香港電腦奧林匹克競賽
Hong Kong Olympiad in Informatics

Graph (I)

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2026-03-14

Acknowledgement

➢ A very large portion of this slide is based on this slide by Christy Cheng {christycty}.

Pre-Requisites

Data Structure (I):

- Queue
- Stack
- Linked lists

Recursion, divide and conquer:

- Recursion

Contents

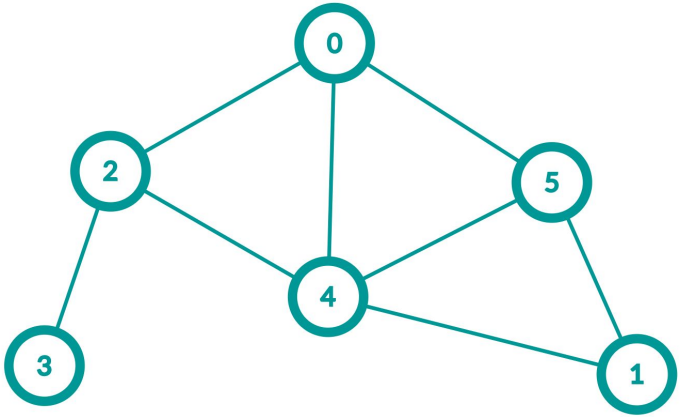
Introduction to Graph

What is a Graph?

A set of vertices/nodes linked by edges

We call a graph (G) = (V,E) where
V = set of vertices ➔ |V| = number of vertices
E = set of edges ➔ |E| = number of edges

Path = sequence of edges which joins a sequence of vertices (usually distinct)

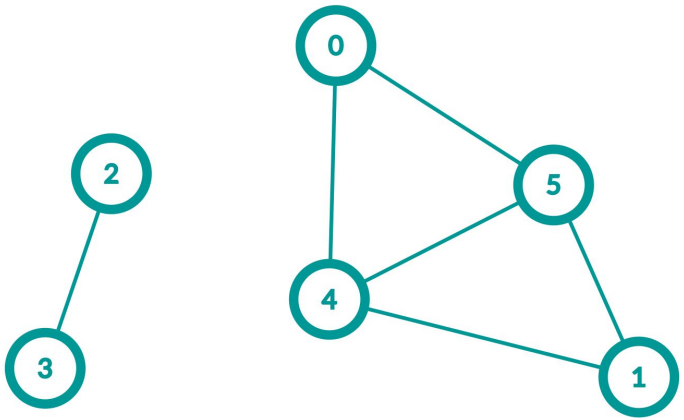


graph with 6 vertices and 8 edges

valid path: 4 → 5 → 0 → 2
invalid path: 1 → 2 → 5

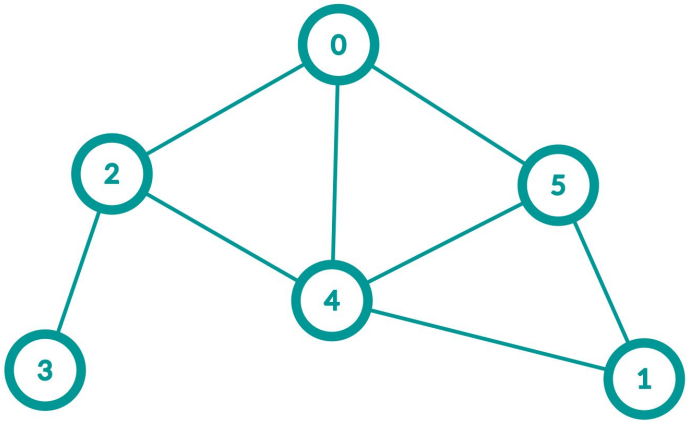
Types of Graphs

Disconnected Graph



{2, 3} and {0, 1, 4, 5} are connected components

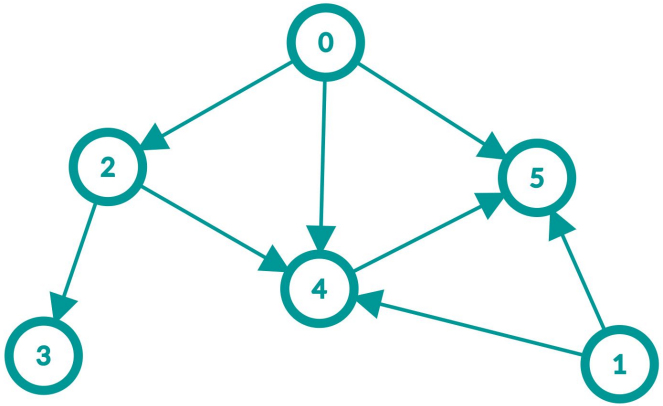
Connected Graph



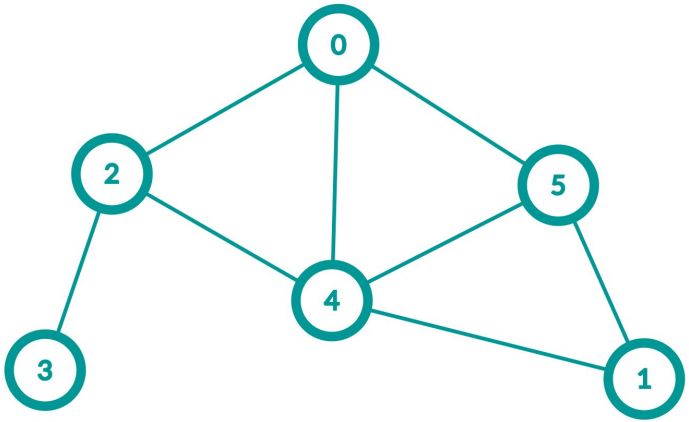
undirected edge = two oppositely directed edges

Types of Graphs

Directed Graph



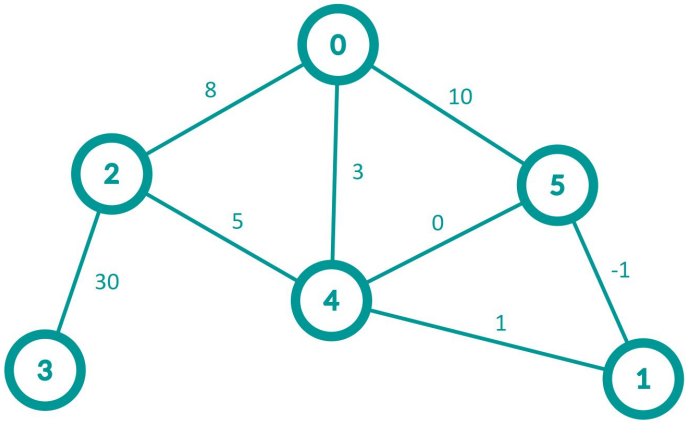
Undirected Graph



undirected edge = two oppositely directed edges

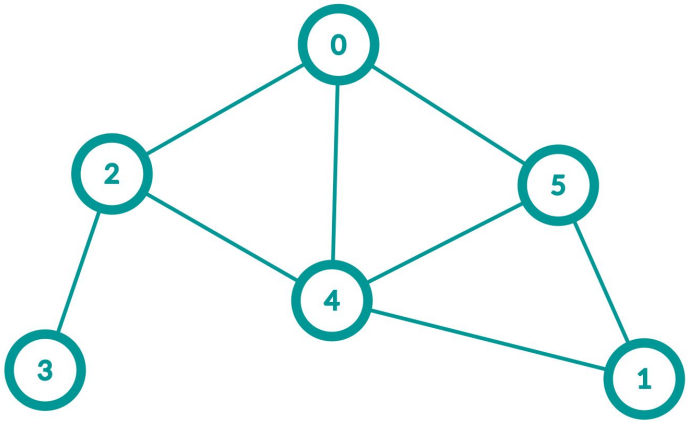
Types of Graphs

Weighted Graph



a value / cost is assigned to each edge

Unweighted Graph



consider as all edges having same weight

Modelling Problems as Graphs

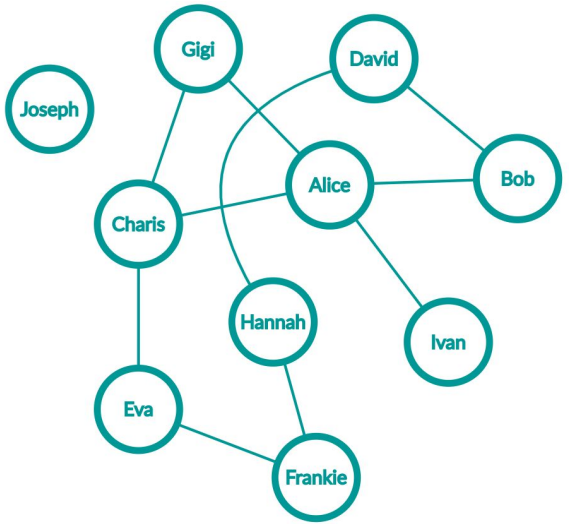
A graph can show relations/transitions between nodes (objects) using edges.

Examples:

- Map
- Maze
- Social networks
- Game states
- Conflict graph

A social network

- nodes = people
- edges = friendship



We can ask:
Do Bob and Charis have any common friends?

Modelling Problems as Graphs

Represent :

- States/ Objects as a vertices
- Possible transitions/ relations as edges
- Transitional costs as weights

It forms a graph and makes things easier to handle.

Modelling Problems as Graphs

Common use cases:

- Find possible paths
- Find shortest path
- Find longest path
- Count number of connected components
- Find nodes reachable with cost less than k
- Find cycles
- ...

States - Water Jug Problem

There are 2 water jugs with capacities N and M litres respectively.

Initially, both of them are empty.

You can perform the following operations for infinitely many times (one operation a time)

1. Empty a jug
2. Fully fill a jug
3. Pour water from one jug to another until either one jug is empty / full

How to get a specific volume K in one of the jugs?

States

The states in this problem would be the volumes of the two jugs (*v1*, *v2*)

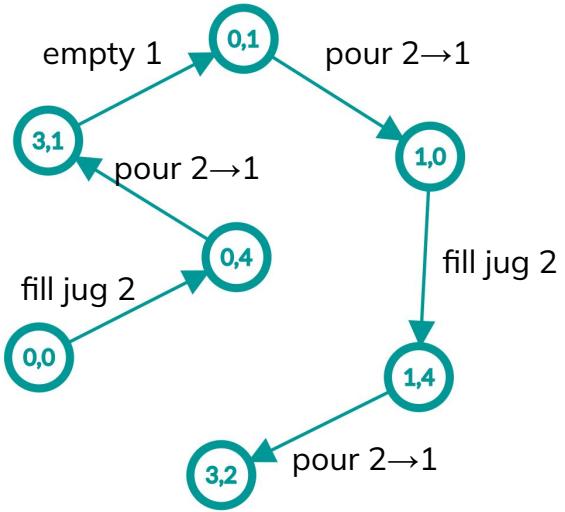
How to we represent them as a graph?

N = 3 *(jug 1)*, M = 4 *(jug 2)*, K = 2 *(goal)*

Initial state = (0, 0); Final state = (3, 2)

We can represent the solution as a path in the graph

We will further discuss this problem later.



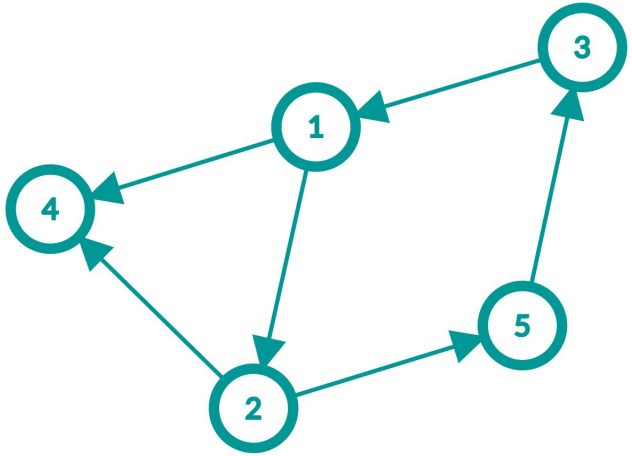
Graph Representation

How can we store a graph?

1. Adjacency Matrix
2. Edge List
3. Adjacency List

Methods needed:

1. Add an edge between two nodes
2. Get a node's neighbors (or parent/child)



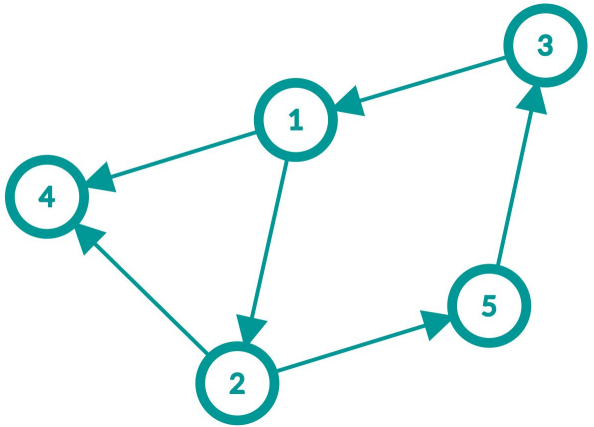
example: can we go from node 2 to node 1?

Adjacency Matrix

A |V|x|V| array representing the relation / weight of every pair of vertices

`adj[i][j] = 1 (or weight)` if there is an edge from vertex *i* to vertex *j*

else `adj[i][j] = 0`



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

Adjacency Matrix - Implementation

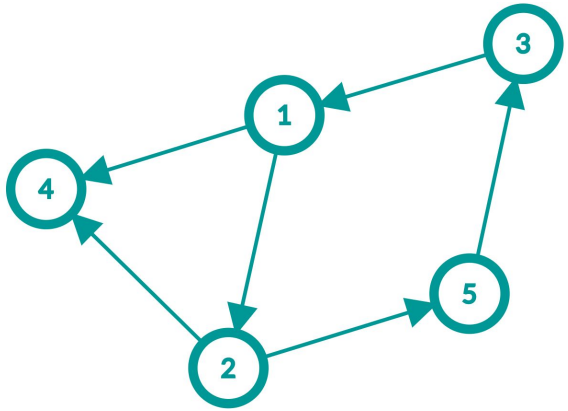
```
adj[V][V]

for i ← 1 to V
for j ← 1 to V
     adj[i][j] = 0

for all edges from u to v:
     adj[u][v] = 1

for all query if u → v exist:
     output adj[u][v]
```

Memory: $O(|V|^2)$

$O(|V|^2)$

$O(|E|)$

O(1) per query



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

Adjacency Matrix - Implementation

How to output all children of node x?

all nodes y that exist an edge x → y

```
for i ← 1 to V
    if adj[x][i] == 1
        output i
```

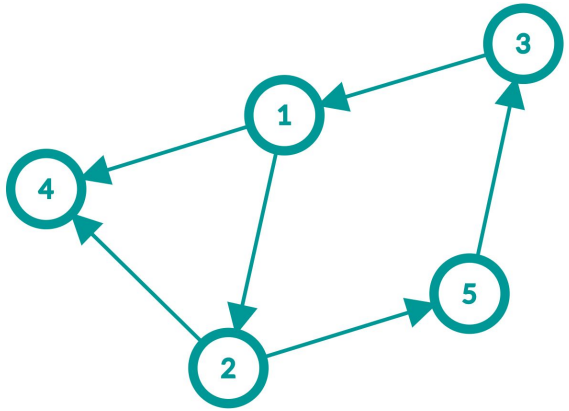
O(|V|) per query

How to output all parents of node x?

all nodes y that exist an edge y → x

```
for i ← 1 to V
    if adj[i][x] == 1
        output i
```

O(|V|) per query

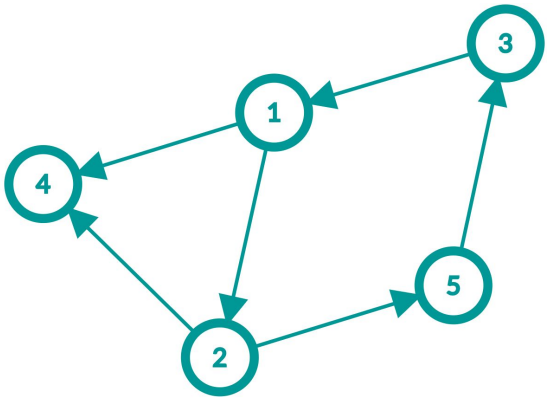


adj	1	2	3	4	5
1	0	1	0	1	0
2	0	0	0	1	1
3	1	0	0	0	0
4	0	0	0	0	0
5	0	0	1	0	0

Edge List

Array of size `|E|` to store each edge:

`from[i]`, `to[i]` and `weight[i]` represent the the origin, the destination and the path weight of the i-th edge respectively.



ID	from	to	weight
1	1	2	1
2	1	4	1
3	2	4	1
4	2	5	1
5	3	1	1
6	5	3	1

Edge List - Implementation

```
edge[E]
edge_no = 0
```

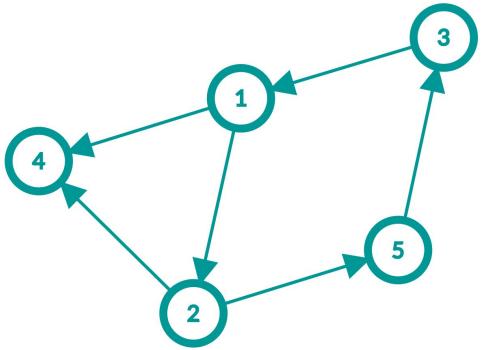
Memory: O(|E|)

```
for all edges from u to v:
    edge[edge_no] = edge(u,v)
    edge_no++
```

O(|E|)

```
for all query if u → v exist:
    for i ← 0 to edge_no - 1:
        if edge[i] is edge(u,v):
            output YES
```

O(|E|) per query



ID	from	to
1	1	2
2	1	4
3	2	4
4	2	5
5	3	1
6	5	3

Edge List - Implementation

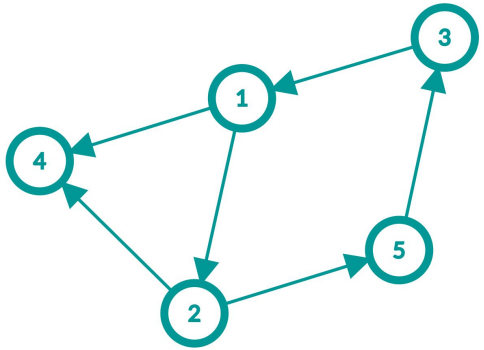
Implement with C++ struct and vector

```
struct Edge {
    int from, to, weight;
};

vector<Edge> edge_list;
```

(easier memory allocation)

Insert edge u → v

```
Edge edge;
edge.from = u, edge.to = v, edge.weight = 1;
edge_list.push_back(edge);
```



ID	from	to
1	1	2
2	1	4
3	2	4
4	2	5
5	3	1
6	5	3

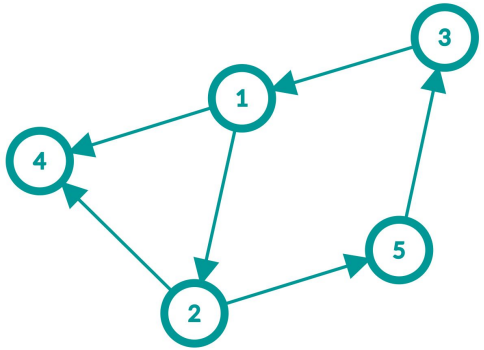
Adjacency List

Idea: for each node, maintain a list of nodes/edges connected to it

in practice, it is usually neighbours in undirected edges, or child for directed edges

The list may be:

- Arrays of size |E|
- Linked lists
- Vectors



1	2	4
2	4	5
3	1	
4		
5	3	

Adjacency List - Implementation

```
neighbour[V]

for all edges from u to v:
     neighbour[u].push_back(v)
     neighbour[v].push_back(u)
```

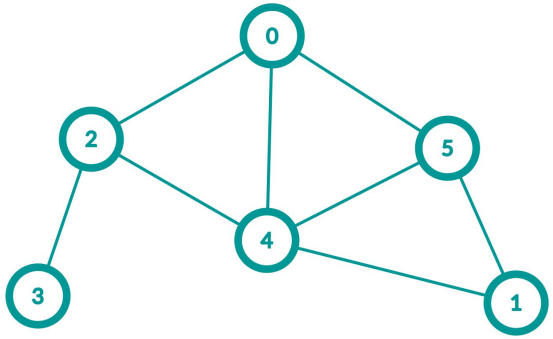
O(|E|)

```
all neighbours of node u
     for x in neighbour[u]:
          output x
```

O(|E|)

```
is v a neighbour of node u
     for x in neighbour[u]:
          if x == v:
               output YES
```

O(|E|) per query



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

fill the matrix

Adjacency List - Implementation

Implement with C++ vector

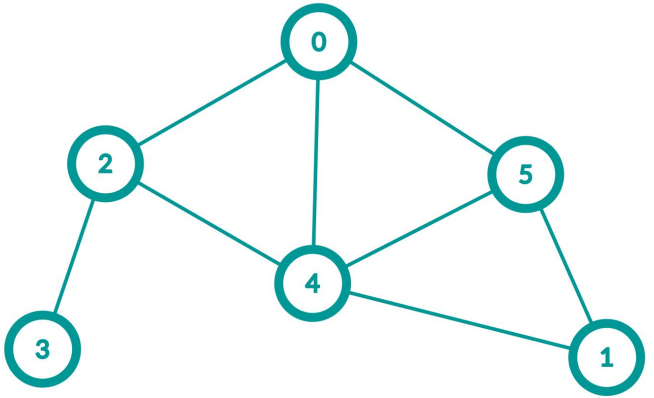
```
vector<int> neighbour[V];
```

Insert edge u - v

```
neighbour[u].push_back(v);
neighbour[v].push_back(u);
```

All neighbours of node u

```
for auto i in neighbour[u]:
    cout << i << " ";
```



Adjacency List - Implementation

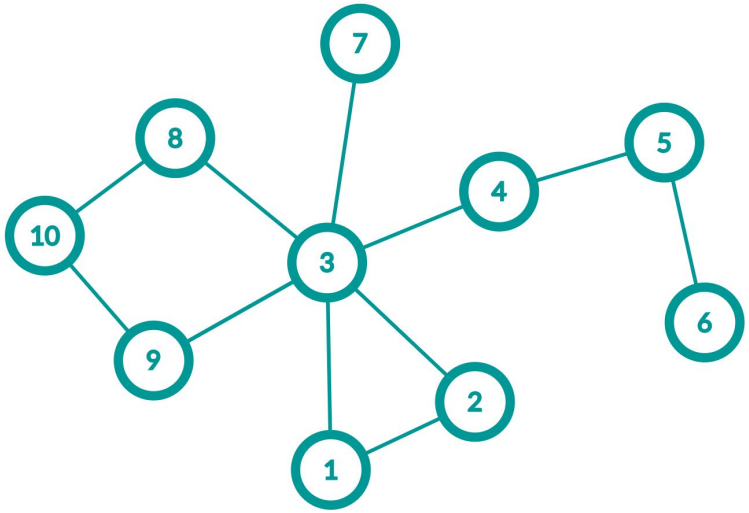
	Array	Linked List	Vector
Declaration	int neighbour[V][E]	Node* neighbour[V]	vector<int> neighbour[V]
Space Efficiency	O(|V||E|)	O(|V|+|E|)	O(|E|)

for more implementation detail, refer to data structure (I)

in practice, just use vector

Practice

Adjacency list for this graph?



