

# Advanced Programming

## INFO135

Lecture 7: Graph

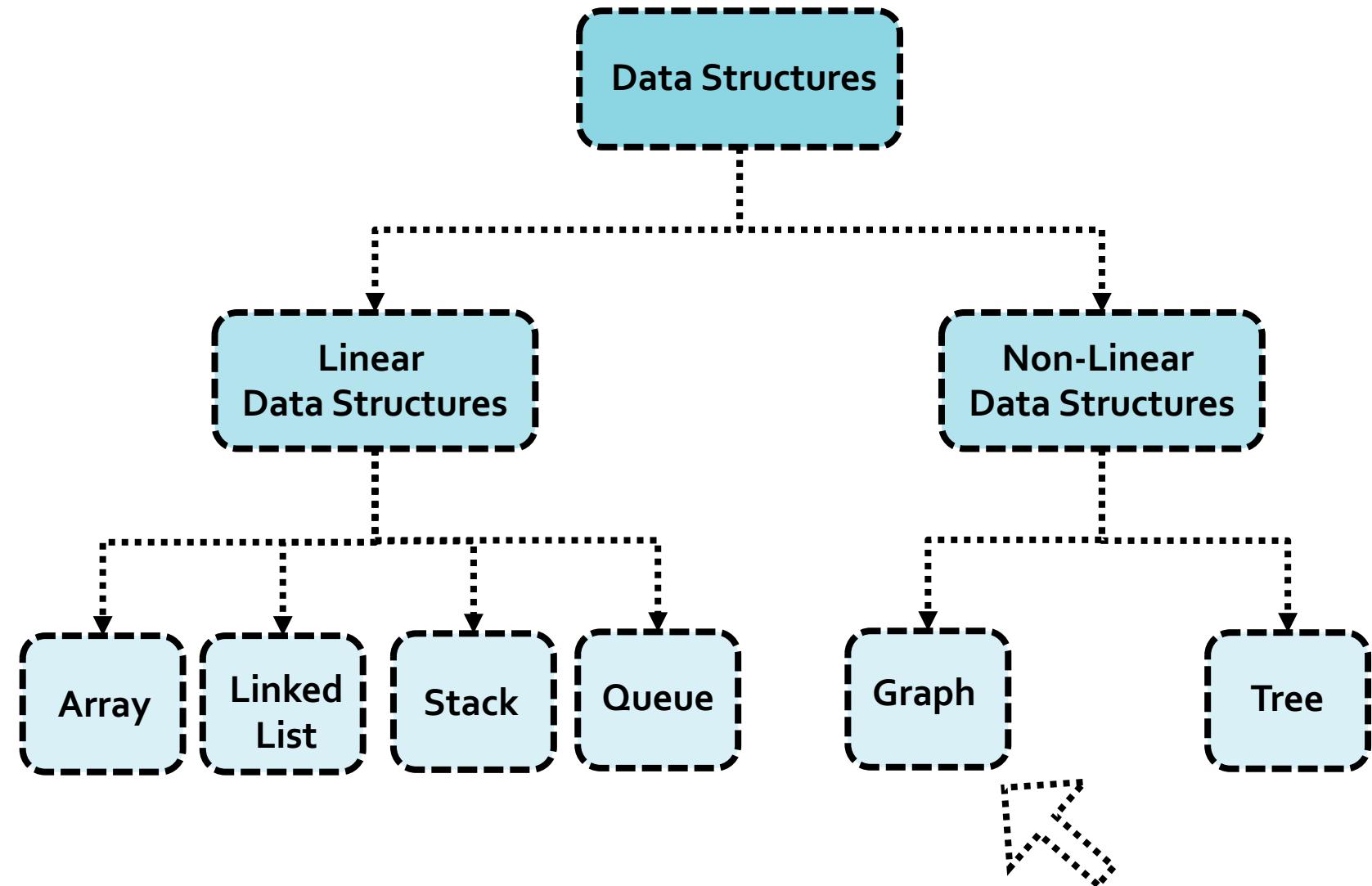
Mehdi Elahi

University of Bergen (UiB)

