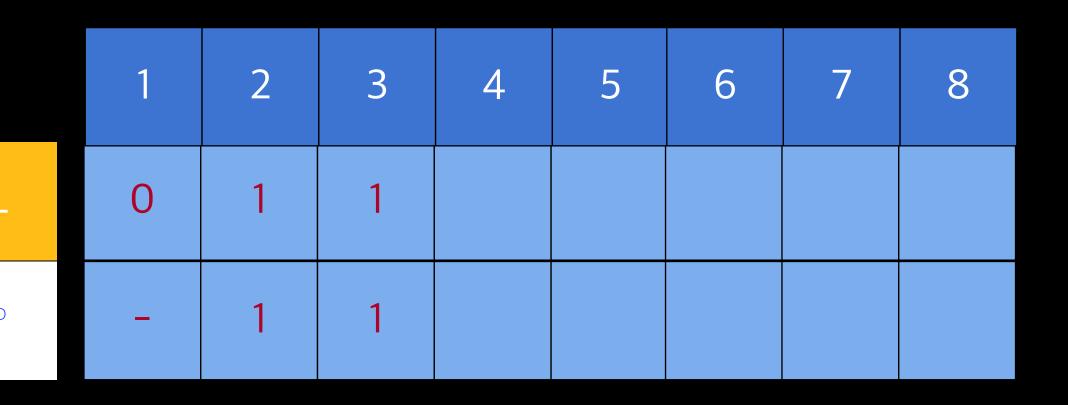


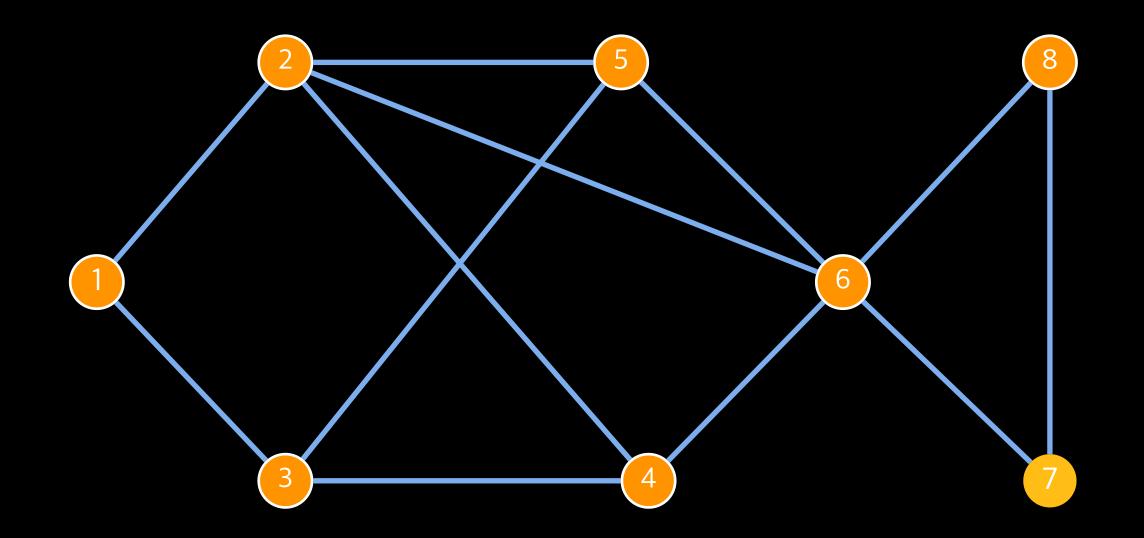


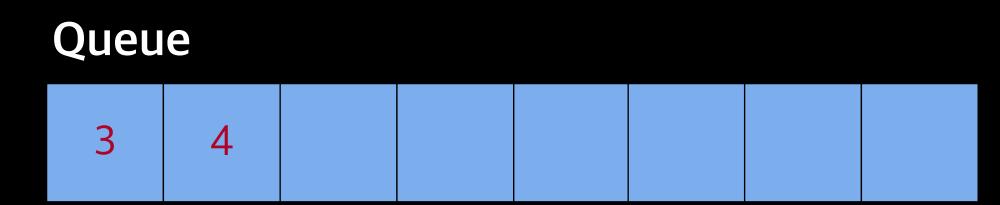




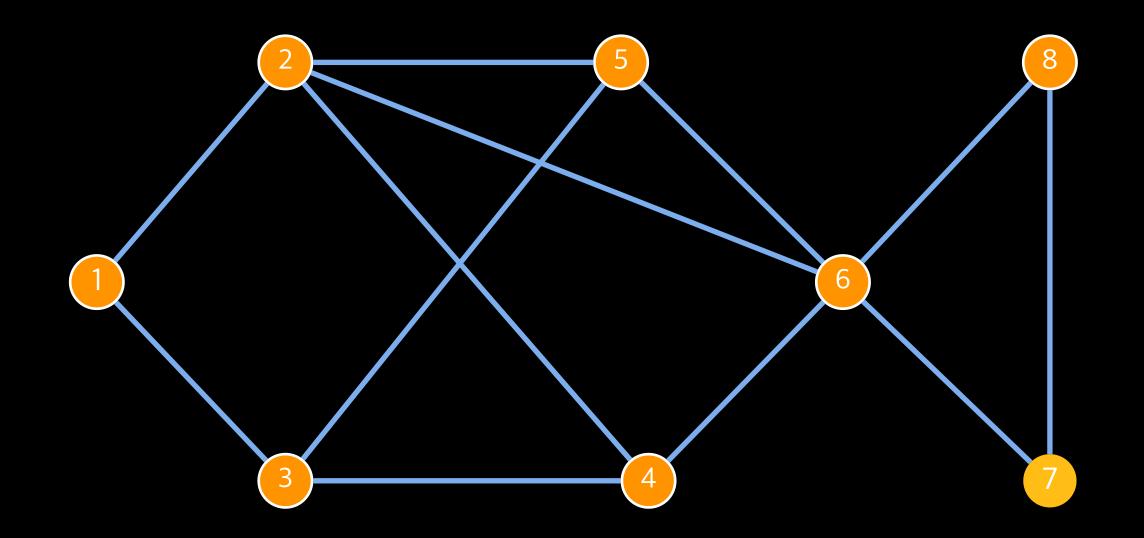