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

Pros and Cons

	Adjacency Matrix	Edge List	Adjacency List
Ease of Implement	High	Middle	Low
Space Efficiency	$O(\|V\|^2)$	$O(\|E\|)$	$O(\|V\|+\|E\|)$
Retrieve 1 edge	$O(1)$	$O(\|E\|)$	*faster than* $O(\|E\|)$
Retrieve all neighbour	$O(\|V\|)$	$O(\|E\|)$	$O(\|V\|+\|E\|)$ for all nodes
Multiple Edges	No	Yes	Yes

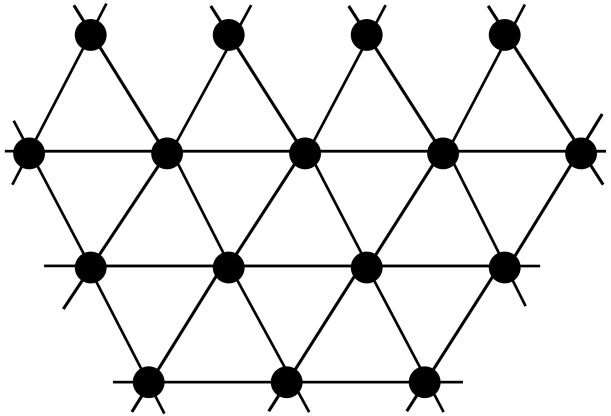
Which to choose?

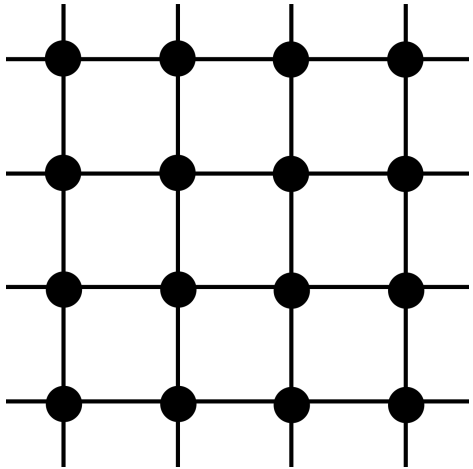
In practice, we almost never use edge list

Adjacency Matrix	Adjacency List
|V| is small (e.g. <= 500)	|V| is large (most cases)
dense edges	sparse edges
query random and unrelated edges	query all edges connected to a node
Use Cases	
Floyd-warshall algorithm	Dijkstra's *(and other BFS-based)* algorithm
	Searching (DFS, BFS)

Grid graphs

In some problems, vertices form a regular tiling (square, triangle...)





Grid graphs

vertices represented by coordinates (row, column)

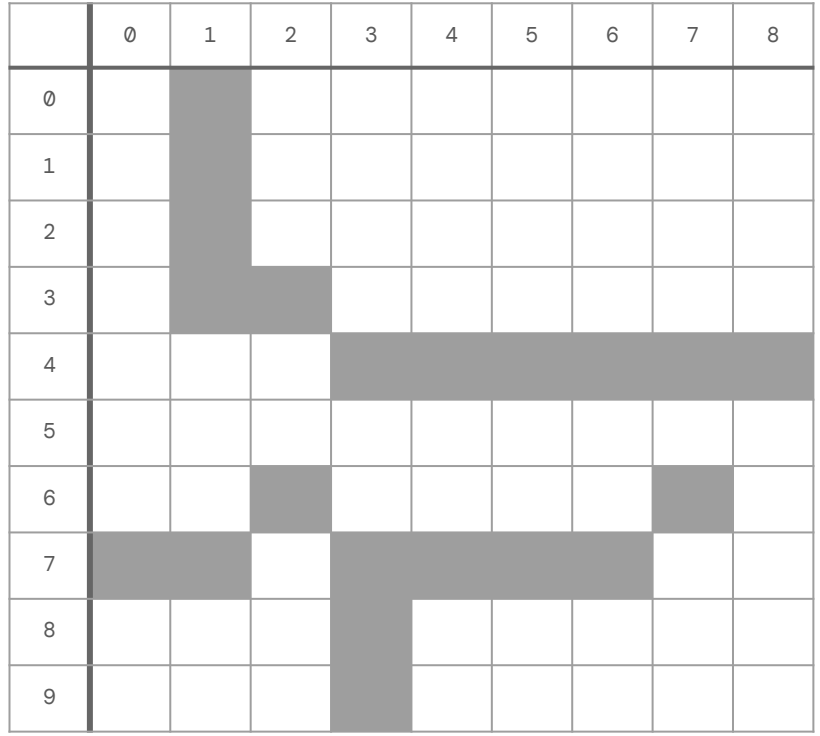
edges are often implicitly given

e.g. in a maze you could only go up/down/left/right

→ edge between each grid and all grids it can visit

in practice, usually touching edges (and corners)

assume that every node has these edges
→ check if these edges are valid as we traverse the graph (e.g. the grid is blocked?)



▨ blocked

Grid graphs - Implementation

1. get neighbors
2. scan all the directions
3. check if the vertex is out-of-bound or invalid

```
N = 3 M = 4
..#.
.#..
....
Neighbors of (1, 2) :(2, 2) (1, 3)
Neighbors of (0, 3) :(1, 3)
```

```
const int MAX_N = 10;
char grid[MAX_N][MAX_N];
int dx[] = {-1, 1, 0, 0};
int dy[] = {0, 0, 1, -1};

void print_neighbors(int r, int c) {
    cout << "Neighbors of (" << r <<  ", " << c << "):";
    for (int i = 0; i < 4; i++){
        int nx = dx[i] + r;
        int ny = dy[i] + c;
        bool out_of_bound = nx < 0 || ny < 0 || nx >= N
             || ny >= M;
        if (out_of_bound) continue;
        bool valid = (grid[nx][ny] == '.');
        if (valid) {
            cout << "(" << nx <<  ", " << ny << ") ";
        }
    }
    cout << "\n";
}
```

Attendance Taking

Practice + 5-minute Break (Until 10:50)

Given a graph of 5 nodes, which has no edges initially.

We want to support the following online queries:

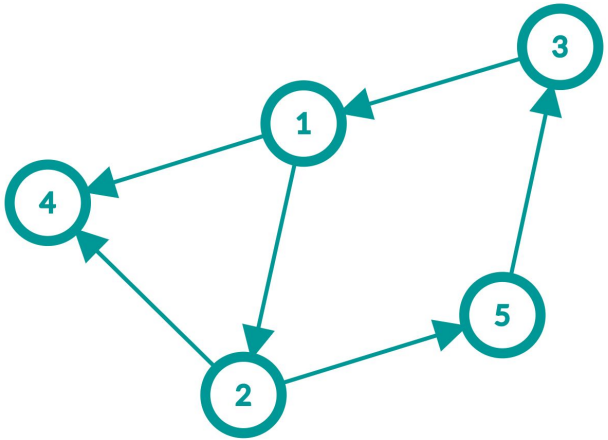
1. Add an edge from node a to node b
2. Remove the edge from node a to node b
3. Output if there is an edge from node a to node b

Each line (representing a query) contains 3 integers: cmd a b

1 4 5 *add an edge from node 4 to node 5*

2 2 5 *remove an edge from node 2 to node 5*

3 3 4 *output if there is an edge from node 3 to node 4*



Try to test with this graph

Graph Traversal

Depth-First Search

Idea:

1. Choose an unvisited path
2. Go as far as possible
3. Go back
4. Explore a new path

Repeat until there is no unvisited path left.

Time complexity

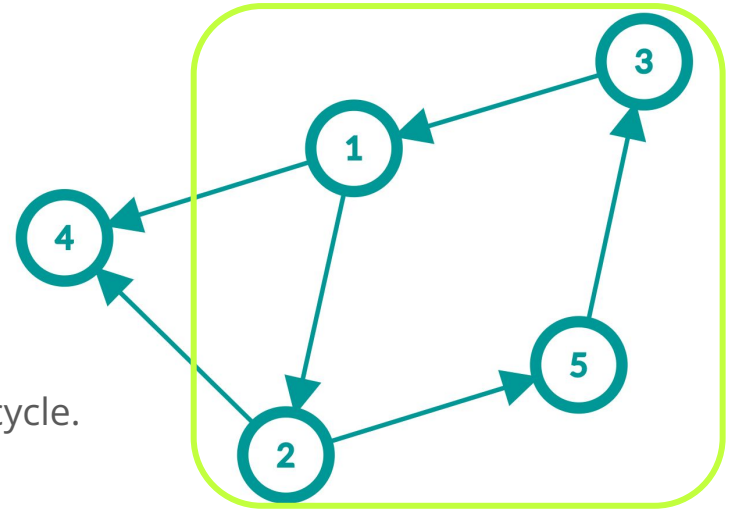
- $O(|V|^2)$ for adjacency matrix
- $O(|V|+|E|)$ for adjacency list

Depth-First Search

There can be cycles - there exists a path that starts from and ends at the same vertex.

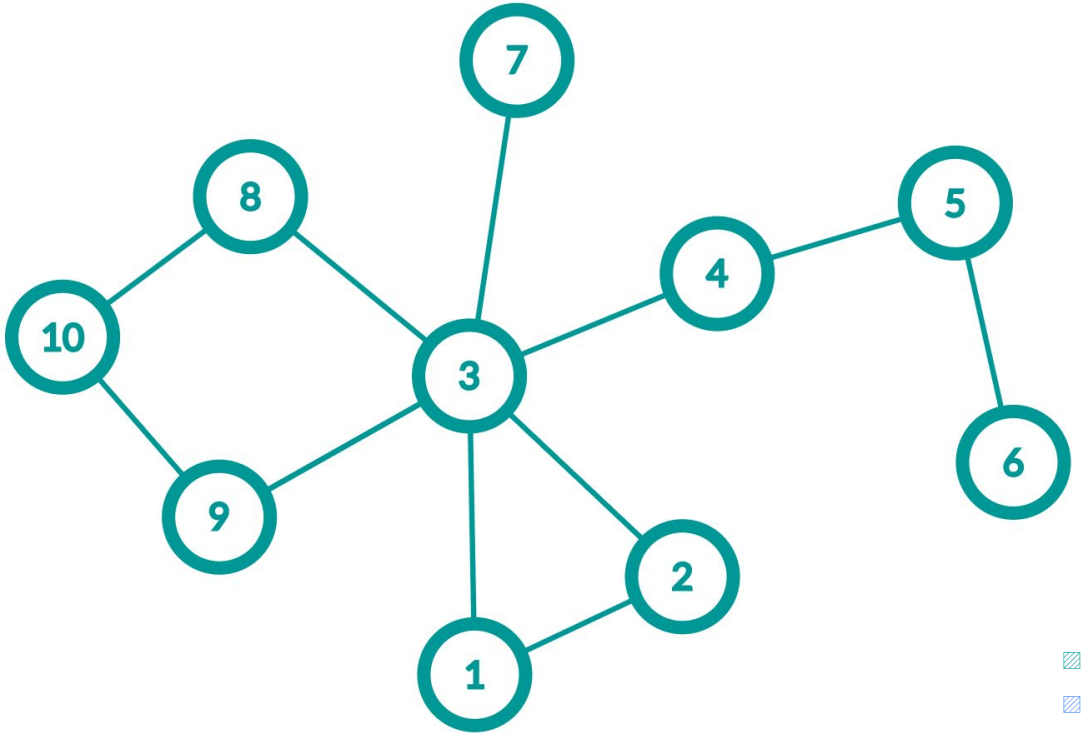
Recall that we do not go on a visited path, so we can stop going forward and treat the current vertex as a ‘dead end’

In this case, not all edges are used
but all vertices are travelled



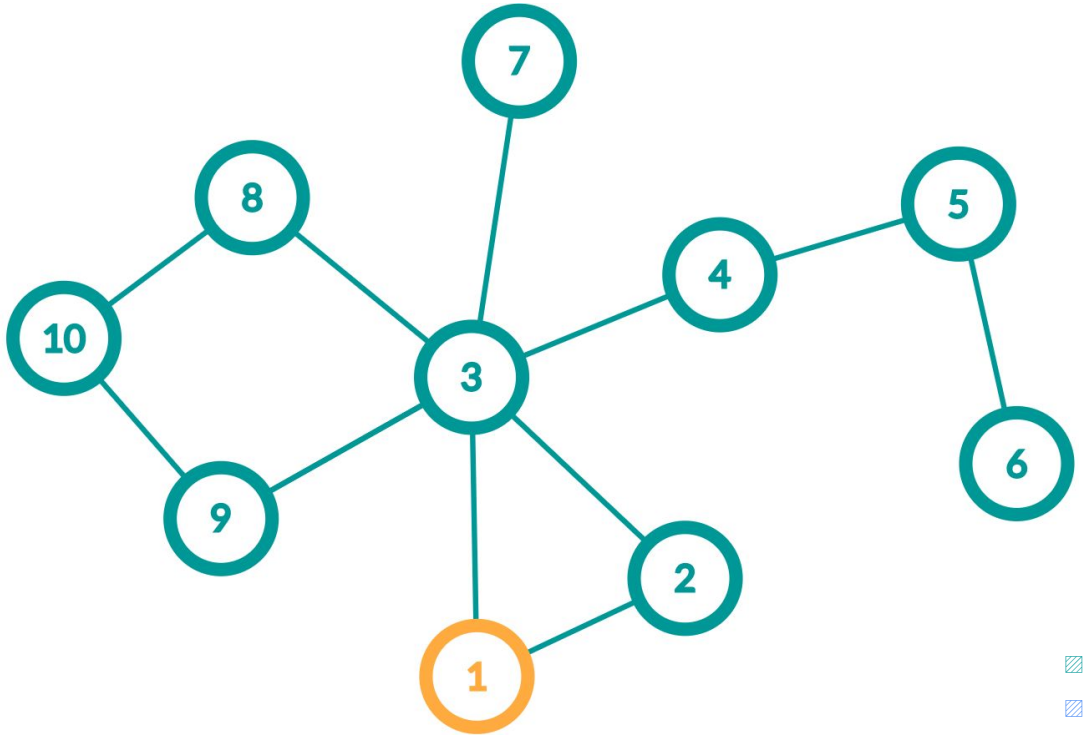
{1, 2, 3, 5} forms a cycle.