# Data Structures

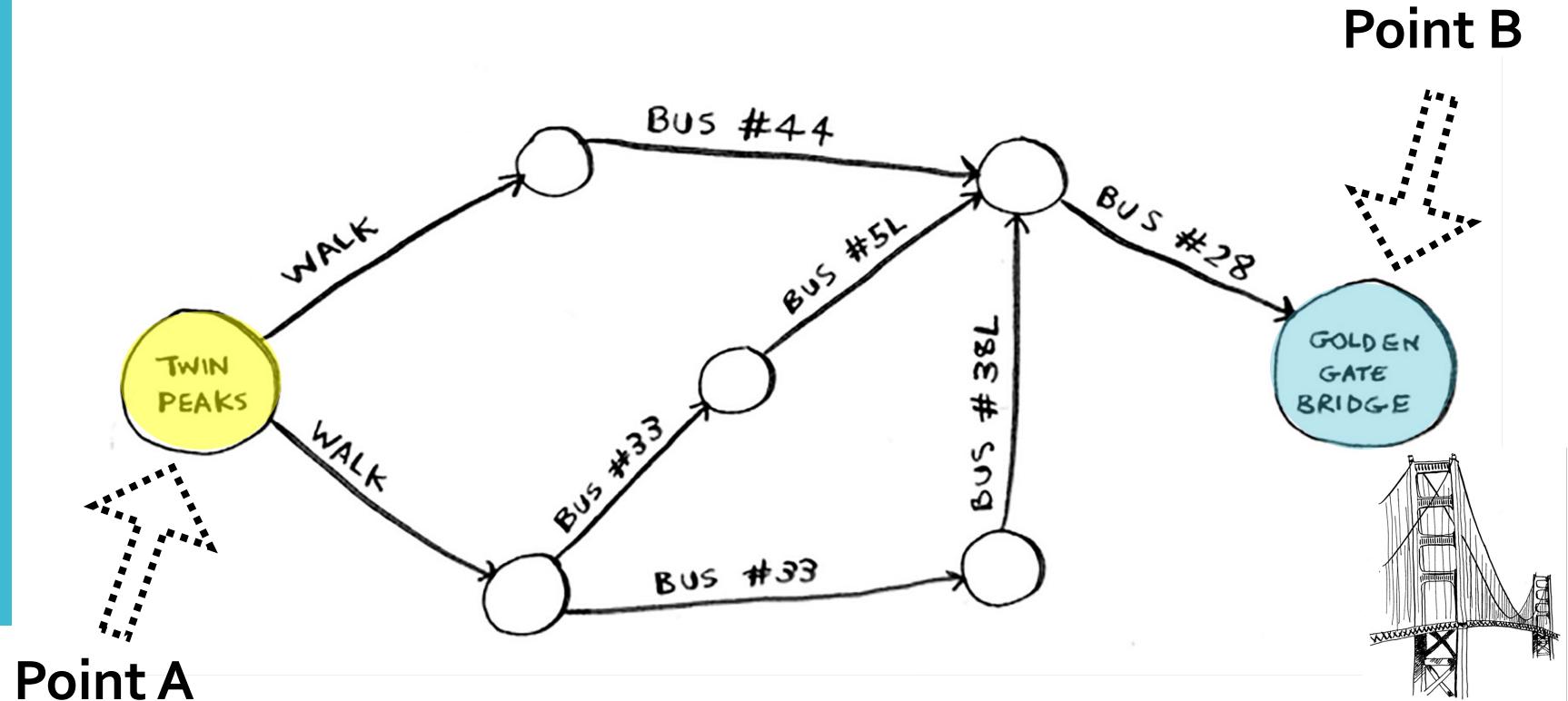
Linear Data Structure	Non-linear Data Structure
Data elements are arranged in a linear order where each element is attached to its <b>previous and next adjacent</b> .	Data elements are attached in <b>hierarchically</b> manner.
<b>Single level</b> is involved.	<b>Multiple levels</b> are involved.
<b>Simpler</b> and easier to implement	More <b>complex</b> and difficult to implement.
Data elements <b>can be traversed</b> in a <b>single run</b> .	Data elements <b>can't be traversed</b> in a <b>single run</b> .