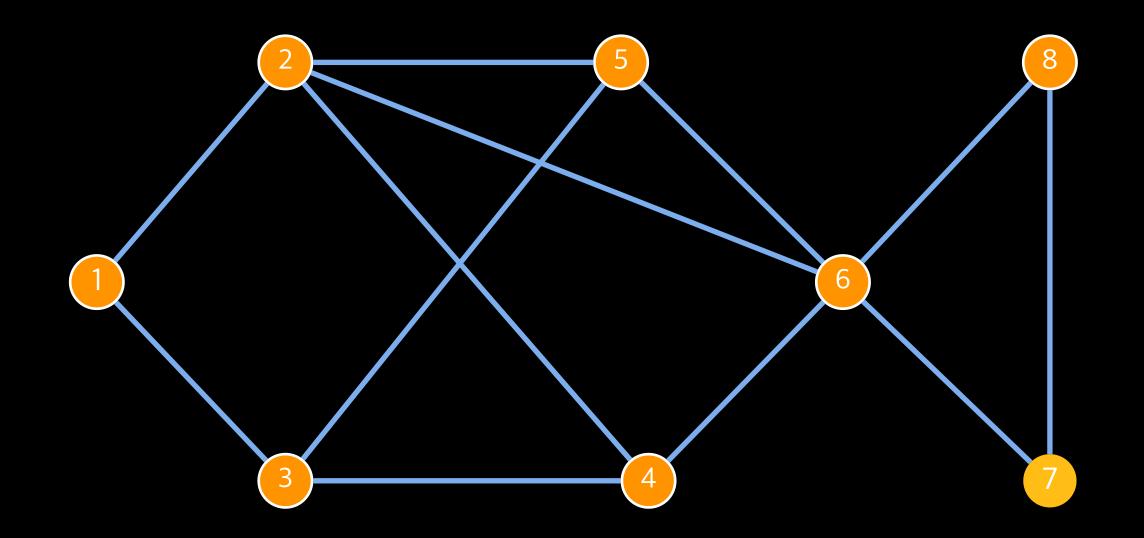


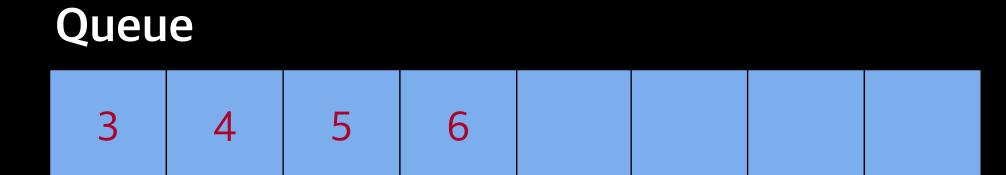
1	2	3	4	5	6	7	8
0	1	1	2				
_	1	1	2				