Function call



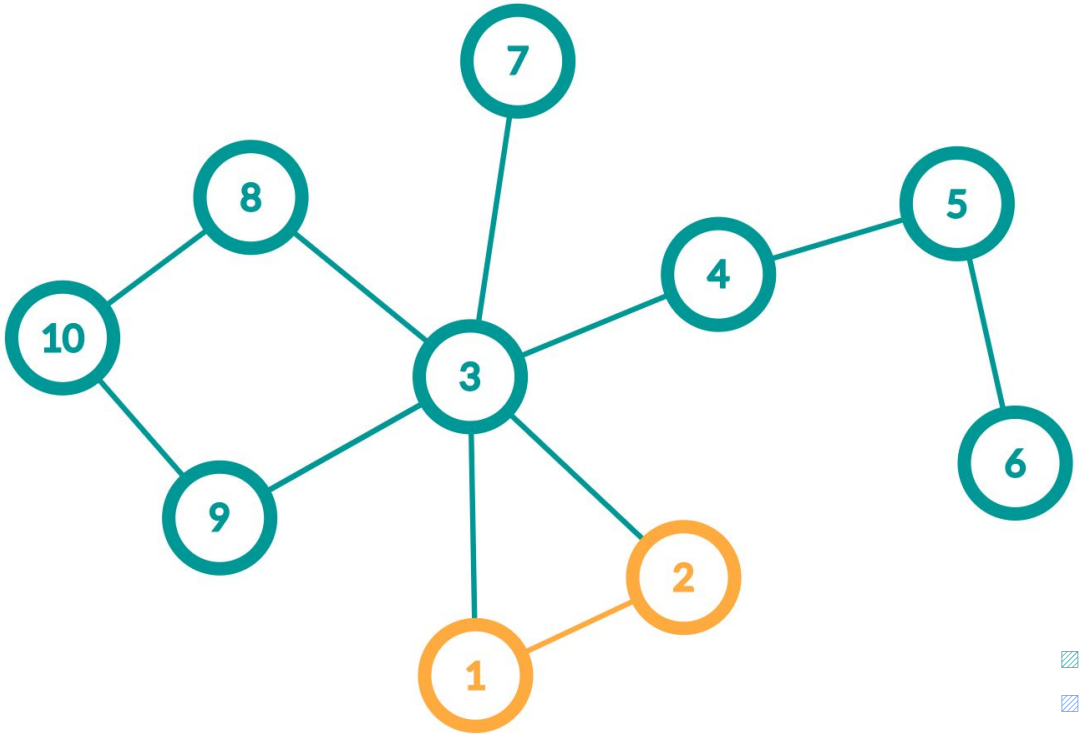
unvisited
visited
dead

DFS(1)
Function call



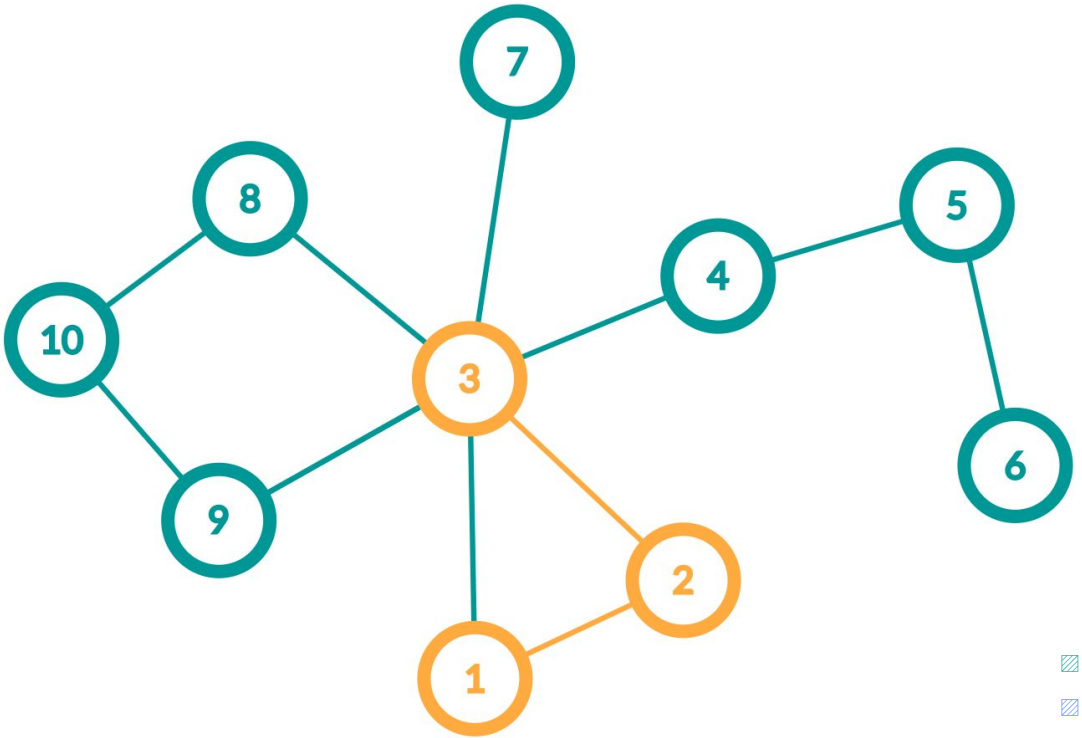
unvisited
visited
dead

DFS(2)
DFS(1)
Function call



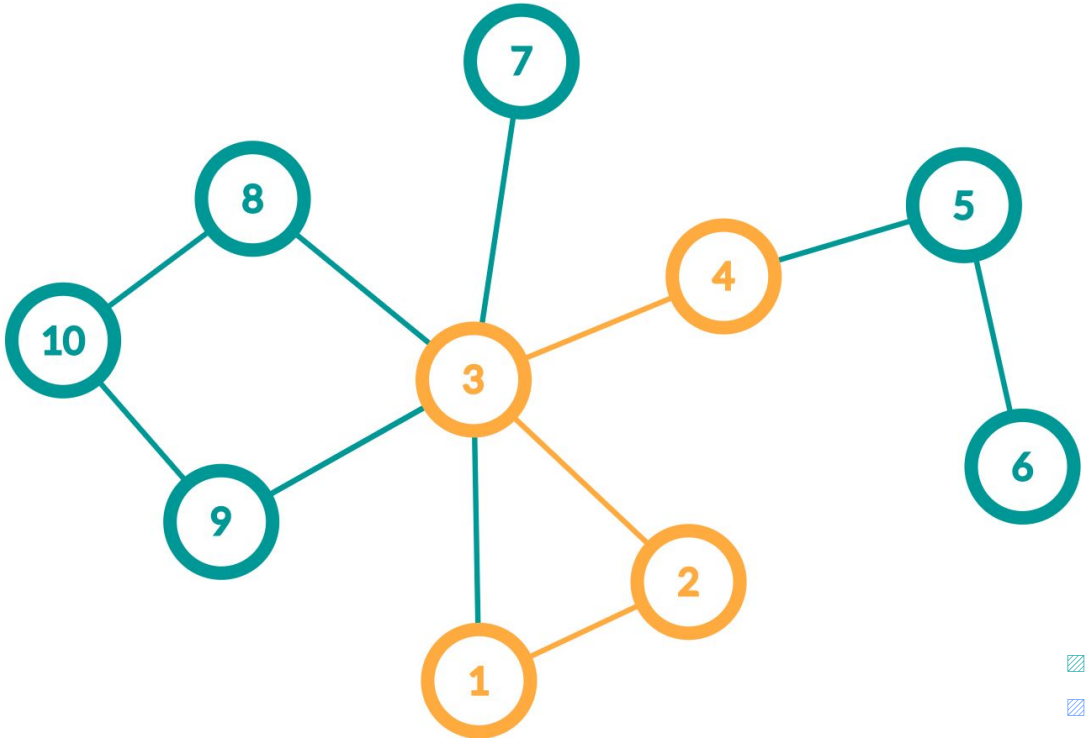
unvisited
visited
dead

DFS(3)
DFS(2)
DFS(1)
Function call

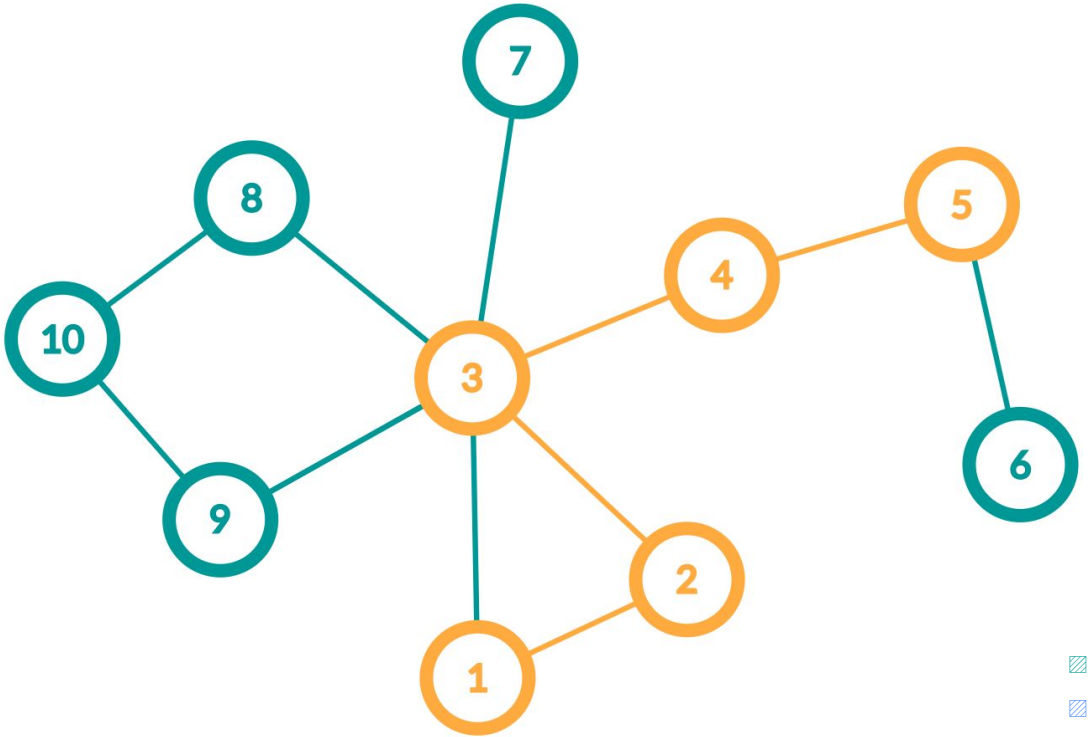


unvisited
visited
dead

DFS(4)
DFS(3)
DFS(2)
DFS(1)
Function call

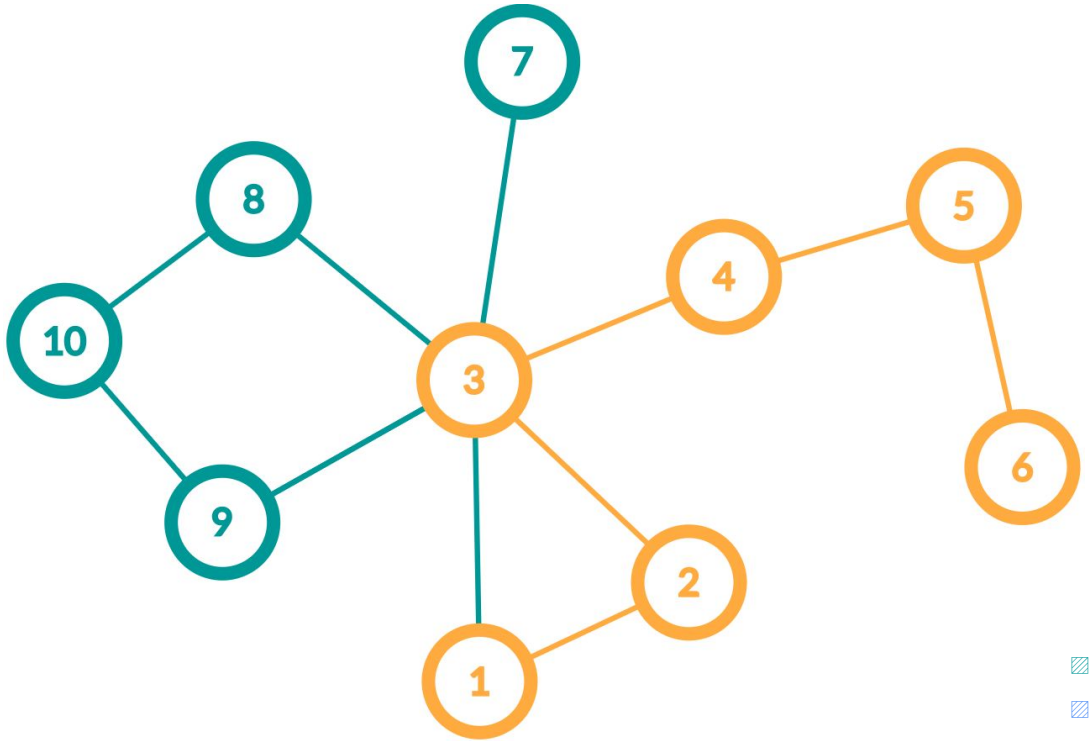


DFS(5)
DFS(4)
DFS(3)
DFS(2)
DFS(1)
Function call

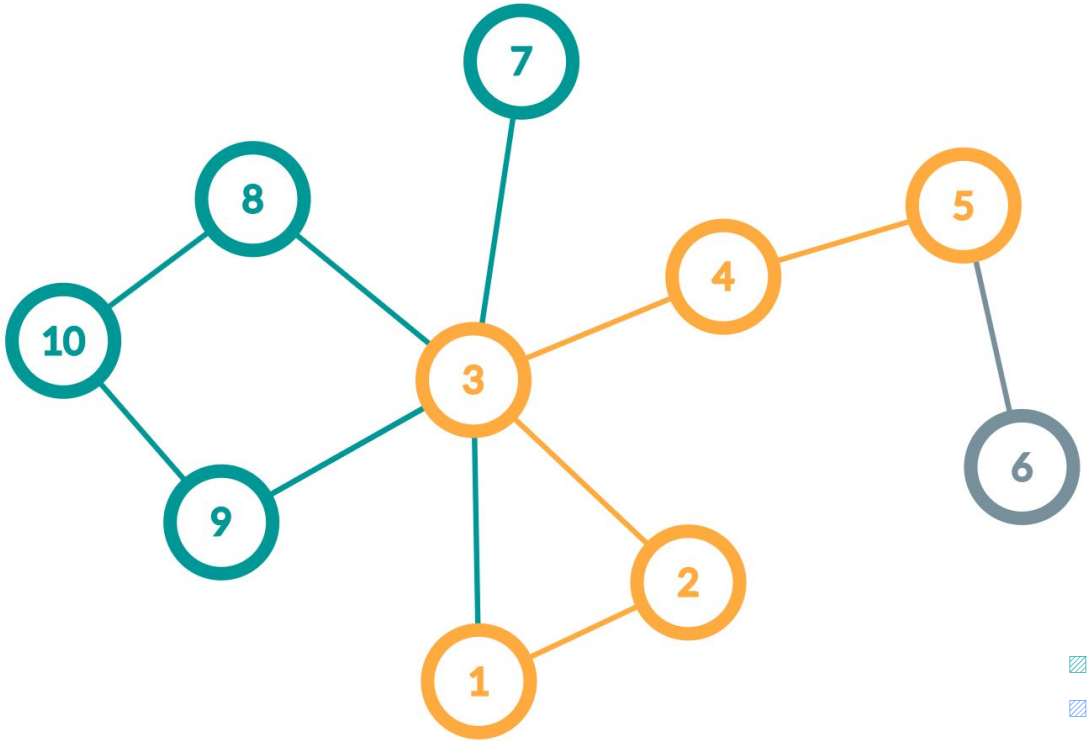


unvisited
visited
dead

DFS(6)
DFS(5)
DFS(4)
DFS(3)
DFS(2)
DFS(1)
Function call

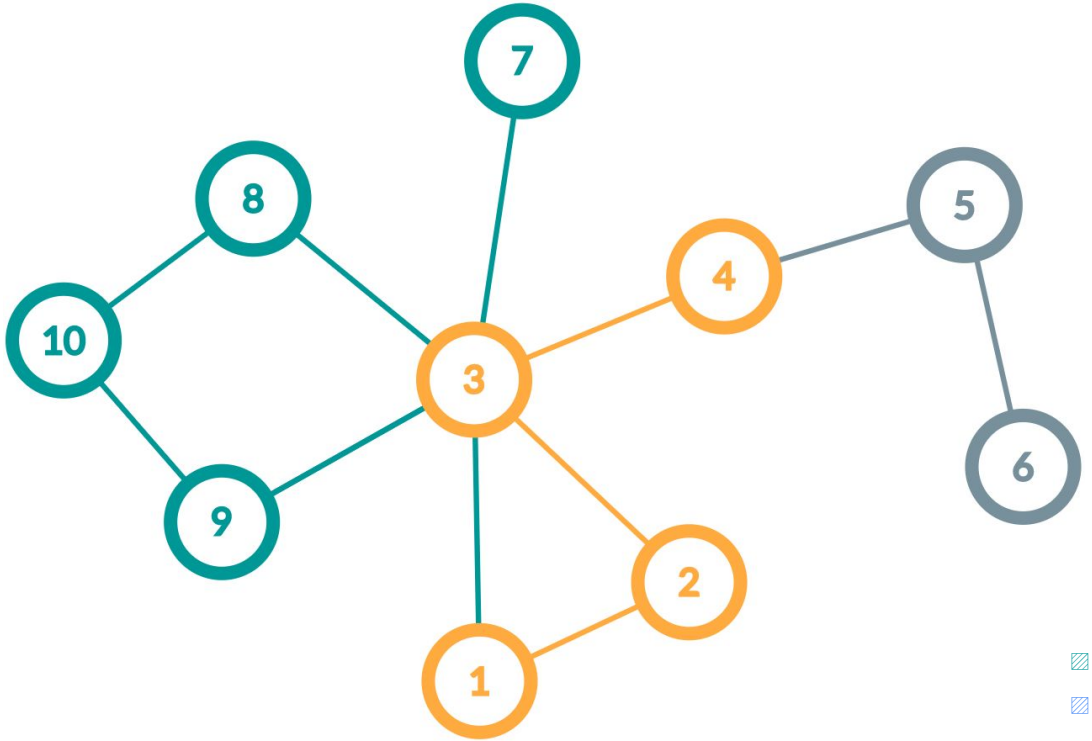


DFS(5)
DFS(4)
DFS(3)
DFS(2)
DFS(1)