# Data Structures



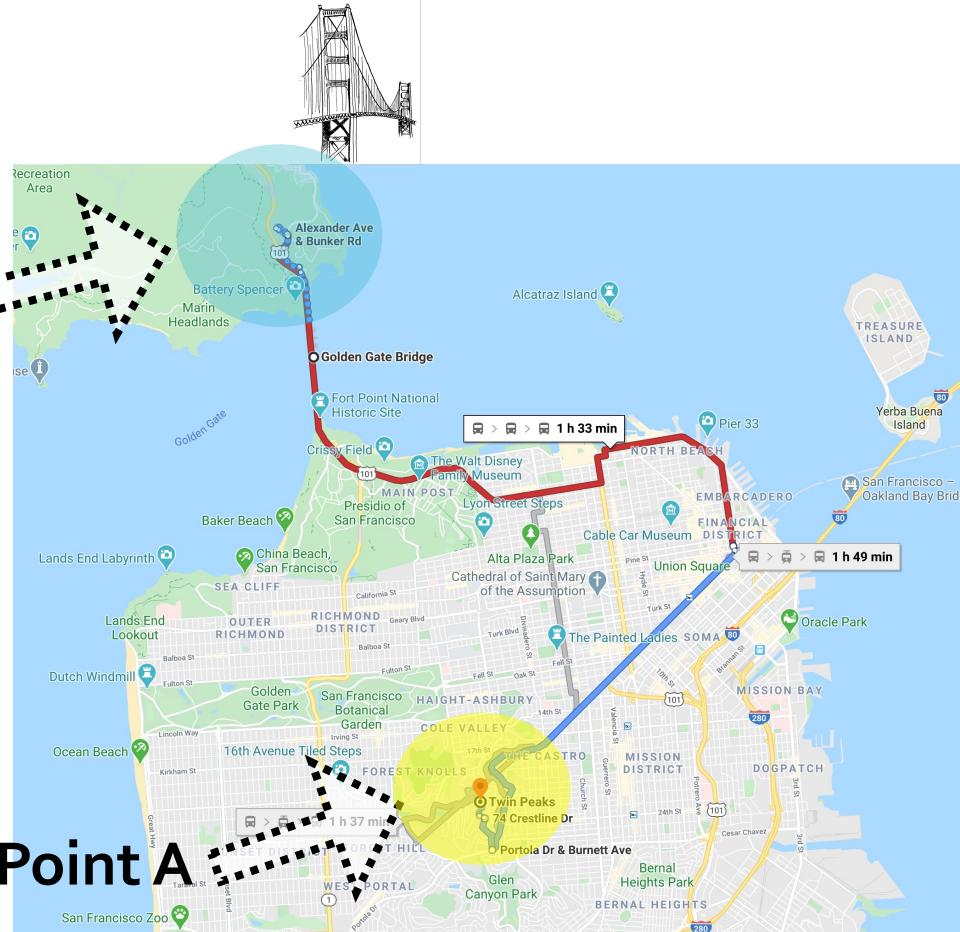
# Graphs

- ❖ Suppose you're in San Francisco, going from **point A** (Twin Peaks) to **point B** (Golden Gate Bridge).
- ❖ You want to get there by bus, with the **fewest steps** (transfers).