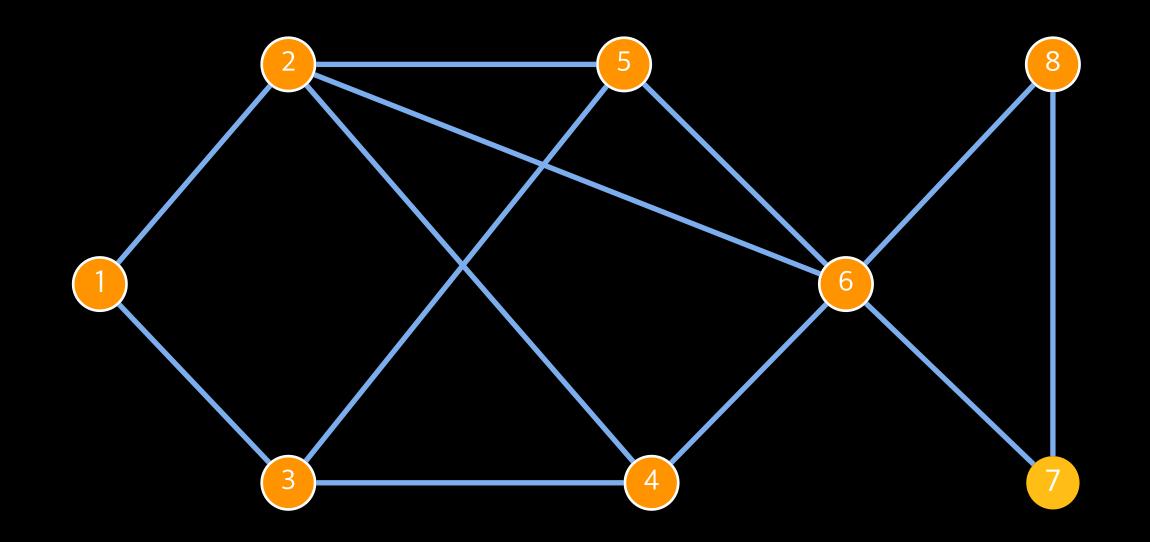
1	2	3	4	5	6	7	8
0	1	1	2	2			
_	1	1	2	2			





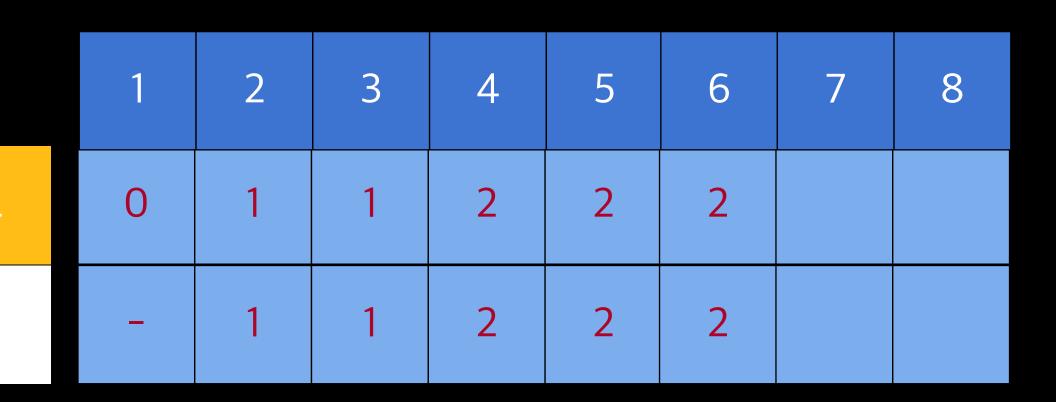
1	2	3	4	5	6	7	8
0	1	1	2	2	2		
_	1	1	2	2	2		