Function call



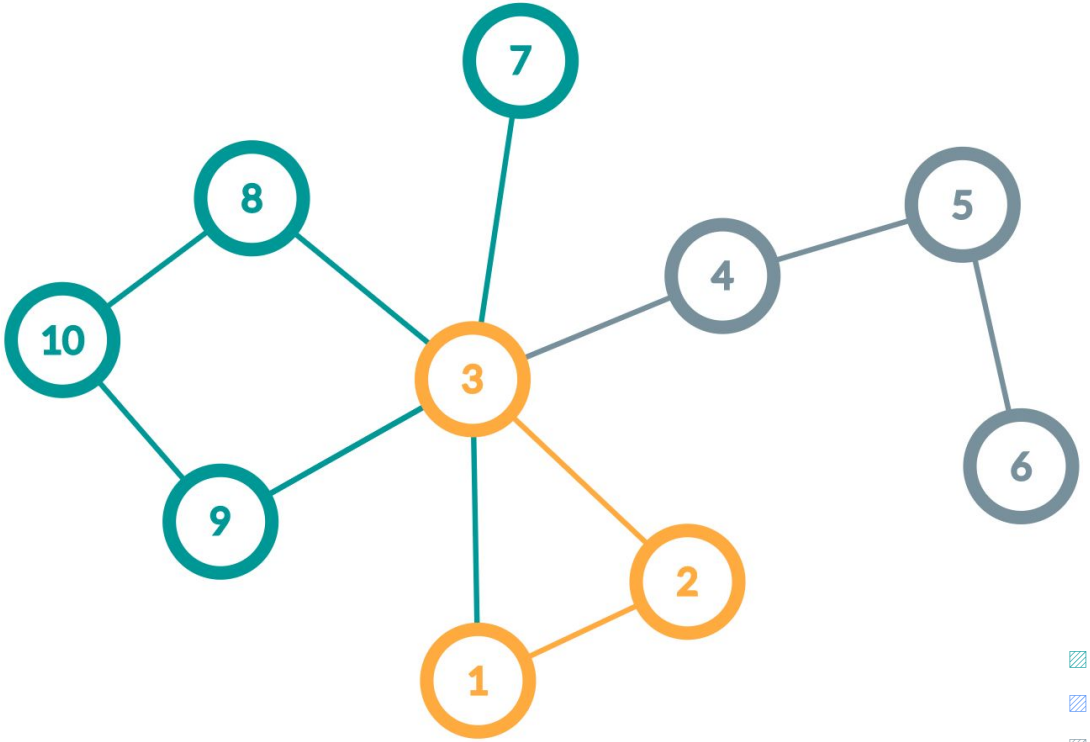
unvisited
visited
dead

DFS(4)
DFS(3)
DFS(2)
DFS(1)
Function call



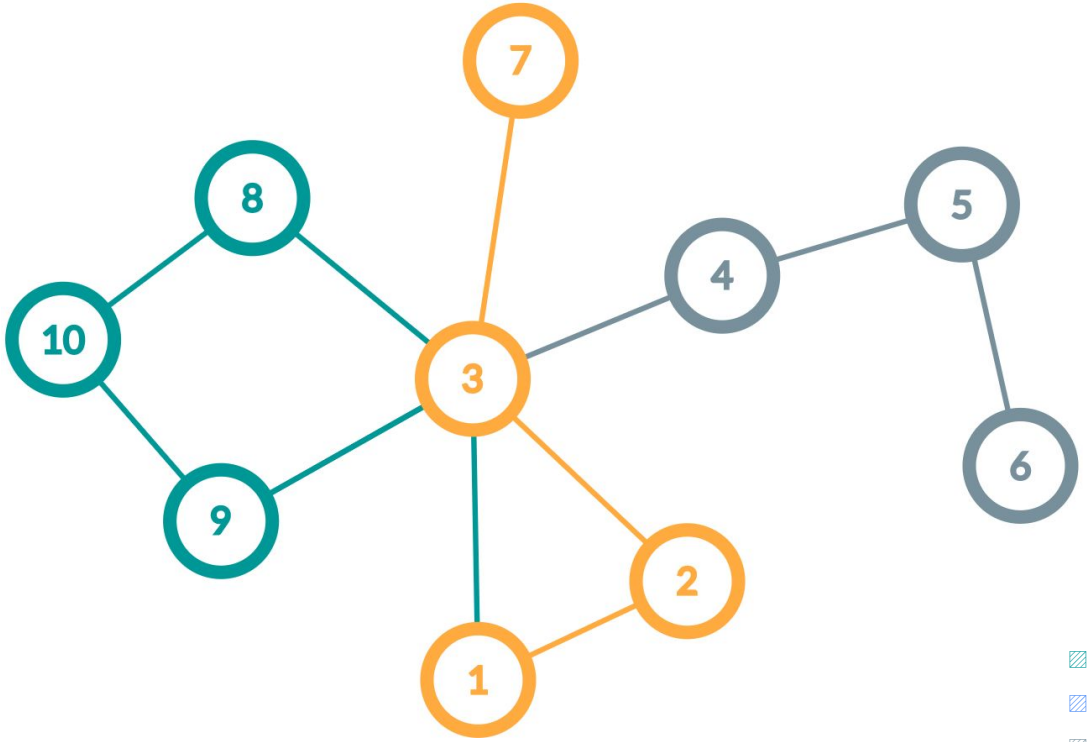
unvisited
visited
dead

DFS(3)
DFS(2)
DFS(1)
Function call



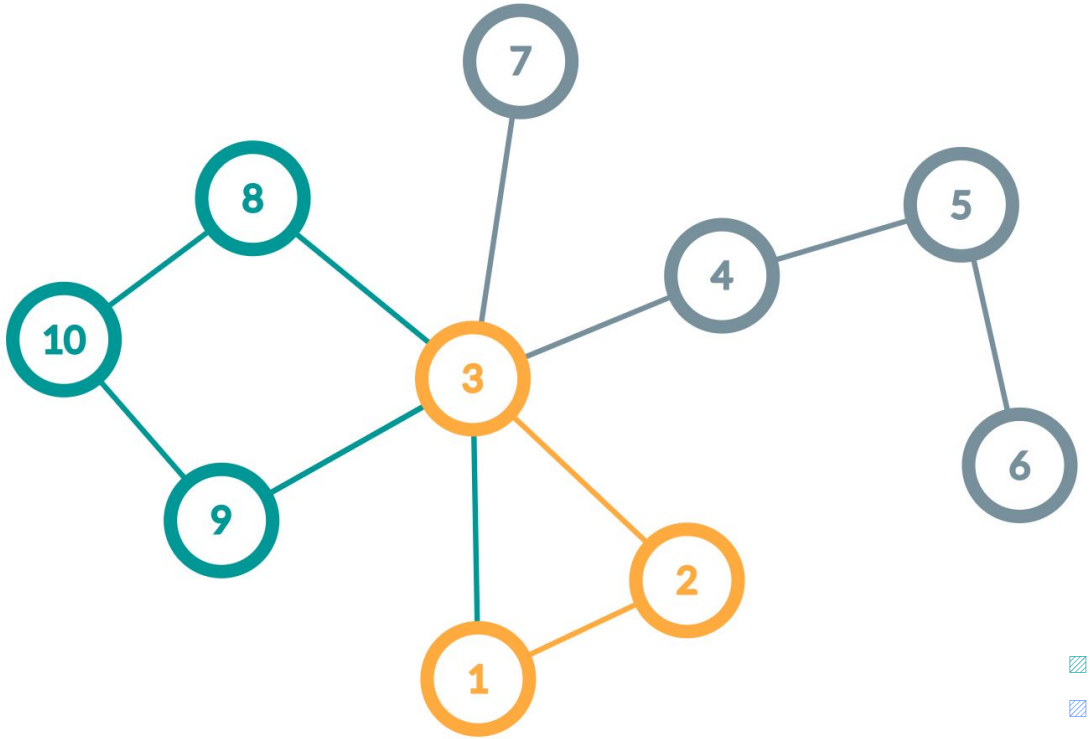
unvisited
visited
dead

DFS(7)
DFS(3)
DFS(2)
DFS(1)
Function call



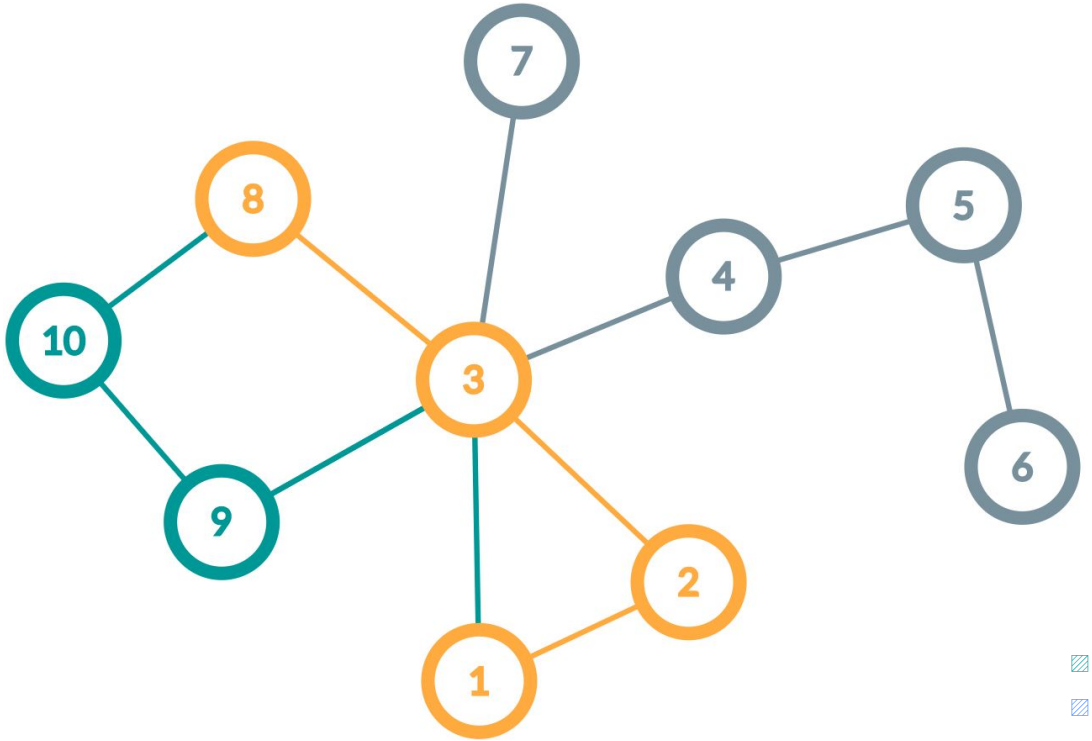
unvisited
visited
dead

DFS(3)
DFS(2)
DFS(1)
Function call



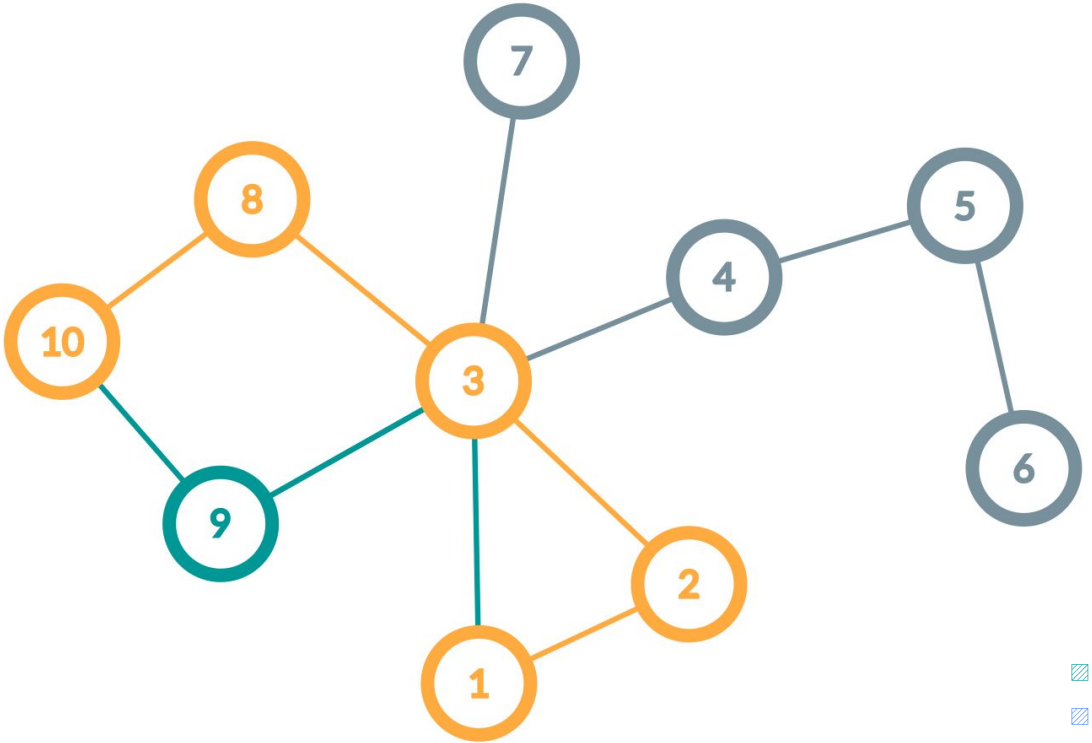
unvisited
visited
dead

DFS(8)
DFS(3)
DFS(2)
DFS(1)
Function call

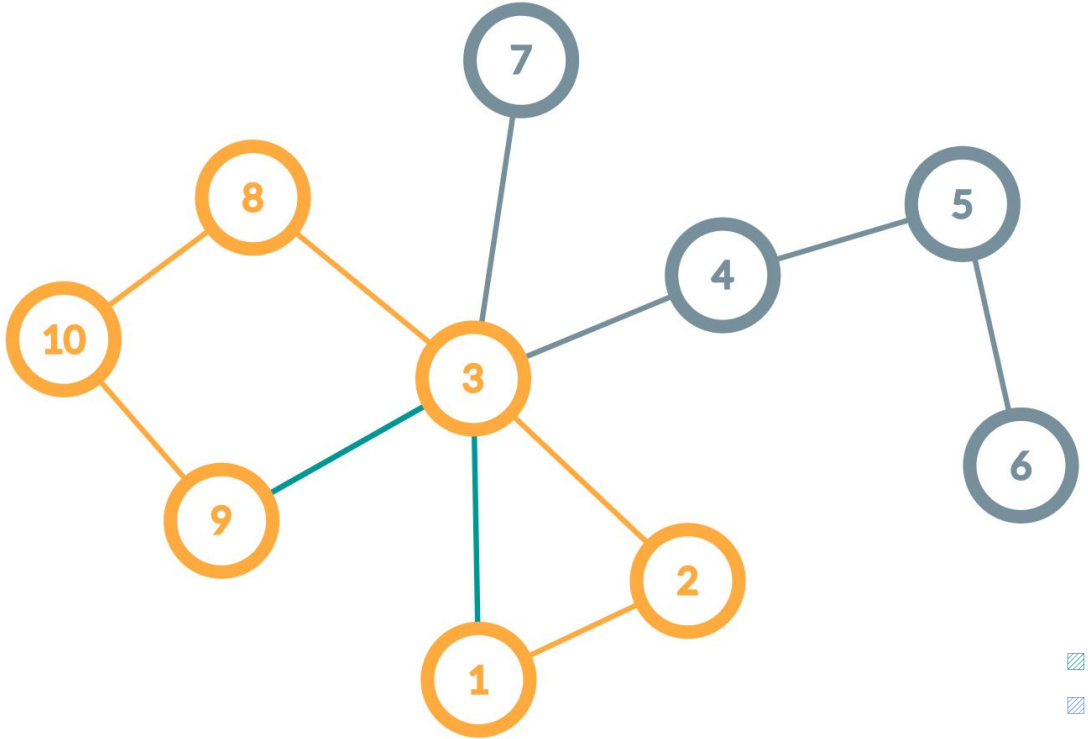


unvisited
visited
dead

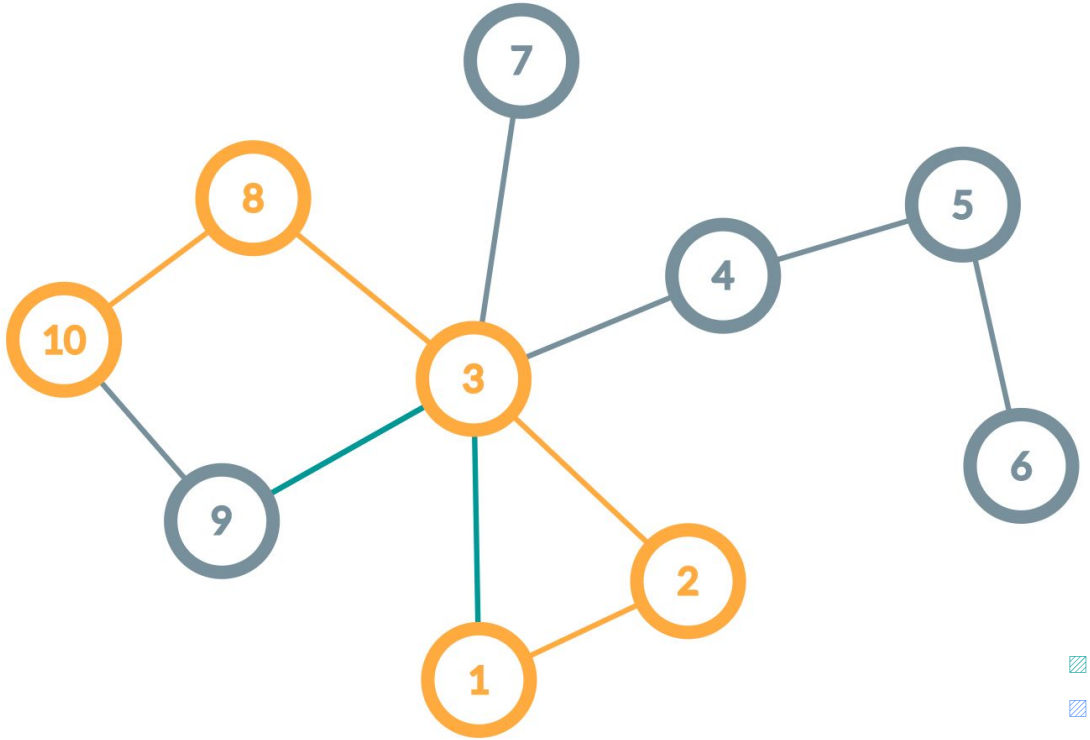
DFS(10)
DFS(8)
DFS(3)
DFS(2)
DFS(1)
Function call