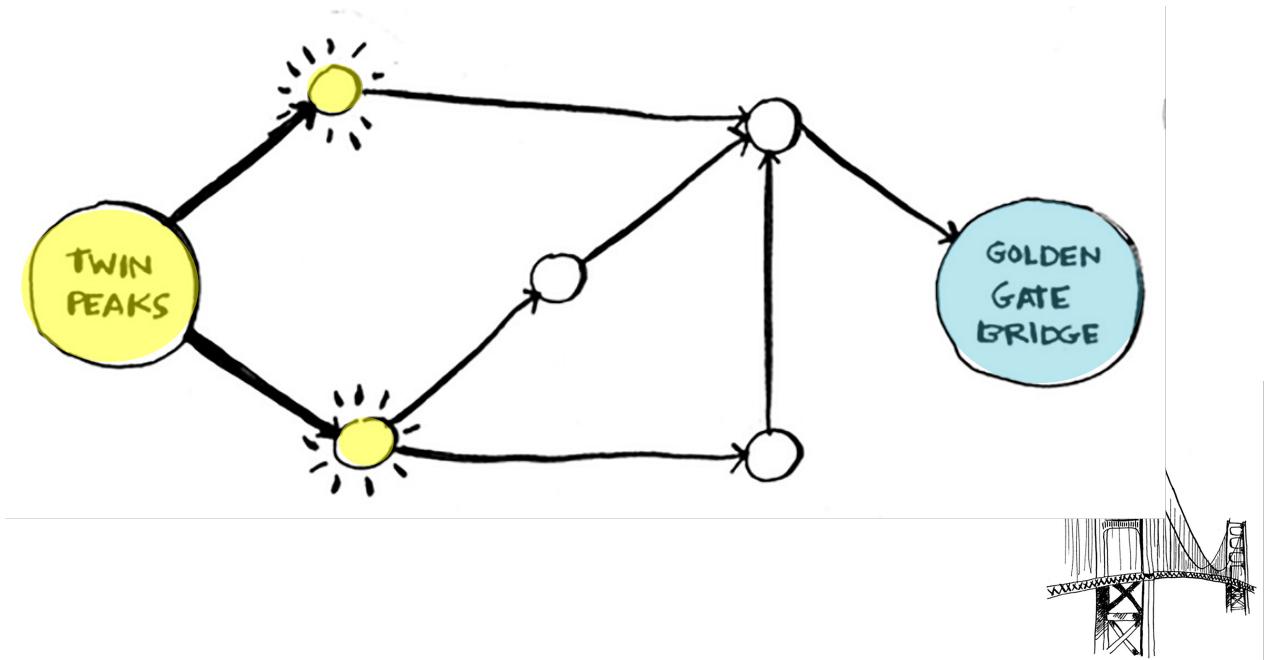
# Graphs

❖ In Google map this will look like the following.