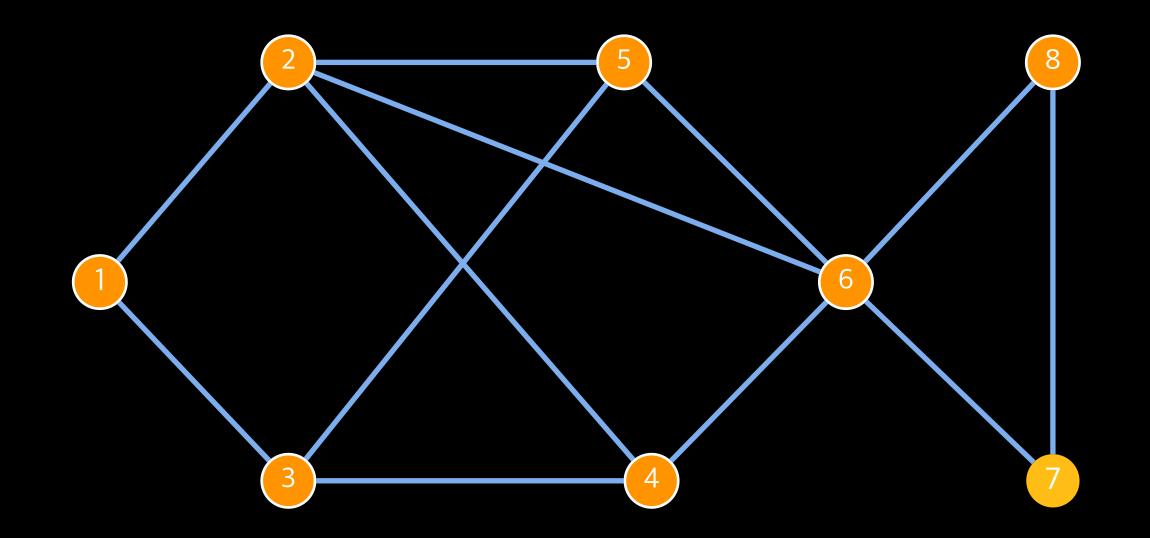
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 append j to queue