DFS(9)
DFS(10)
DFS(8)
DFS(3)
DFS(2)
DFS(1)
Function call

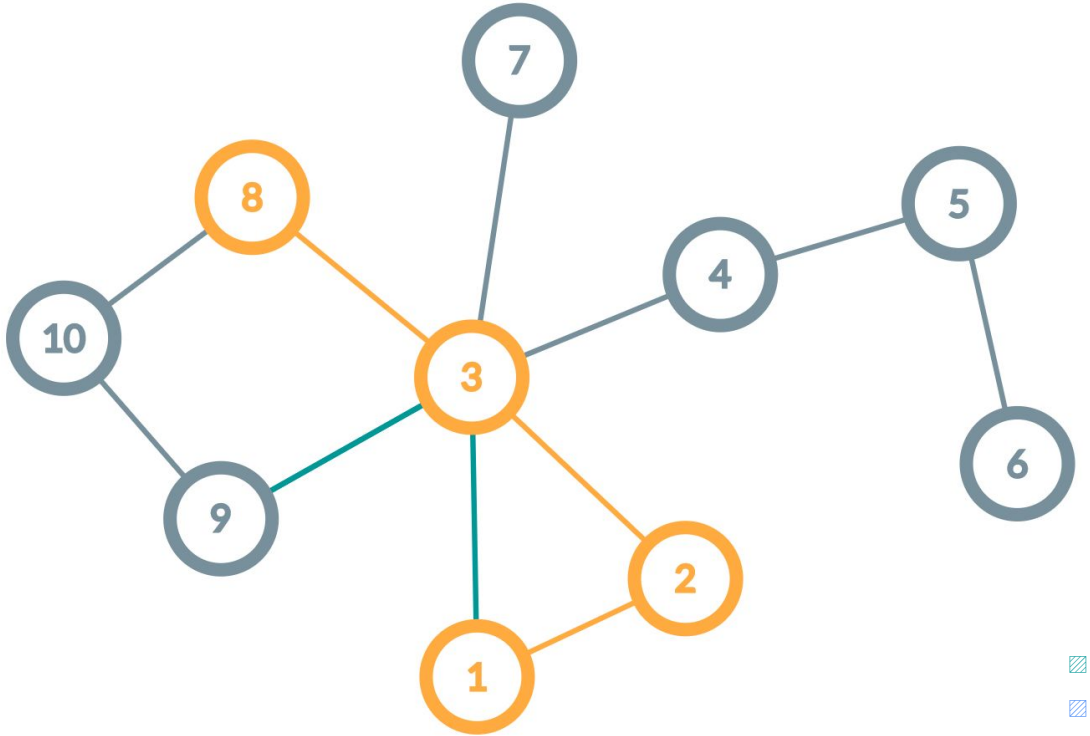


DFS(10)
DFS(8)
DFS(3)
DFS(2)
DFS(1)
Function call

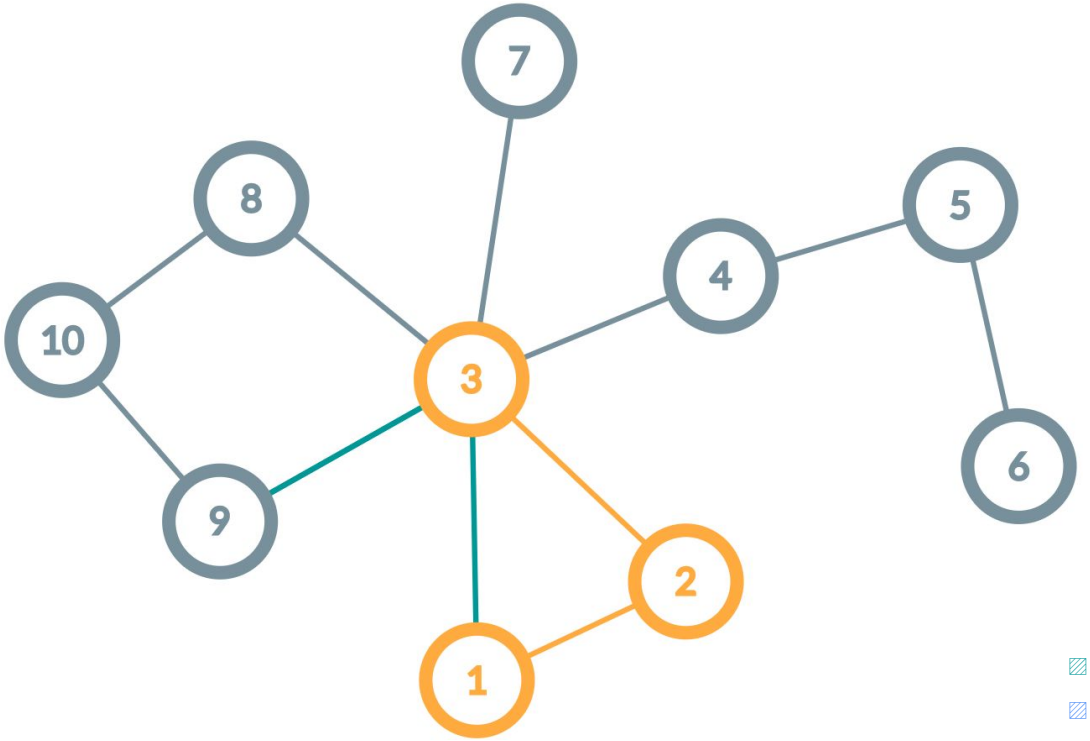


unvisited
visited
dead

DFS(8)
DFS(3)
DFS(2)
DFS(1)
Function call

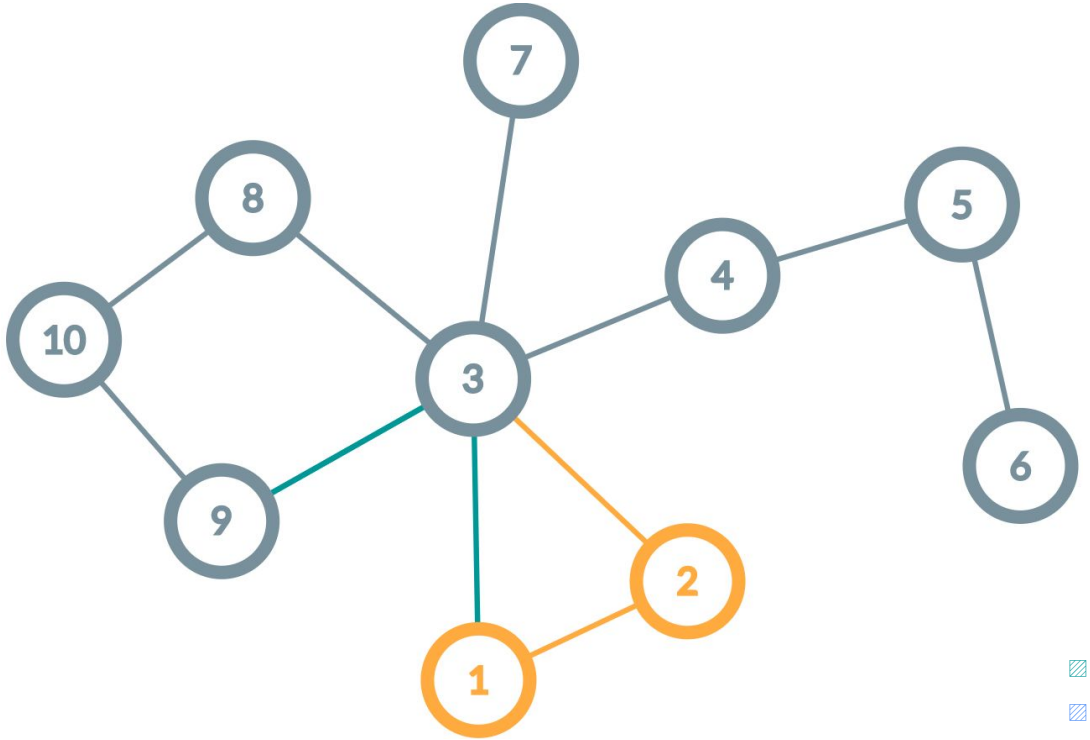


DFS(3)
DFS(2)
DFS(1)
Function call



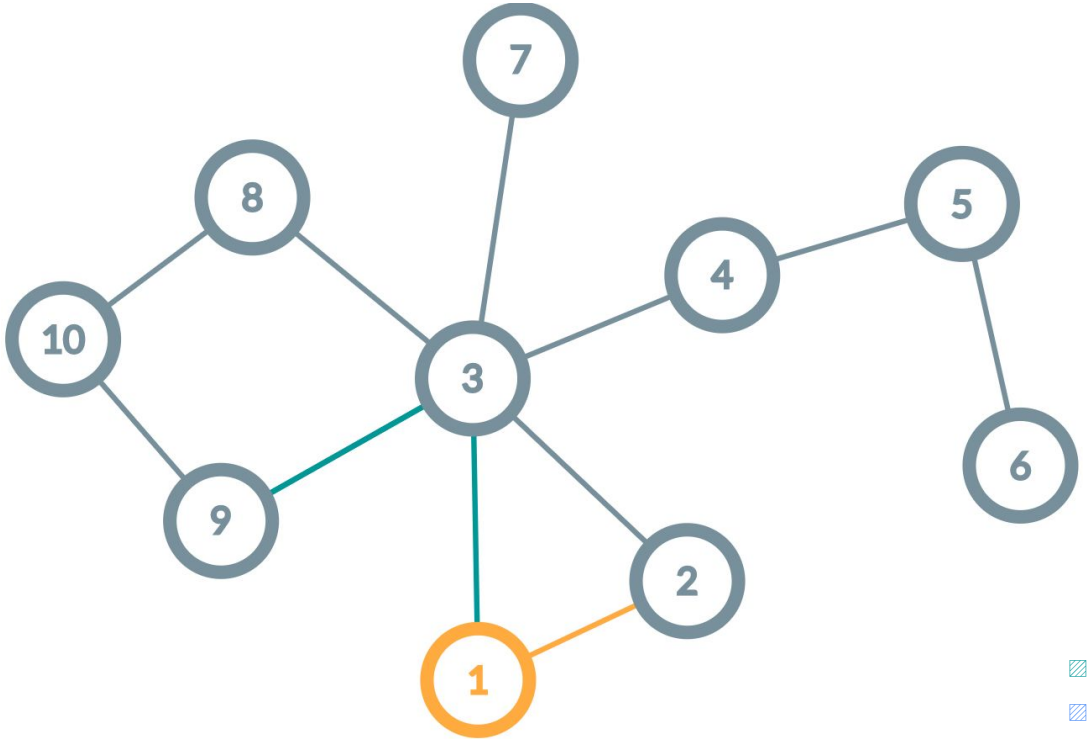
unvisited
visited
dead

DFS(2)
DFS(1)
Function call

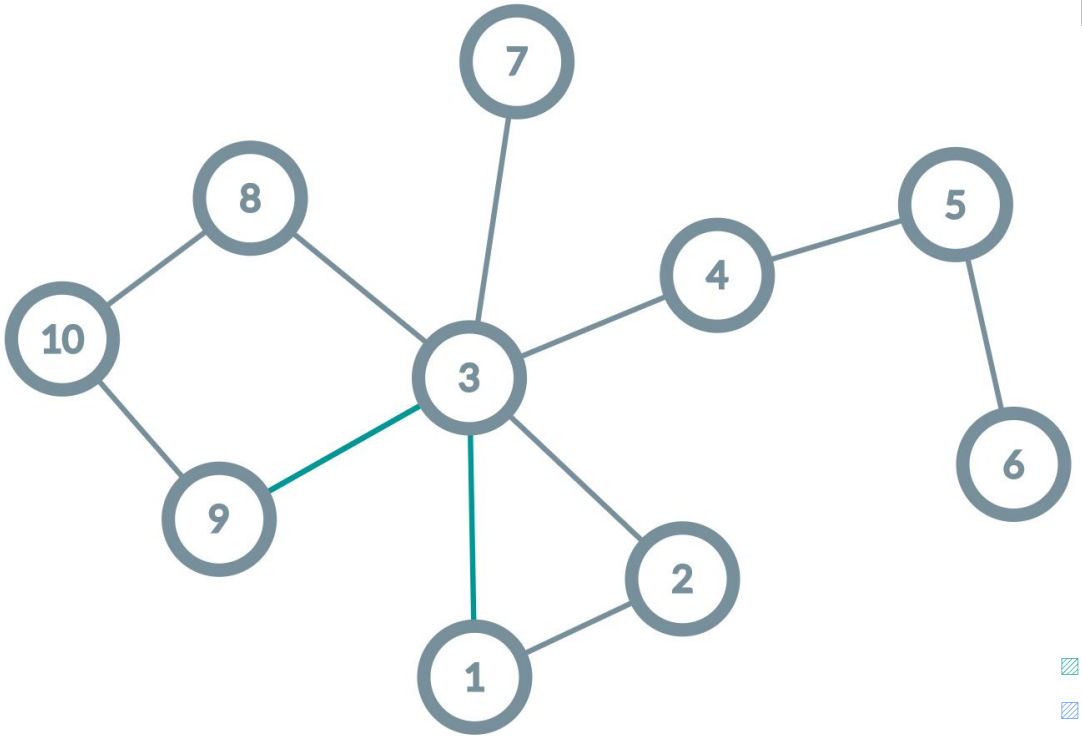


unvisited
visited
dead

DFS(1)
Function call



Function call



DONE!!

unvisited
visited
dead

Depth-First Search

(1) Iterative (stack)

```
push source vertex (denote as u) into stack S
while (S is not empty)
    pop the top element (x) in S
    if x is not visited
        push all the unvisited vertices that are neighbors of x into S
        mark x as visited
```

Depth-First Search

(2) Recursive *(more common)*

```
Procedure DFS(vertex x)
    mark x as visited
    For all vertices that are neighbors of x and are unvisited (v)
        Call DFS(v)

To start the search, call DFS(u) (source vertex)
```

Demo - HKOJ C800 Depth First Search

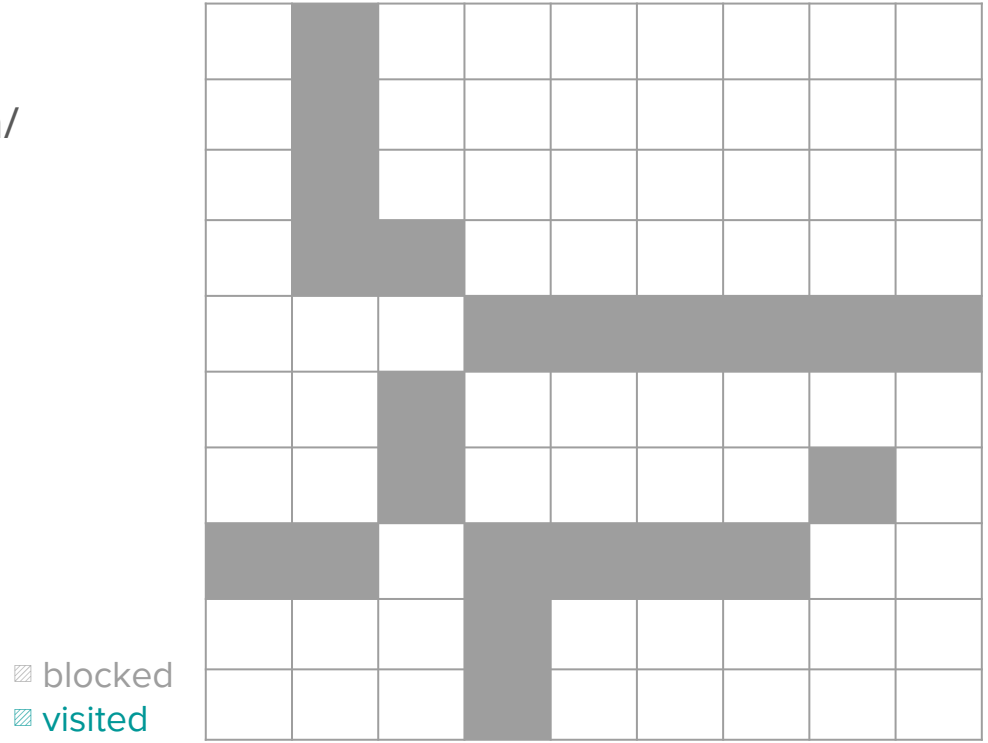
Flood Fill - DFS

From a empty cell, you can go up/down/ left/right to reach another empty cell.

Mutually reachable cells form a region.

How many regions are there?

How large is each region?

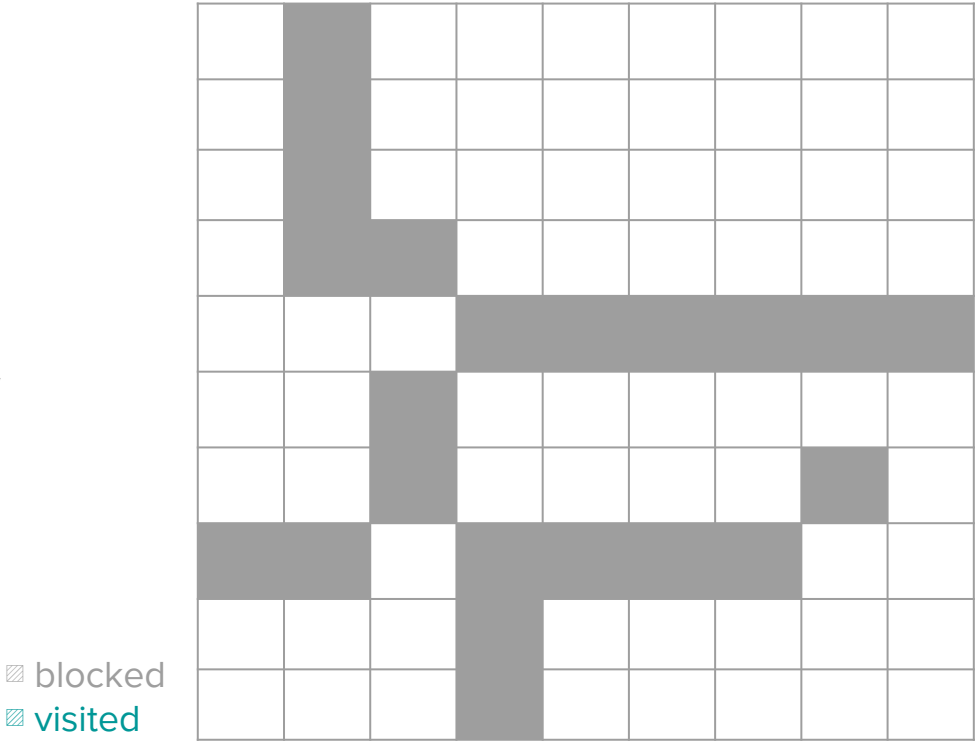


Flood Fill - DFS

We name each cell by its coordinate (row, column)

For example, top-left cell: (1, 1)
bottom-right: (10, 9)

When we detect an unvisited empty cell, we run DFS on it and count the number of cells in this region.



Flood Fill - DFS

We name each cell by its coordinate (row, column)

For example, top-left cell: (1, 1)
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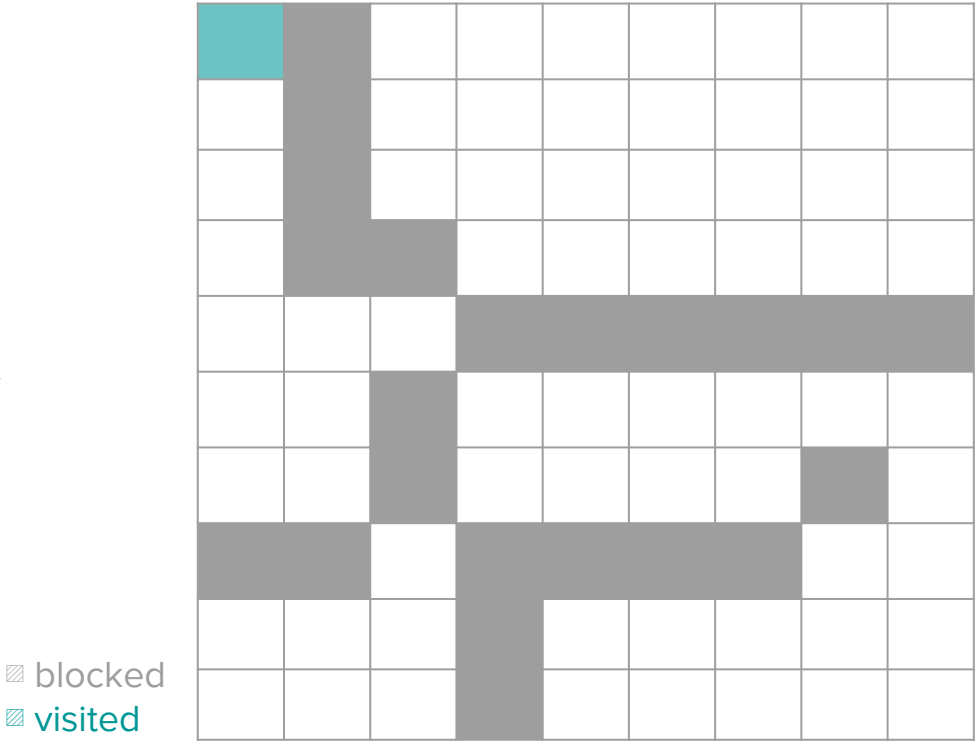
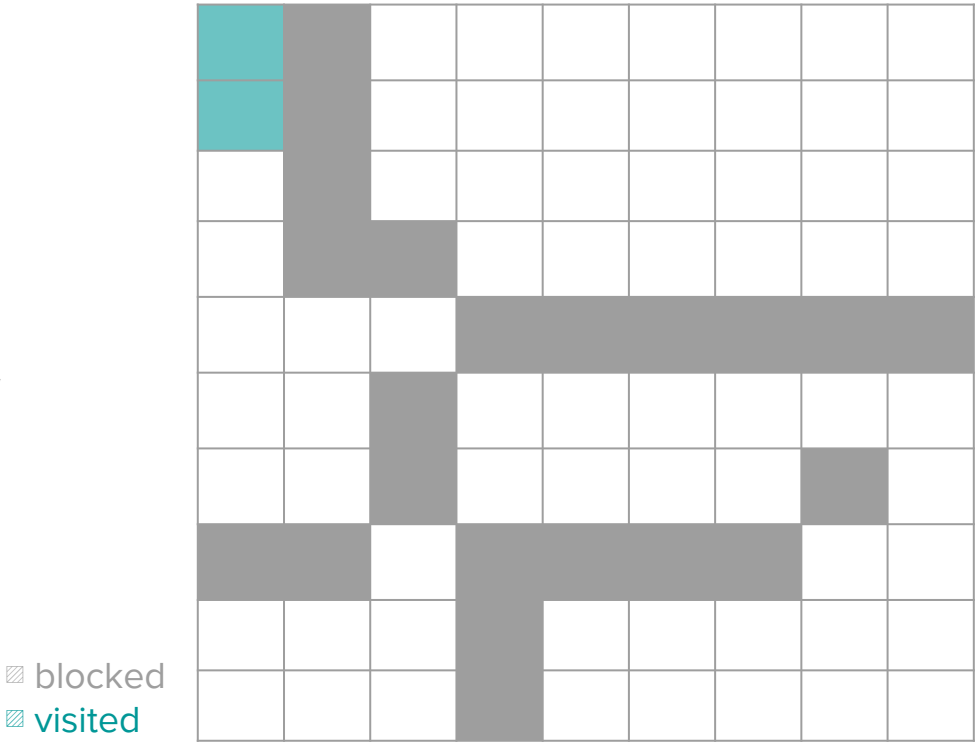


Flood Fill - DFS

We name each cell by its coordinate (row, column)

For example, top-left cell: (1, 1)
bottom-right: (10, 9)

When we detect an unvisited empty cell, we run DFS on it and count the number of cells in this region.

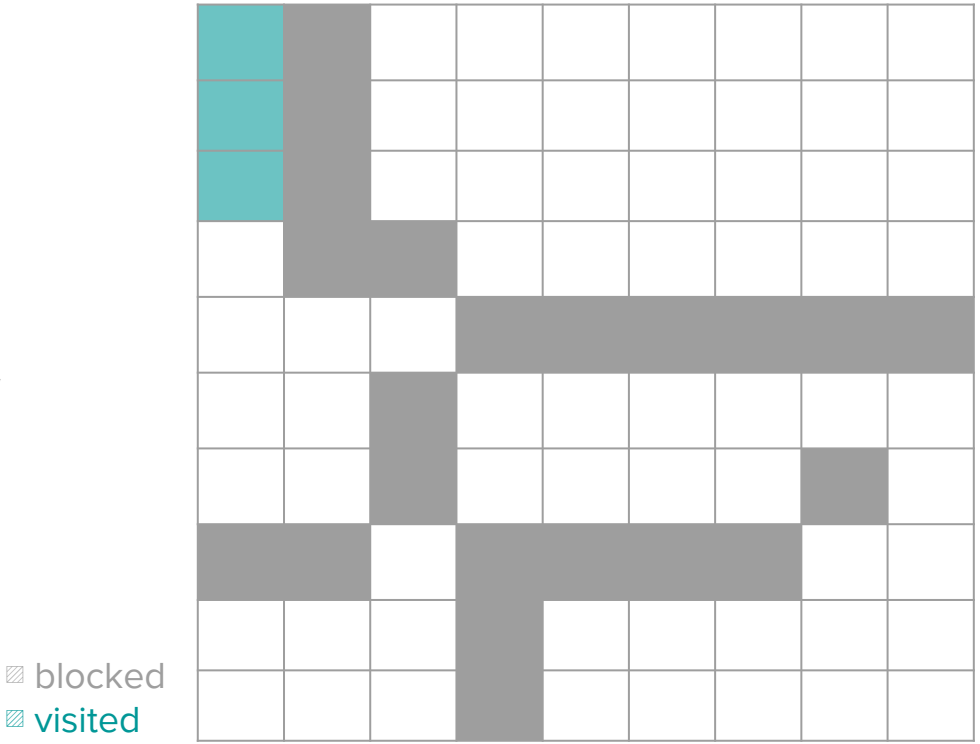


Flood Fill - DFS

We name each cell by its coordinate (row, column)

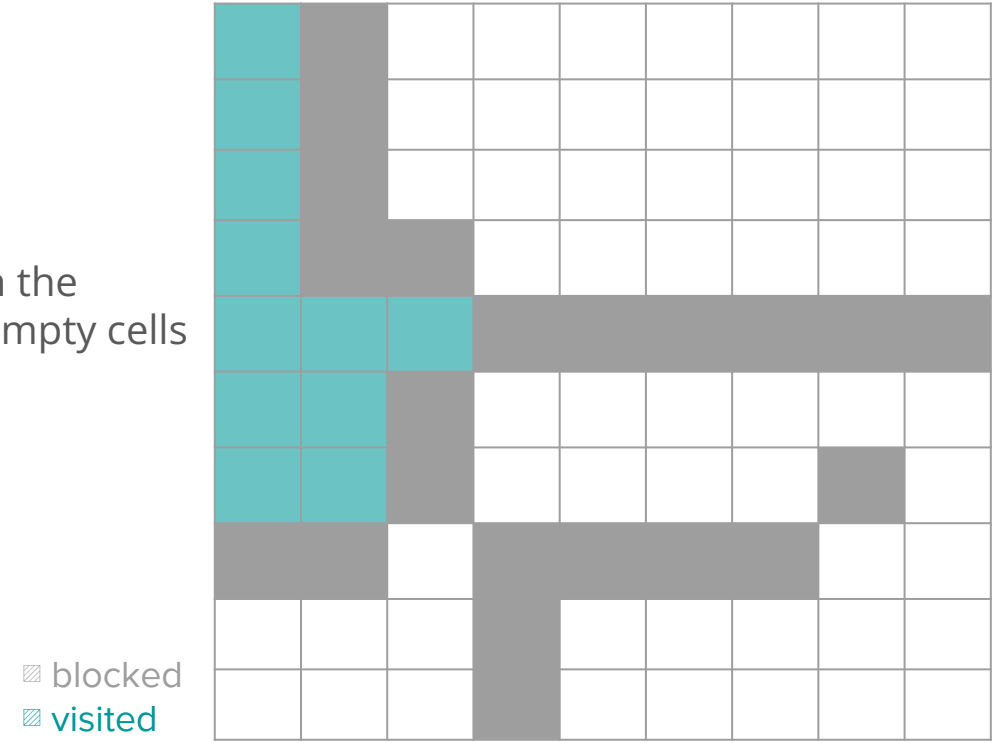
For example, top-left cell: (1, 1)
bottom-right: (10, 9)

When we detect an unvisited empty cell, we run DFS on it and count the number of cells in this region.



Flood Fill - DFS

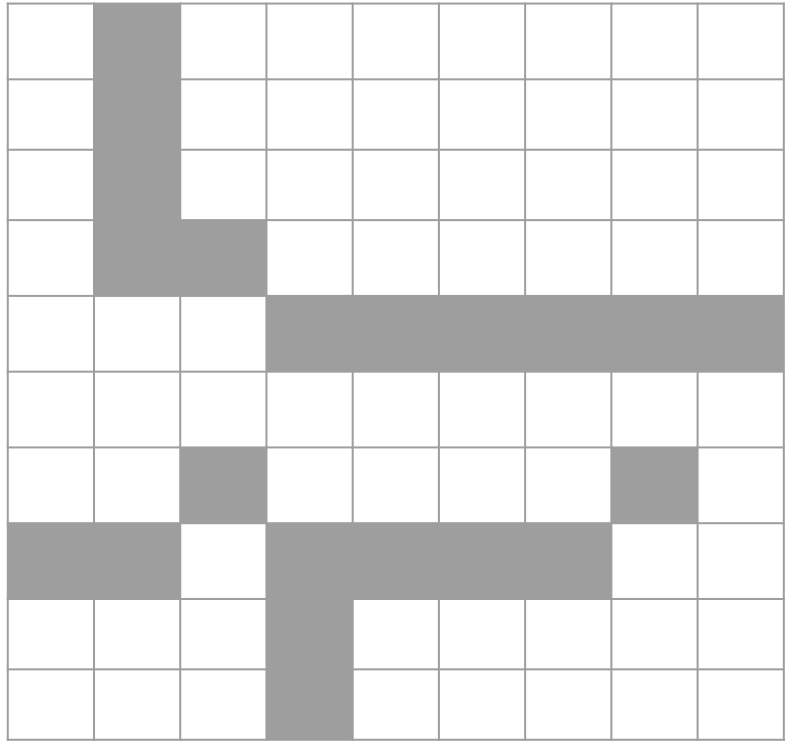
After visiting a region, we perform DFS on the remaining unvisited empty cells until all empty cells are visited.



Maze - DFS

Find a way from the top-left corner (1, 1) to the bottom-right corner (10, 9)

You can go up / down / left / right.



blocked

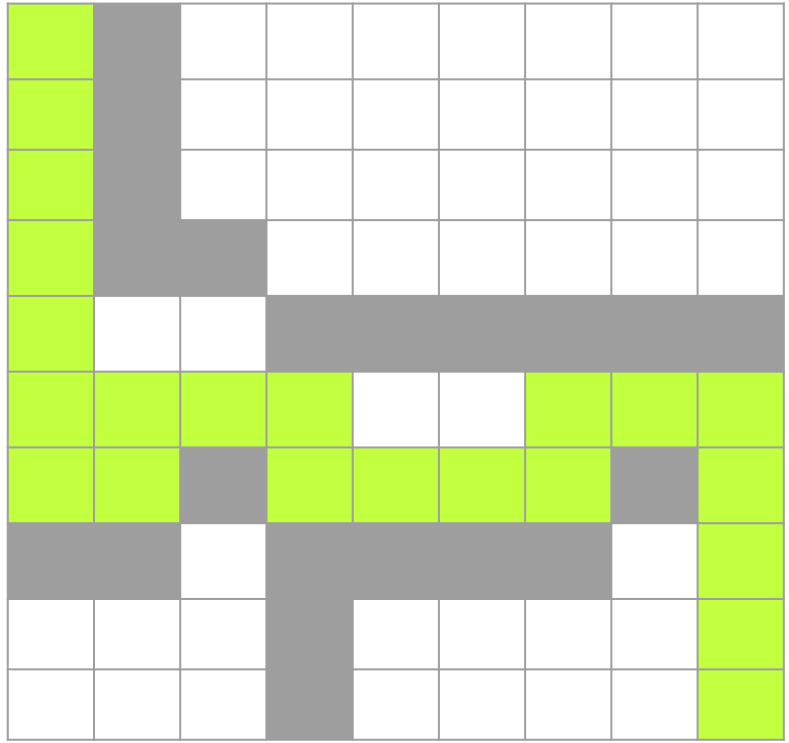
one of the possible paths

Maze - DFS

Call DFS(1, 1)

When the destination is reached, backtrack the path.

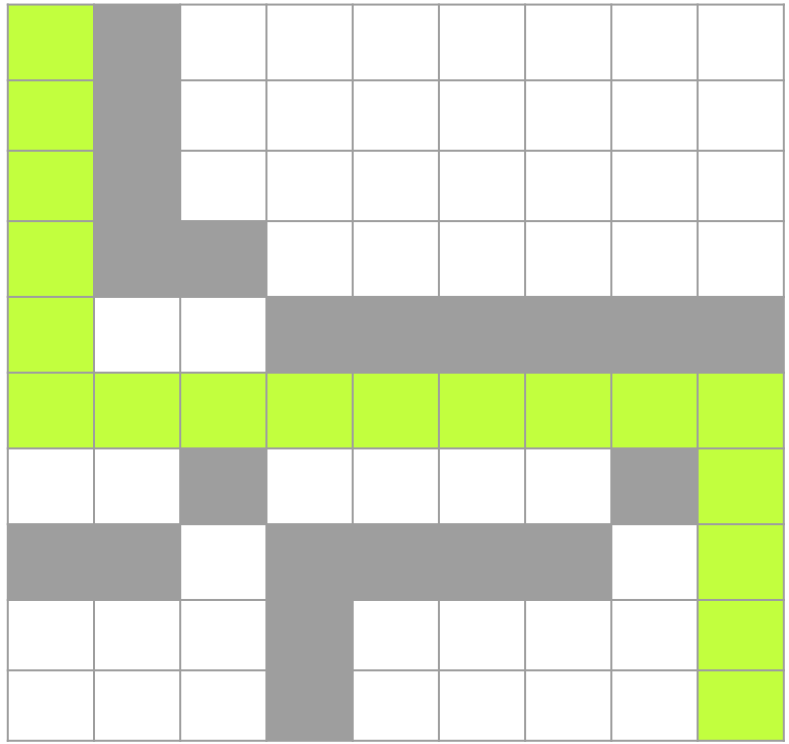
To backtrack, maintain a list containing parent of each nodes in the current path



Maze - DFS

Note that the path **may not** be shortest

(To find shortest paths using concepts similar to DFS, read about Iterative Deepening DFS, it was covered in 2018 Graph II materials)
– Fixing depth of traversal for each search



blocked
one of the possible paths

Demo - HKOJ C801 Flood Fill

Breadth-First Search

Though Depth-First Search can provide us a path from *u* (source node) to *v* (any node reachable), it may not be the shortest path.

Given the graph is **unweighted**, we can use Breadth-First Search (BFS) to find the shortest path from *u* to *v*.

For weighted graph shortest path, refer to Graph (II).

Breadth-First Search

Starting from the source vertex, we visit all reachable nodes using minimum of *i* steps.

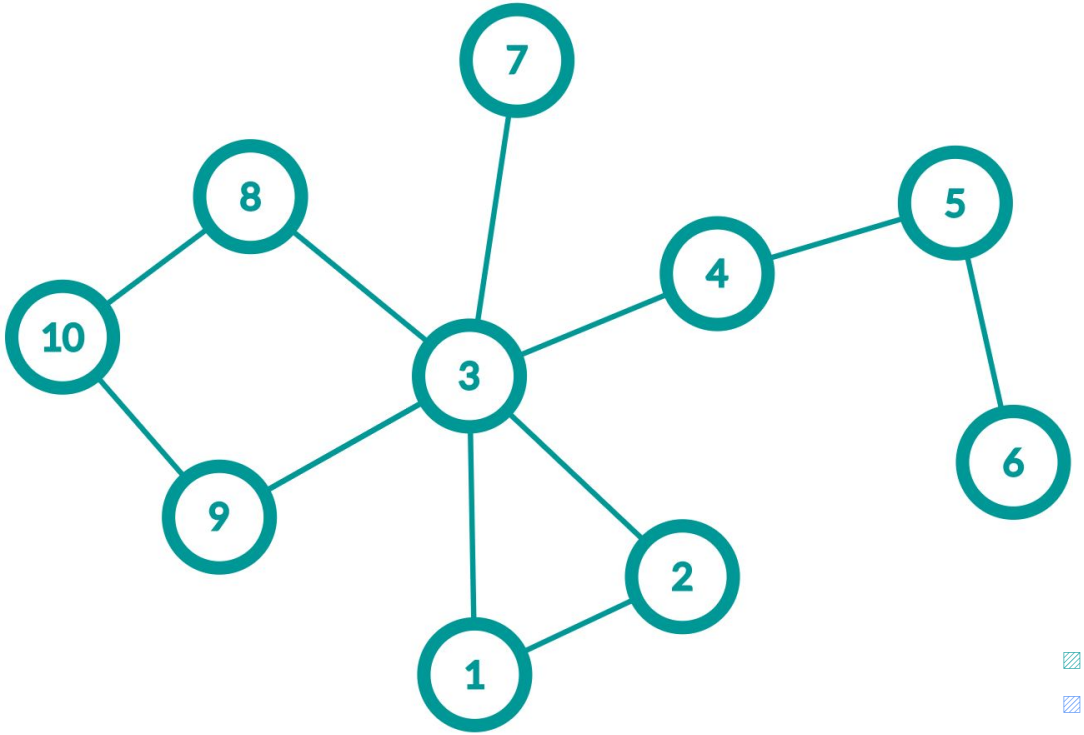
(1) Visit all the neighbors of *u* {x : set of vertices that are 1-step reachable from *u*}

(2) Visit all the unvisited neighbors of {x} {y : set of vertices that are 2-step reachable from *u*}

(3) Visit all the unvisited neighbors of {y} {z : set of vertices that are 3-step reachable from *u*}