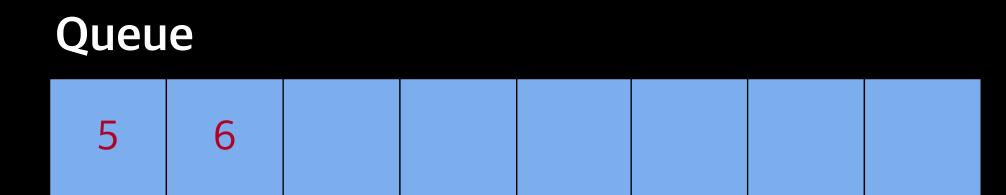


#### Queue

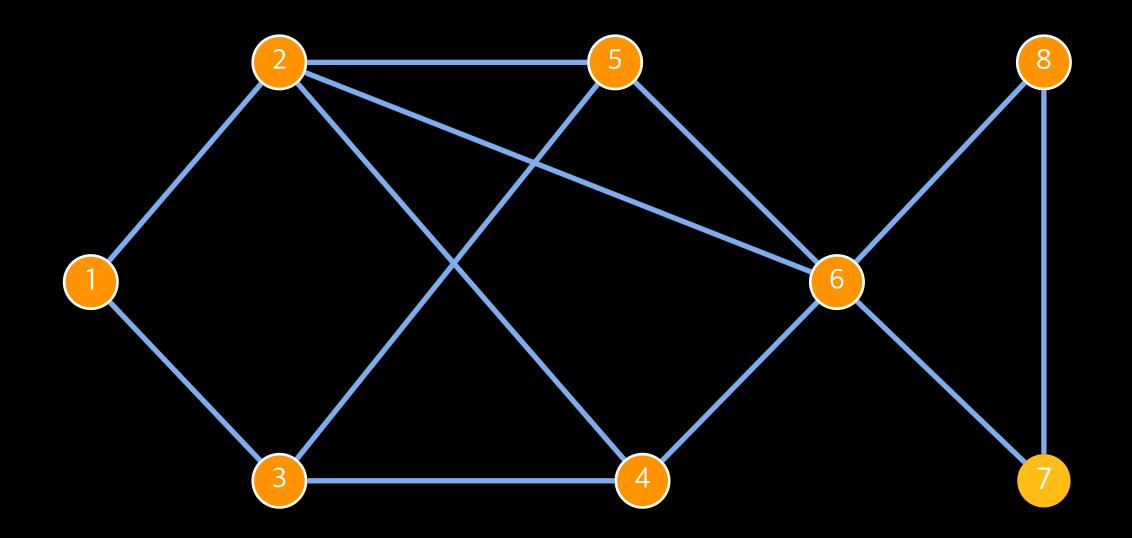
4	5	6			





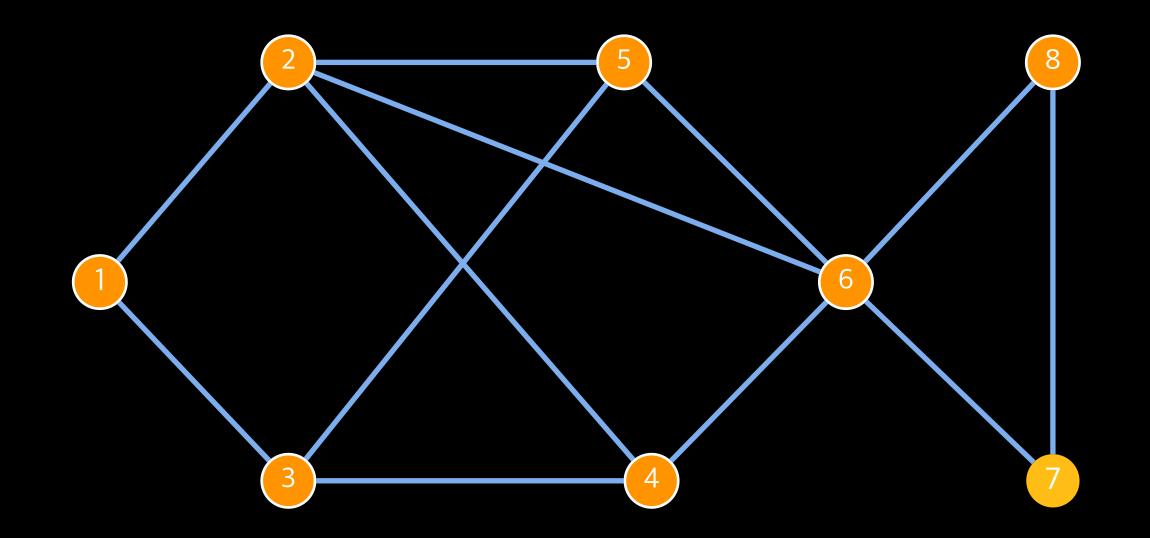


1	2	3	4	5	6	7	8
0	1	1	2	2	2		
_	1	1	2	2	2		



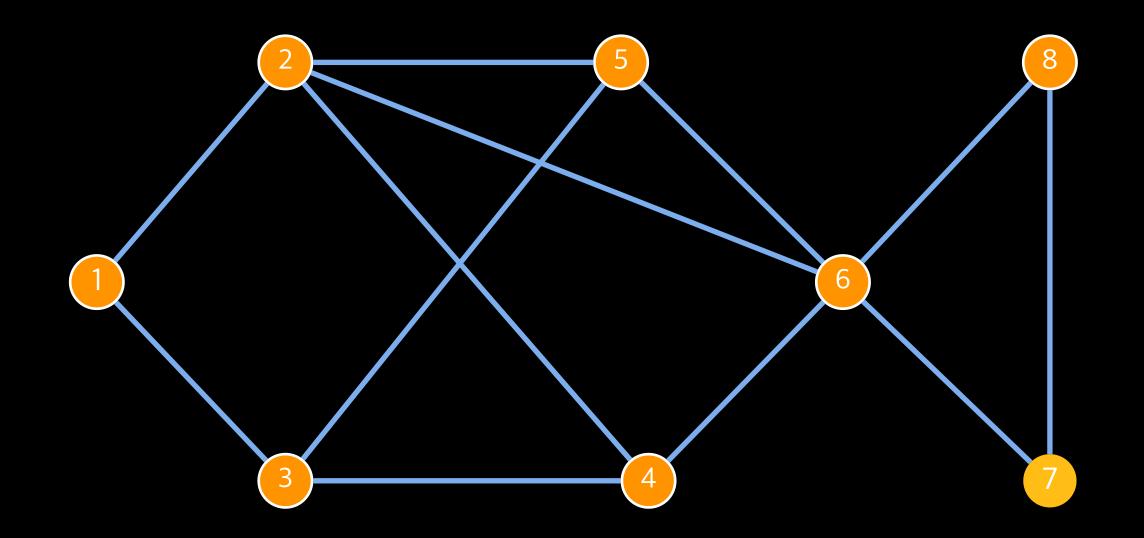


1	2	3	4	5	6	7	8
0	1	1	2	2	2		
_	1	1	2	2	2		



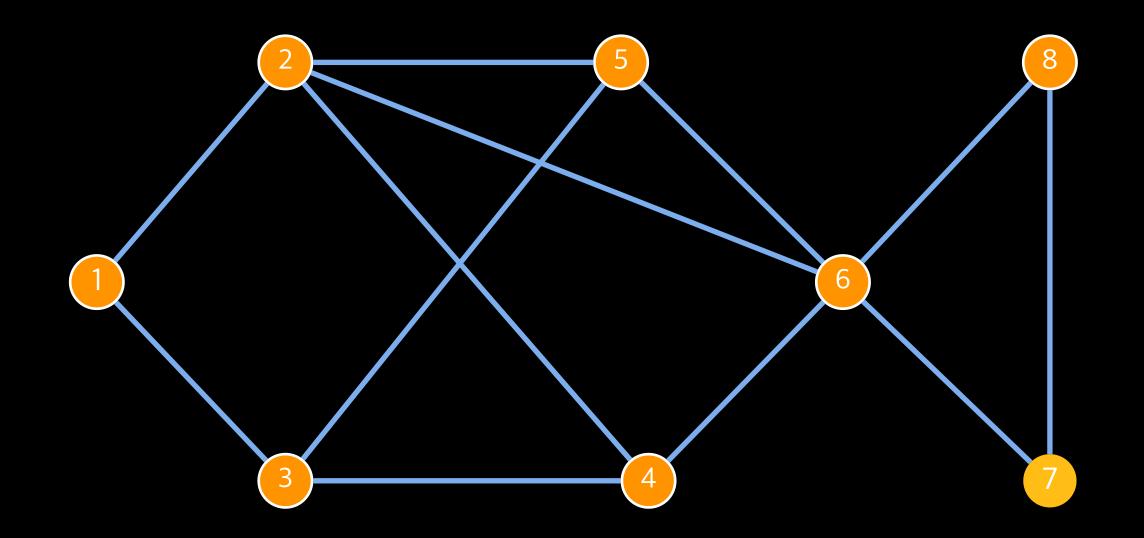


1	2	3	4	5	6	7	8
0	1	1	2	2	2	3	
_	1	1	2	2	2	6	



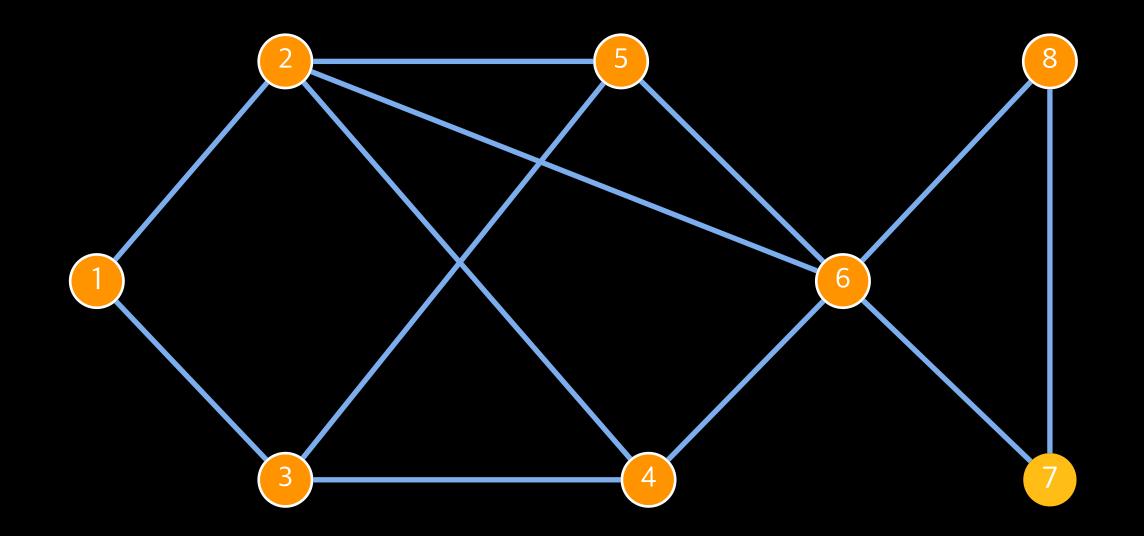


1	2	3	4	5	6	7	8
0	1	1	2	2	2	3	3
_	1	1	2	2	2	6	6





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





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

## SHORTEST PATH

- ◆ BFS with *level*[] gives the shortest path to each node in terms of number of edges
- lacktriangle The modified BFS computes shortest paths, iff  $e_{jw}=1, \forall e_{jw}\in E$
- ♦ If the edges are weighted money, time, distance and  $\neq 1$ , then the level[] may not give the optimal shortest path in terms of the # of edges
  - OR, the path walked may not be the shortest in spite of small number of edges