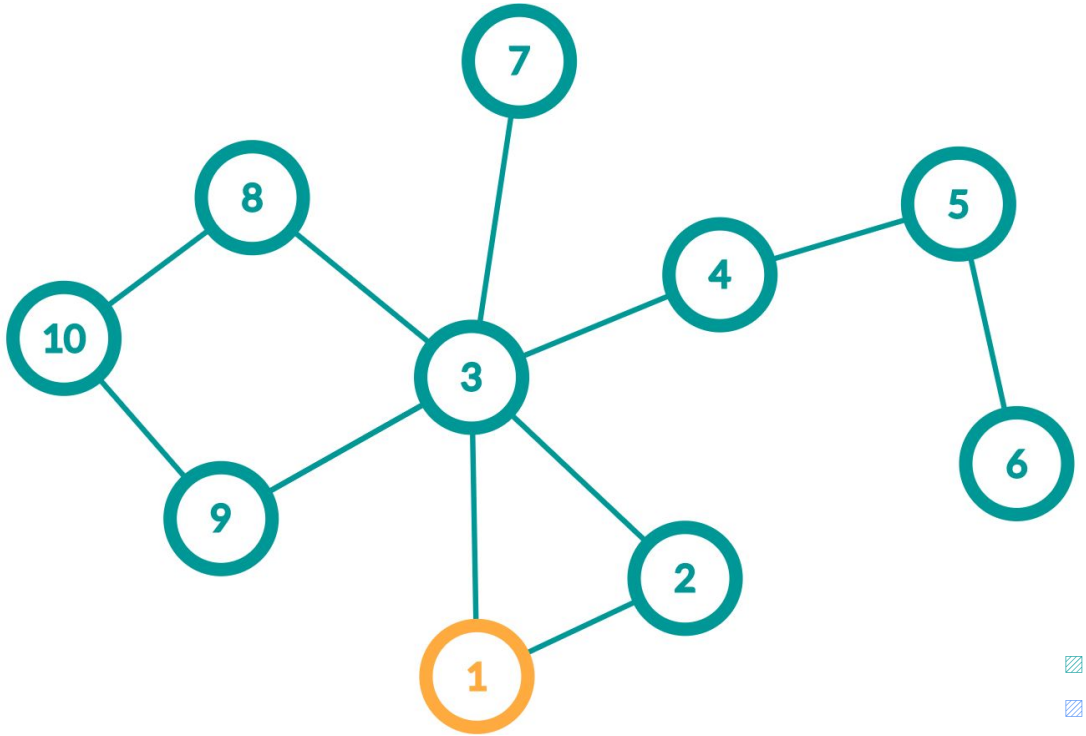
... until no new vertices are visited

Queue

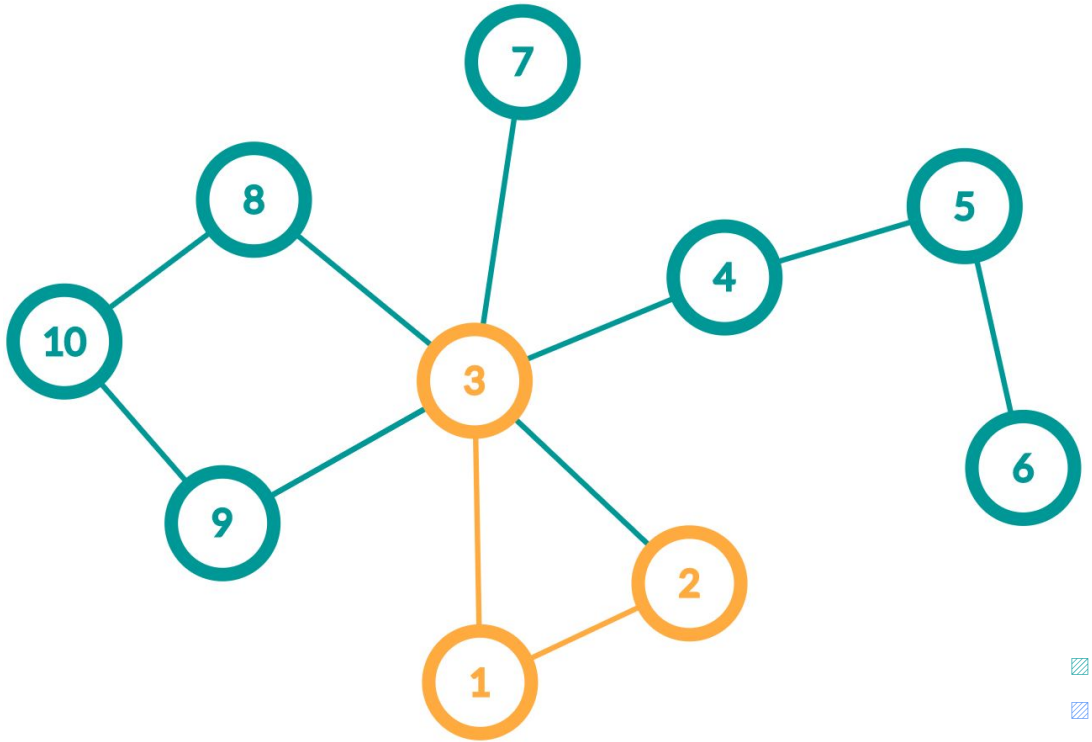


unvisited
visited
dead

Queue
1

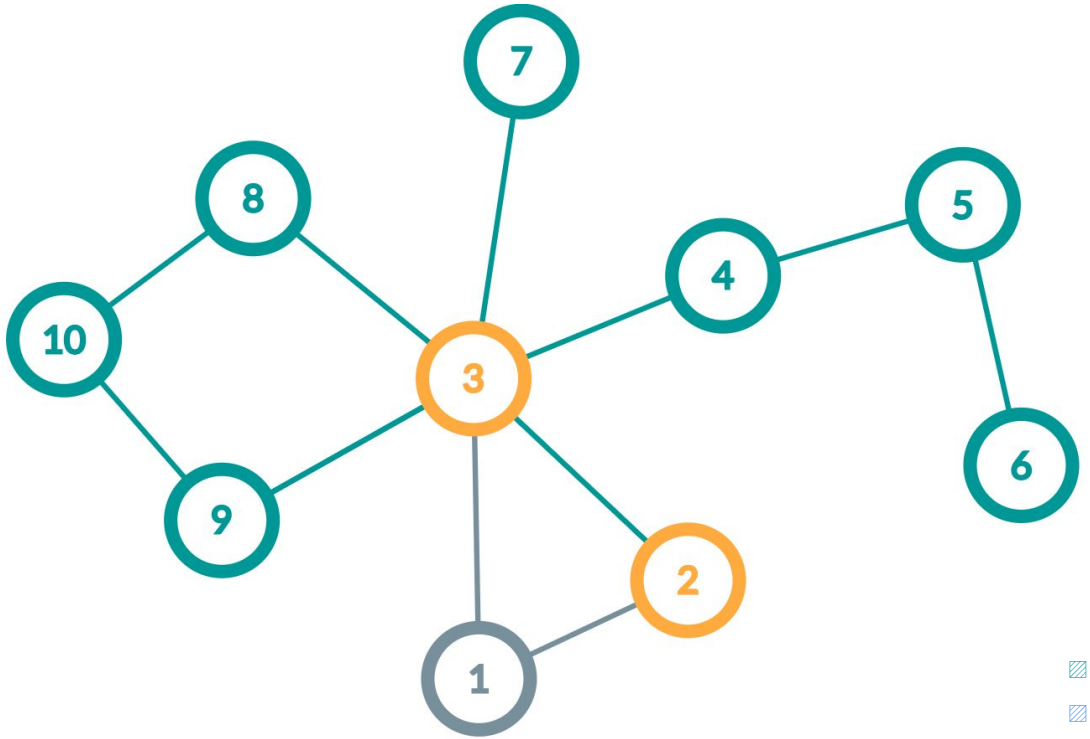


Queue
1
2
3



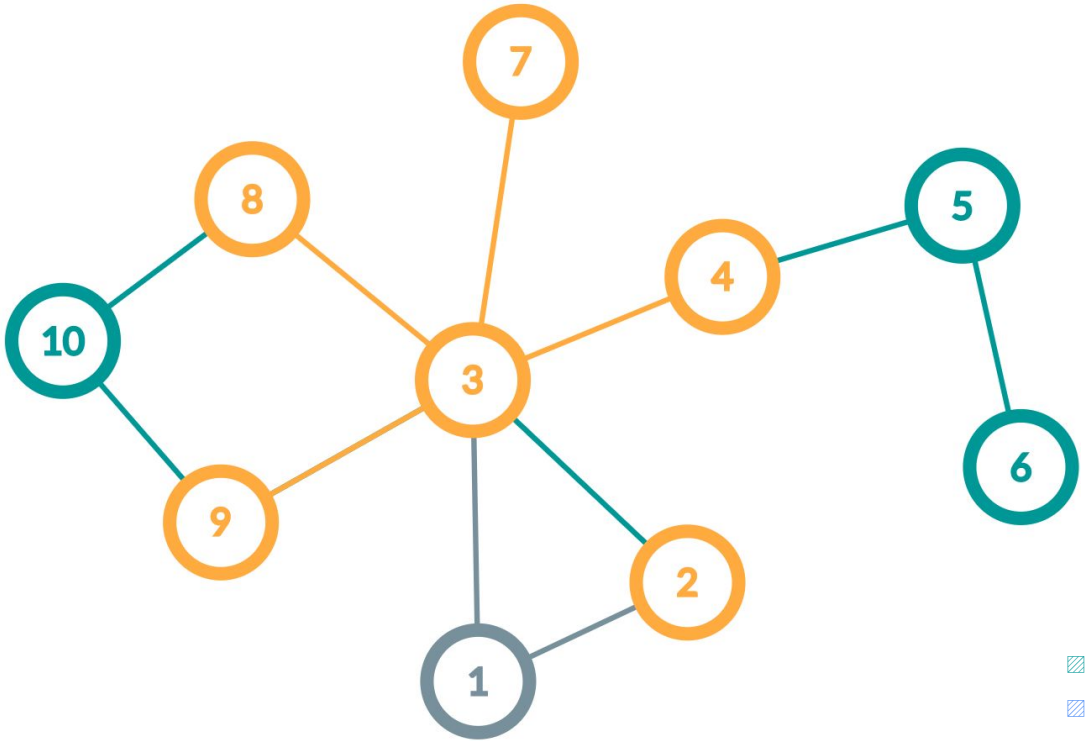
unvisited
visited
dead

Queue
2
3

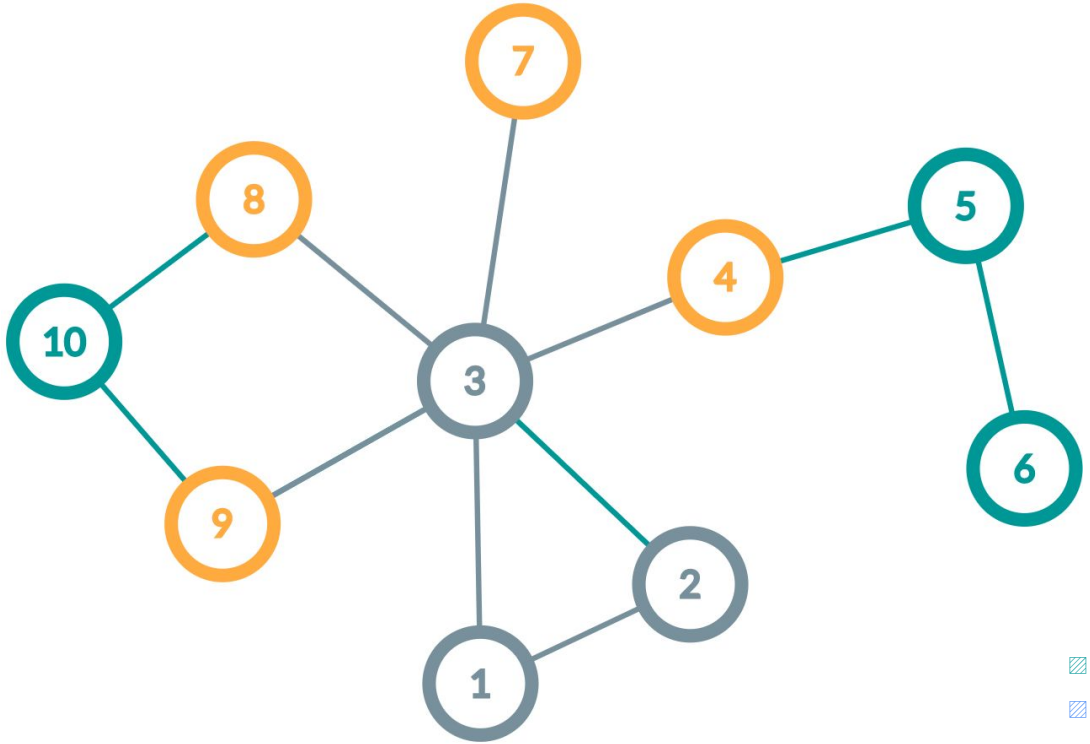


unvisited
visited
dead

Queue
2
3
4
7
8
9

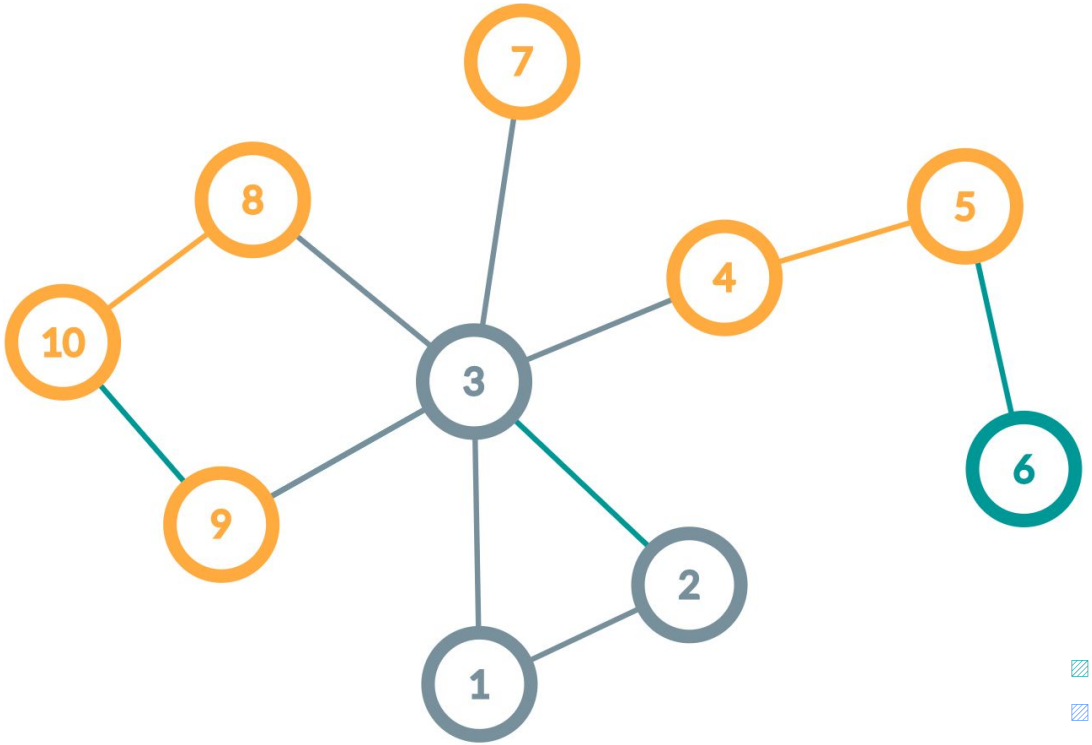


Queue
4
7
8
9



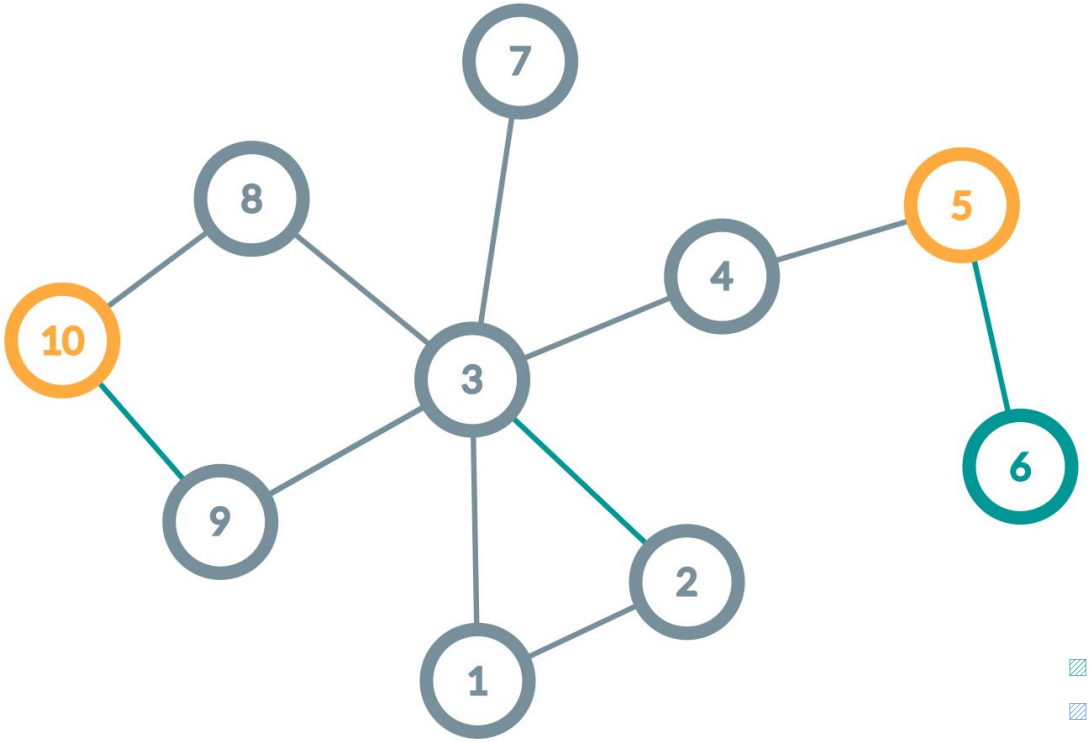
unvisited
visited
dead

Queue
4
7
8
9
5
10



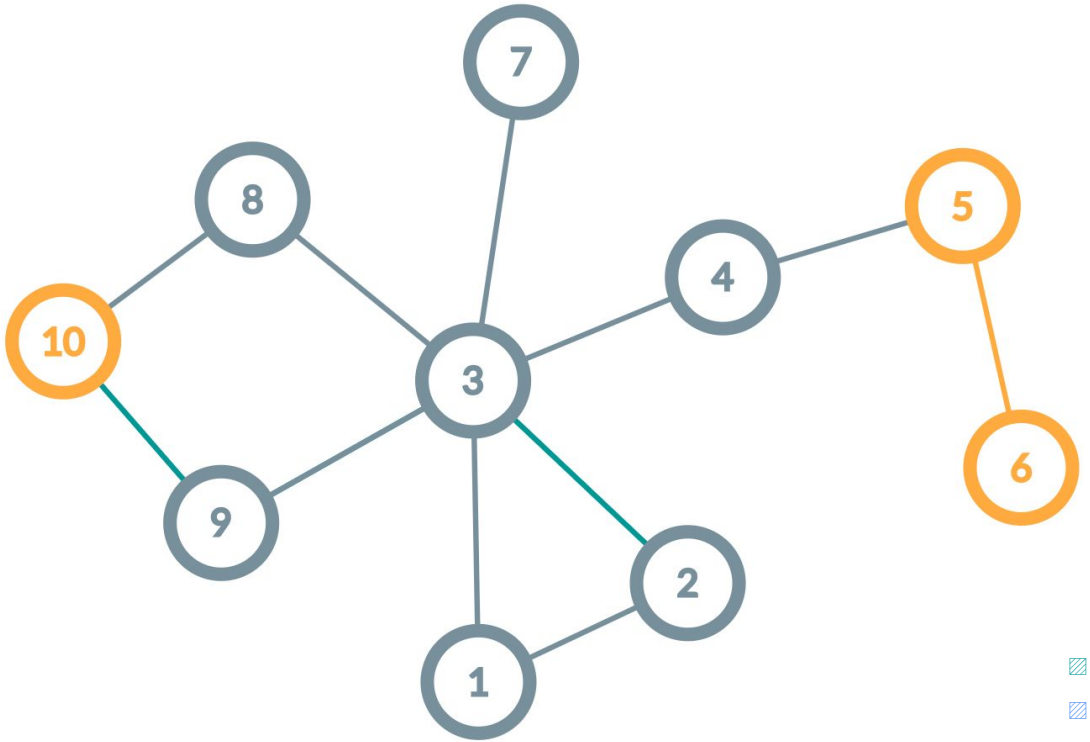
unvisited
visited
dead

Queue
5
10



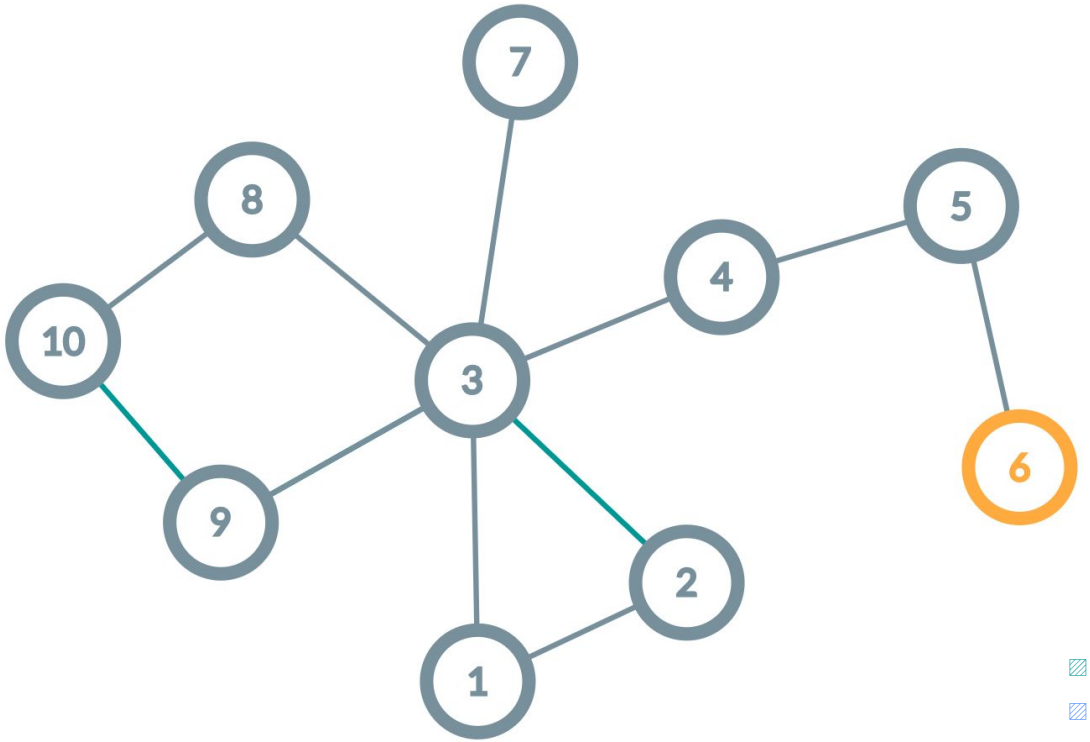
unvisited
visited
dead

Queue
5
10
6



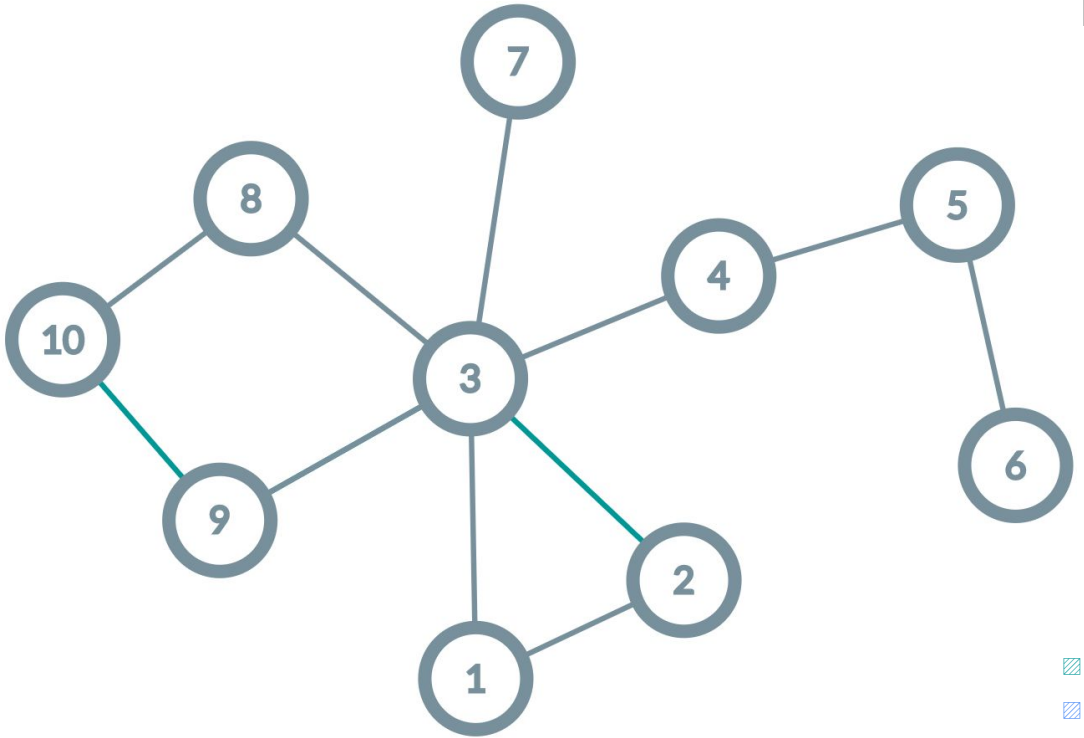
unvisited
visited
dead

Queue
6



unvisited
visited
dead

Queue



DONE!!

unvisited
visited
dead

Breadth-First Search

Push source vertex (denote as u) into queue Q
Mark u as visited
While (Q is not empty)
    Pop the front element (x) in Q
    Push all the **unvisited** vertices that are **neighbors of x** into Q
    Mark them as visited

Demo - HKOJ B200 Breadth First Search

BFS - Water Jug Problem

There are 2 water jugs with capacities N and M litres respectively.

Initially, both of them are empty.

You can perform the following operations for infinitely many times (one operation a time)

1. Empty a jug
2. Fully fill a jug
3. Pour water from one jug to another until either one jug is empty / full

How to get a specific volume K in one of the jugs?

BFS - Water Jug Problem

N = 3, M = 4, K = 2

States:

(0, 0)	(0, 1)	(0, 2)	(0, 3)	(0, 4)
(1, 0)	(1, 1)	(1, 2)	(1, 3)	(1, 4)
(2, 0)	(2, 1)	(2, 2)	(2, 3)	(2, 4)
(3, 0)	(3, 1)	(3, 2)	(3, 3)	(3, 4)

initial state
target states

Run BFS from (0,0)

BFS - Water Jug Problem

Transitions from (x, y)

1. Empty a jug (x, 0) and (0, y)
2. Fully fill a jug (x,M) and (N, y)
3. Pour water from the first jug to the second (x + y - M, M) or (0, x+ y)
4. Pour water from the second jug to the first (N, x+ y - N) or (x + y, 0)

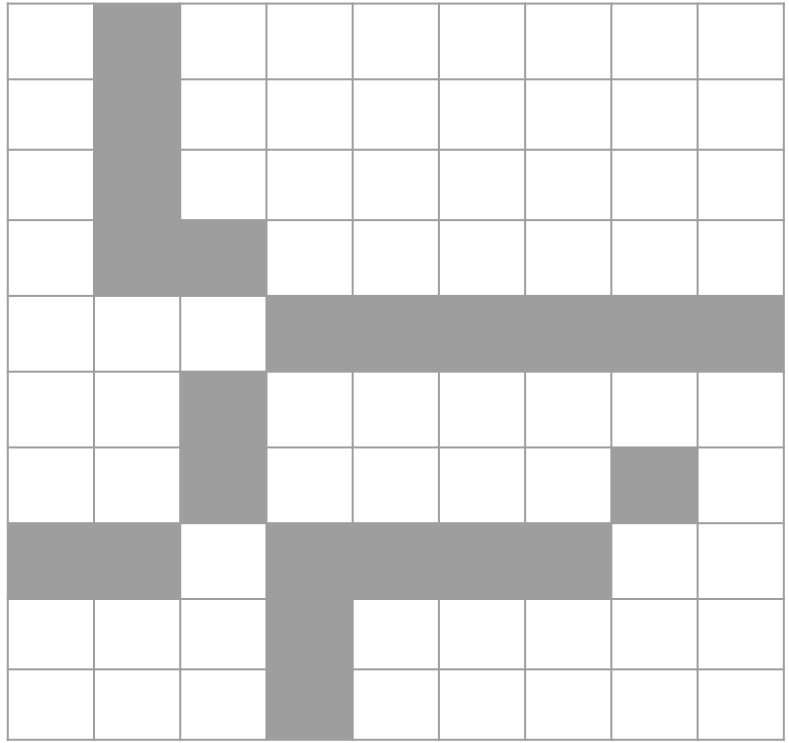
BFS - Maze Problem with Bomb

There is a N x M maze.

You can go U/ D/ L/ R one cell each move.

You have B bombs to destroy a wall of one cell

Find the **shortest path** from (1, 1) to (N, M)



BFS - Maze Problem with Bomb

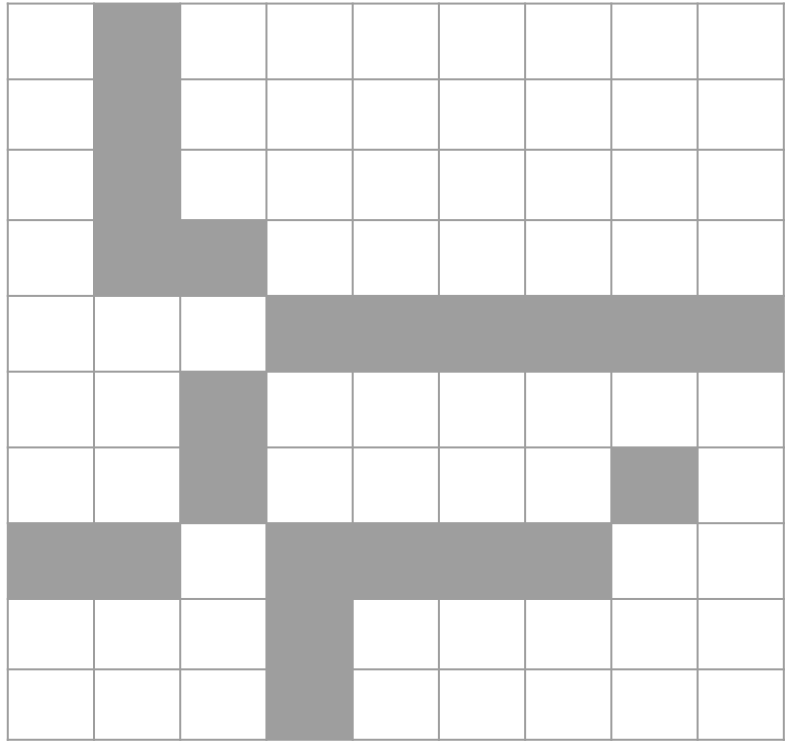
States: (x, y, b) [you are in cell (x, y) and you have b bombs]

Transitions

- (x, y + 1, b) if (x, y + 1) is not blocked
- (x, y + 1, b - 1) if (x, y + 1) is blocked and b is greater than 0

Same for other directions

Run BFS.



Multi-source BFS - Flood Fill

Given N sources, find the distance between each cell from the nearest source.

Method 1: BFS from one source at a time

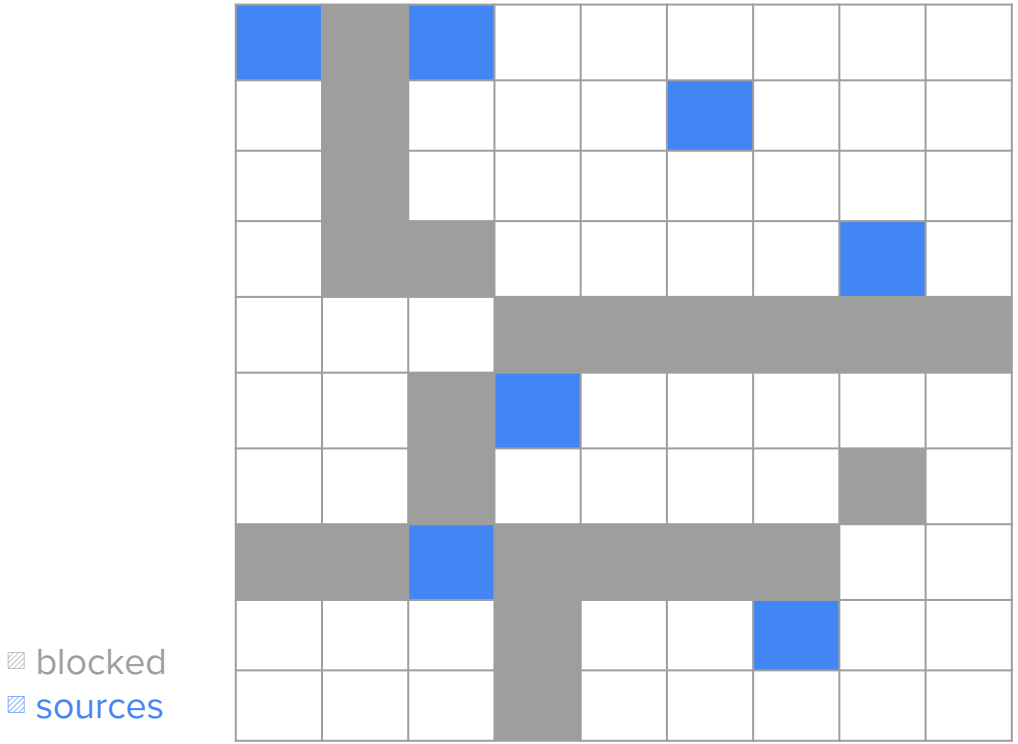
Time complexity: O(N * (V + E))

Method 2: BFS from all sources (perform BFS once)

Time complexity: O(V + E)

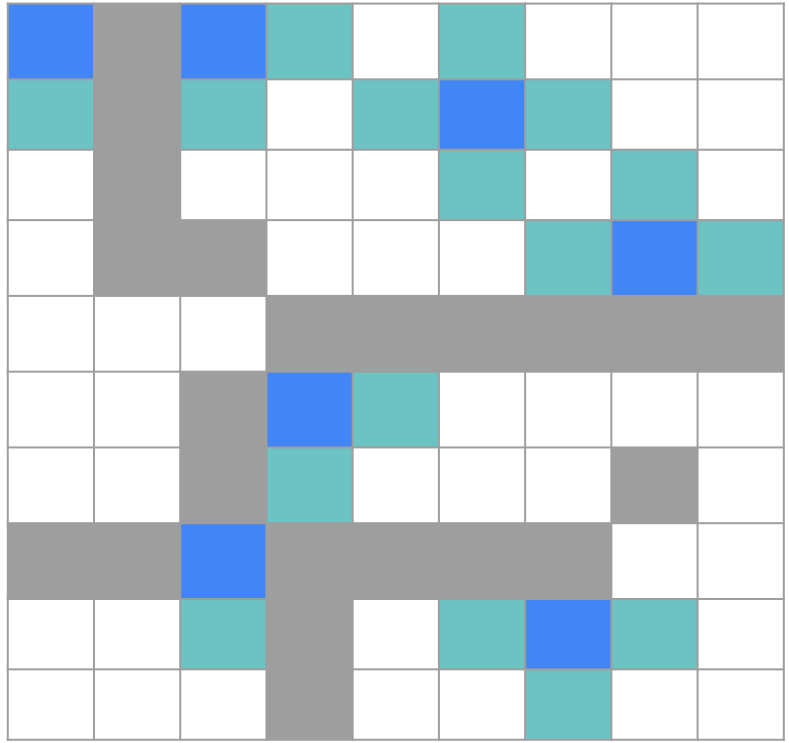
Multi-source BFS

Multi-source BFS on flood fill...



Multi-source BFS

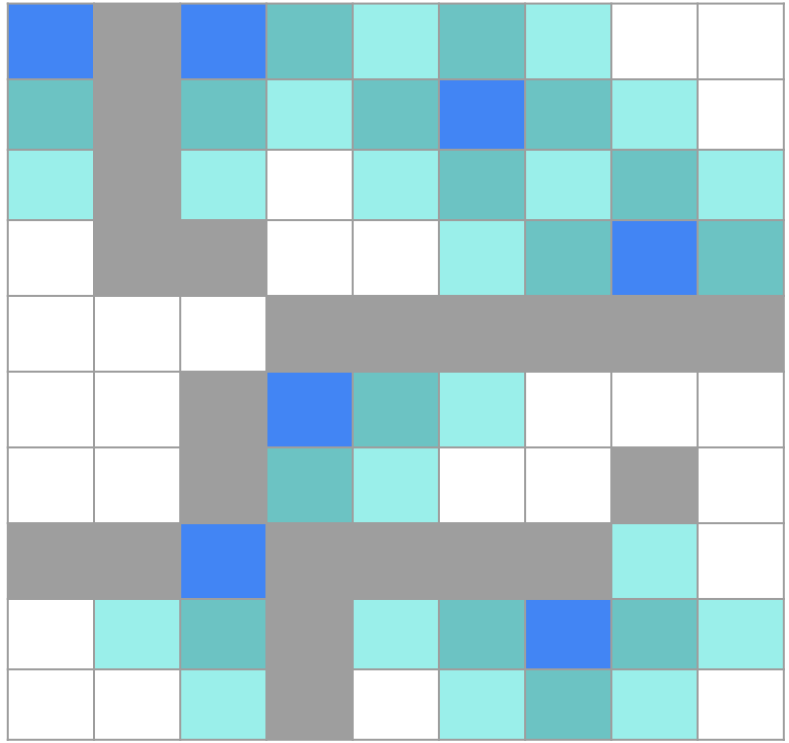
Multi-source BFS on flood fill...



blocked
dist = 1
sources

Multi-source BFS

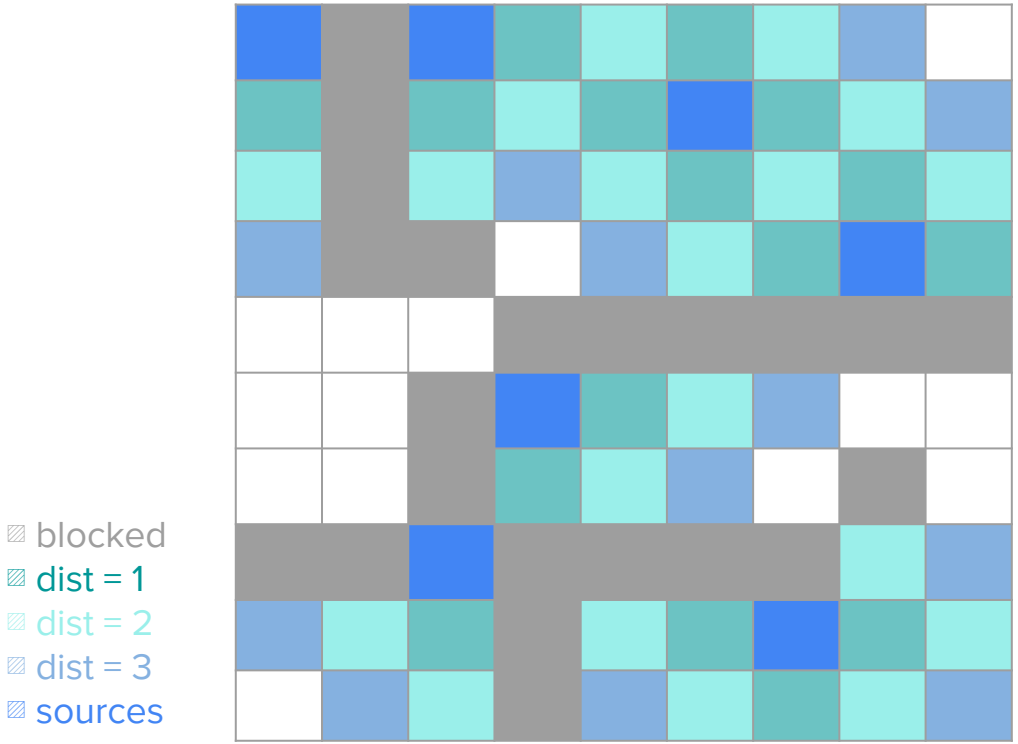
Multi-source BFS on flood fill...



Multi-source BFS

Multi-source BFS on flood fill...

And so on...



Demo - HKOJ B201 Multisource BFS

Bidirectional search

To speed up finding unweighted shortest path from u to v

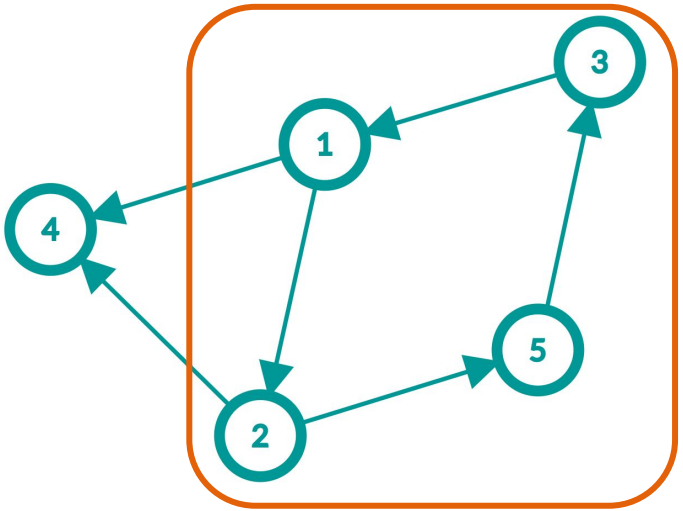
we can perform BFS with both u and v as sources

Stop when two paths meet in the middle

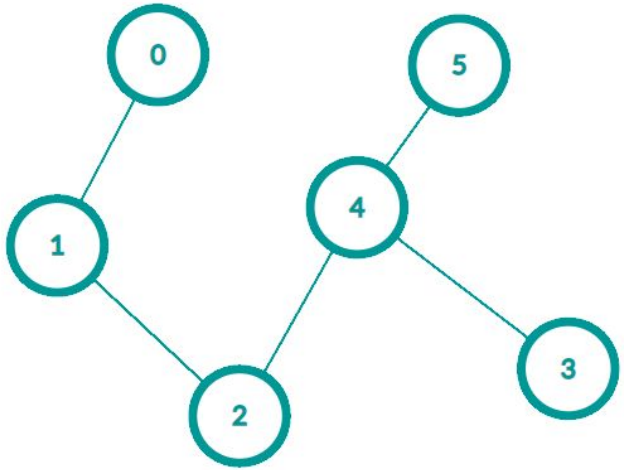
Special Graphs

Cycles

Graph with cycles

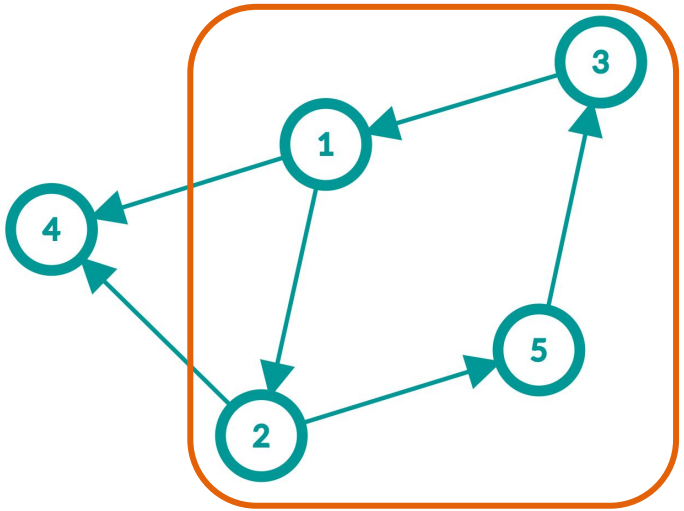


Acyclic graph



Cycles

Graph with cycles



How to detect cycles in a graph?

Recall searching, we avoid going to visited nodes again... why?

It indicates a cycle!

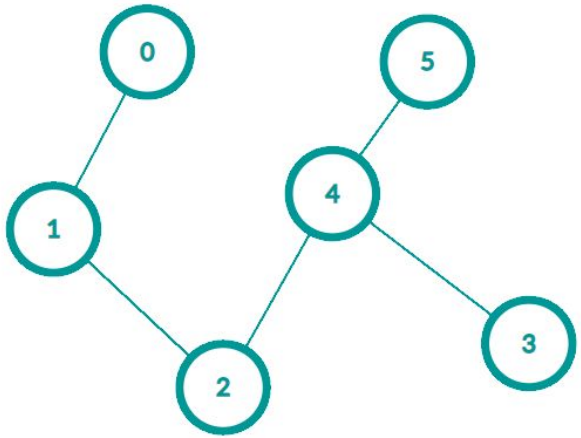
When we visit a node the second time, we know it is in a cycle

Trees

- A connected graph with |V|-1 edges
- A connected graph without cycles
- A graph with exactly one path between every pair of vertices

You will learn more about trees in Graph(II) and (III)

- Minimum spanning tree
- Lowest common ancestor
- Tree traversal
- ...



Chains

A tree with all vertices having 2 neighbors, except the two ends having only 1

Properties:

- Acyclic
- No branches

* Traverse from the ends

usually appear in earlier subtasks

Stars

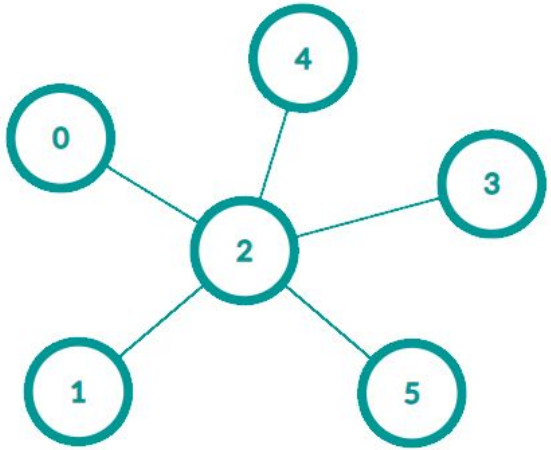
A tree with one internal node and remaining nodes as leaves

Properties:

- Acyclic
- Max distance between 2 nodes = 2

* Special handling on the internal node

usually appear in earlier subtasks

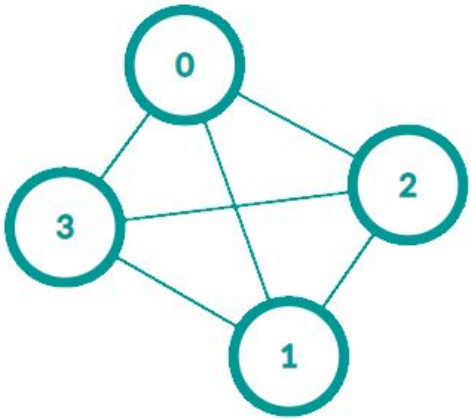


Complete Graph

Every pair of distinct vertices is connected by a unique edge

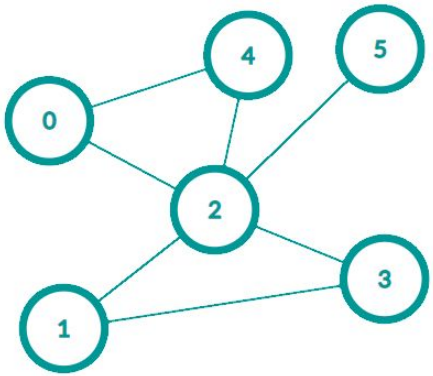
Properties:

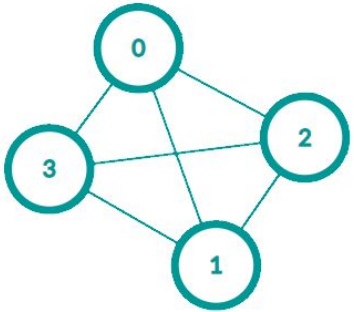
- `|E| = |V| * (|V| - 1) / 2`



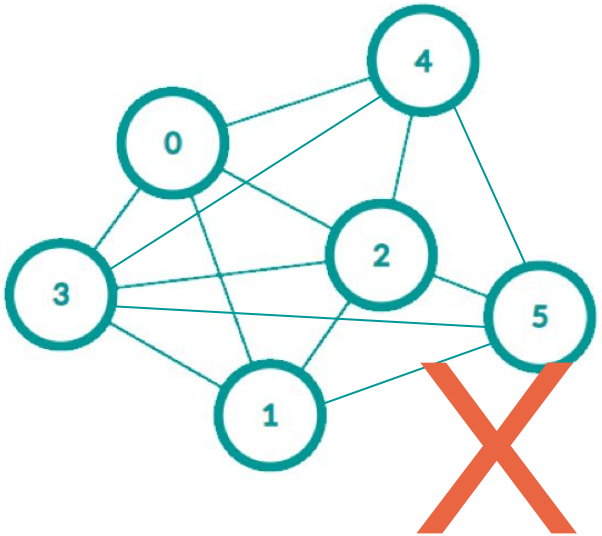
Planar Graph

Vertices and edges in such graph can be drawn in a plane such that no two edges intersect





(can move node 3 inside 0,1,2)



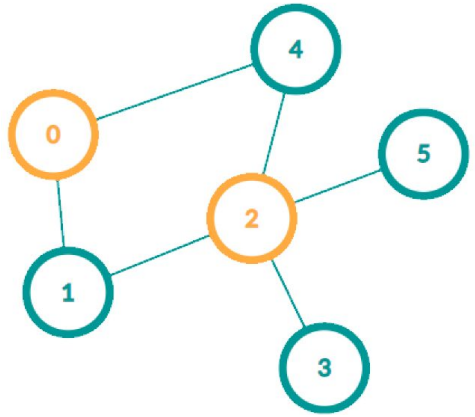
Bipartite Graph

Vertices can be divided into two disjoint and independent sets U and V, such that every edge connects a vertex in U to a vertex in V.

You can assign one of two colours for each node such that all edges have different colour on two side

Properties:

- Does not contain odd cycles



Bipartite Graph

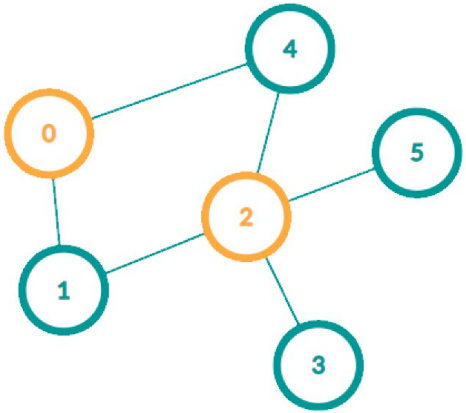
Algorithm to check if the graph is bipartite / color the graph into a bicolor graph

Perform DFS/BFS on any node

    Assign 0 and 1 alternately to every node according to the depth

    If a node is assigned 0, assign 1 to its neighbors, and vice versa

If any edge has same number on two sides, the graph is not bipartite



Attendance Taking

Practice Problems

HKOJ 01035 Patrol Area

HKOJ 01067 Maze

HKOJ M1311 Dokodemo Door

HKOJ T022 Bomber Man

HKOJ M0911 Theseus and the Minotaur

HKOJ 30422 Knights in FEN

CF 1033A King Escape

CF 893C Rumor

<dfs or equivalent / graph tag in codeforces>

演算法筆記

Graph Theory by Teb's Lab