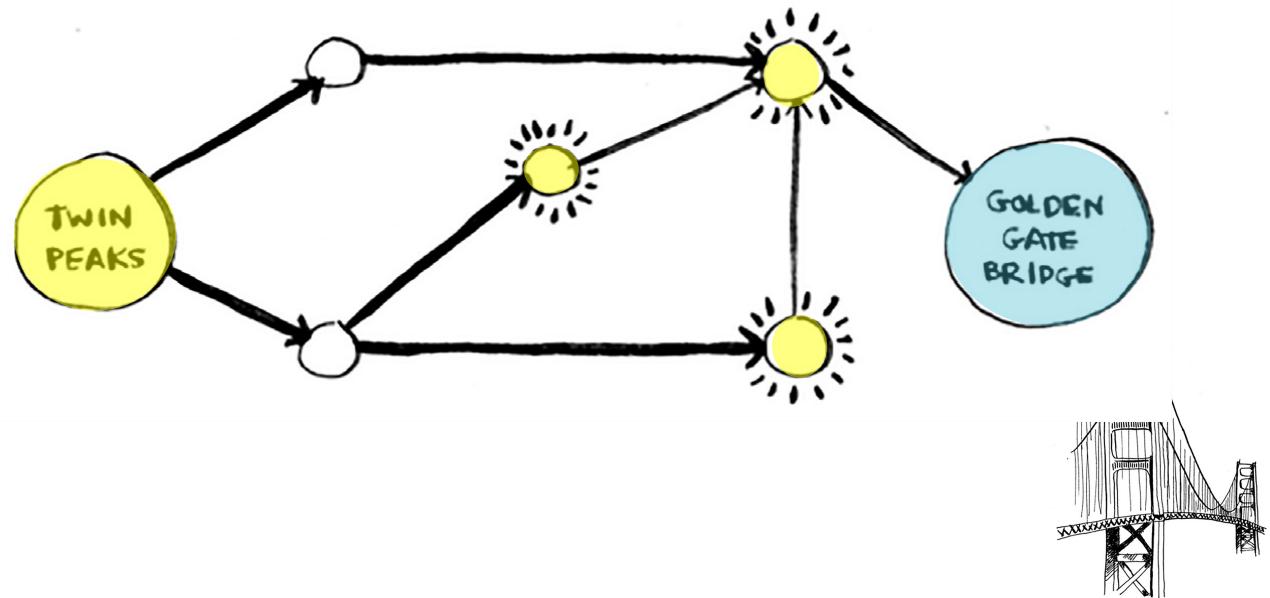
# Graphs

- ❖ Can you describe an algorithm to find the **path with the fewest steps?**
- ❖ Lets check: Start from Twin Peaks with one step.
- ❖ You **can't reach to bridge** in one step.



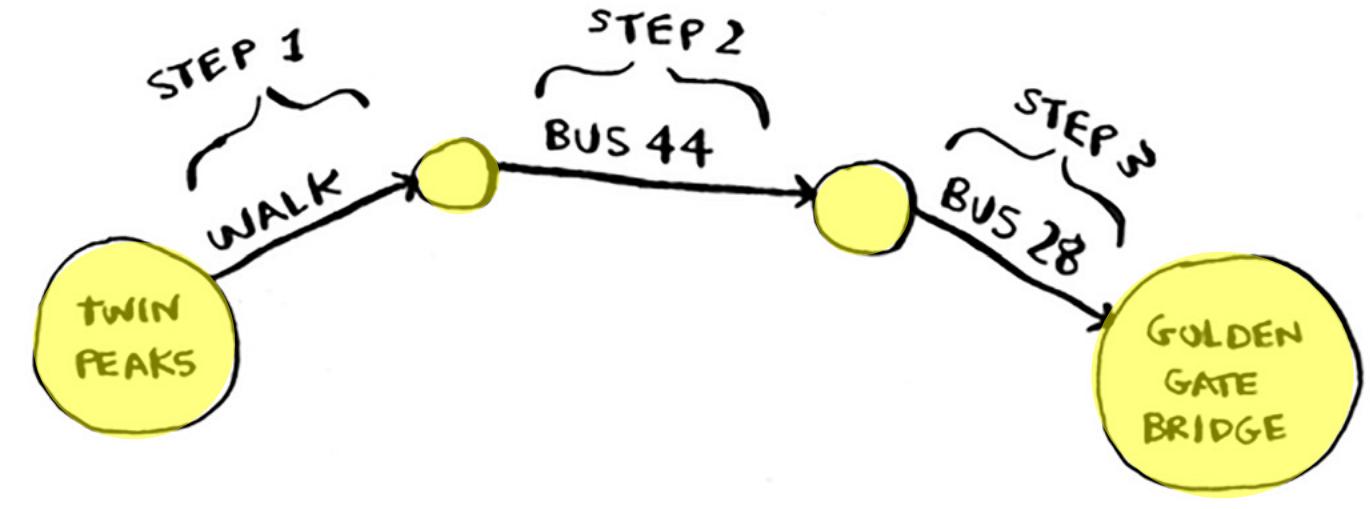
# Graphs

- ❖ What about **two steps**? No, you can't.
- ❖ What about **three steps**?



# Graphs

- ❖ Now the **Golden Gate Bridge** shows up.
- ❖ So it takes minimum **three steps** to get from Twin Peaks to the bridge, with this rout.



# Graphs

- ❖ There are other routes that will get you to the bridge too, but **they're longer** (four steps).
- ❖ This problem is known as **Shortest-path Problem**:
  - ❖ the shortest route to your friend's house.
  - ❖ the smallest number of moves to checkmate in a game of chess.
- ❖ Algorithms that can find the path in a Graph:
  - ❖ **Breadth-First Search (BFS)**
  - ❖ **Depth-First Search (DFS)**

# Graphs

- ❖ We address the **shortest path problem** by:
  - ❖ **modelling** the problem as a **Graph**.
  - ❖ **solving** it by using algorithms such as **BFS** or **DFS**.

- ❖ Q: But what is a **Graph**?
- ❖ A: It is a structure that models a set of points (nodes or vertices) and their connections (edges).

# Graphs

## ❖ Examples of Graphs.

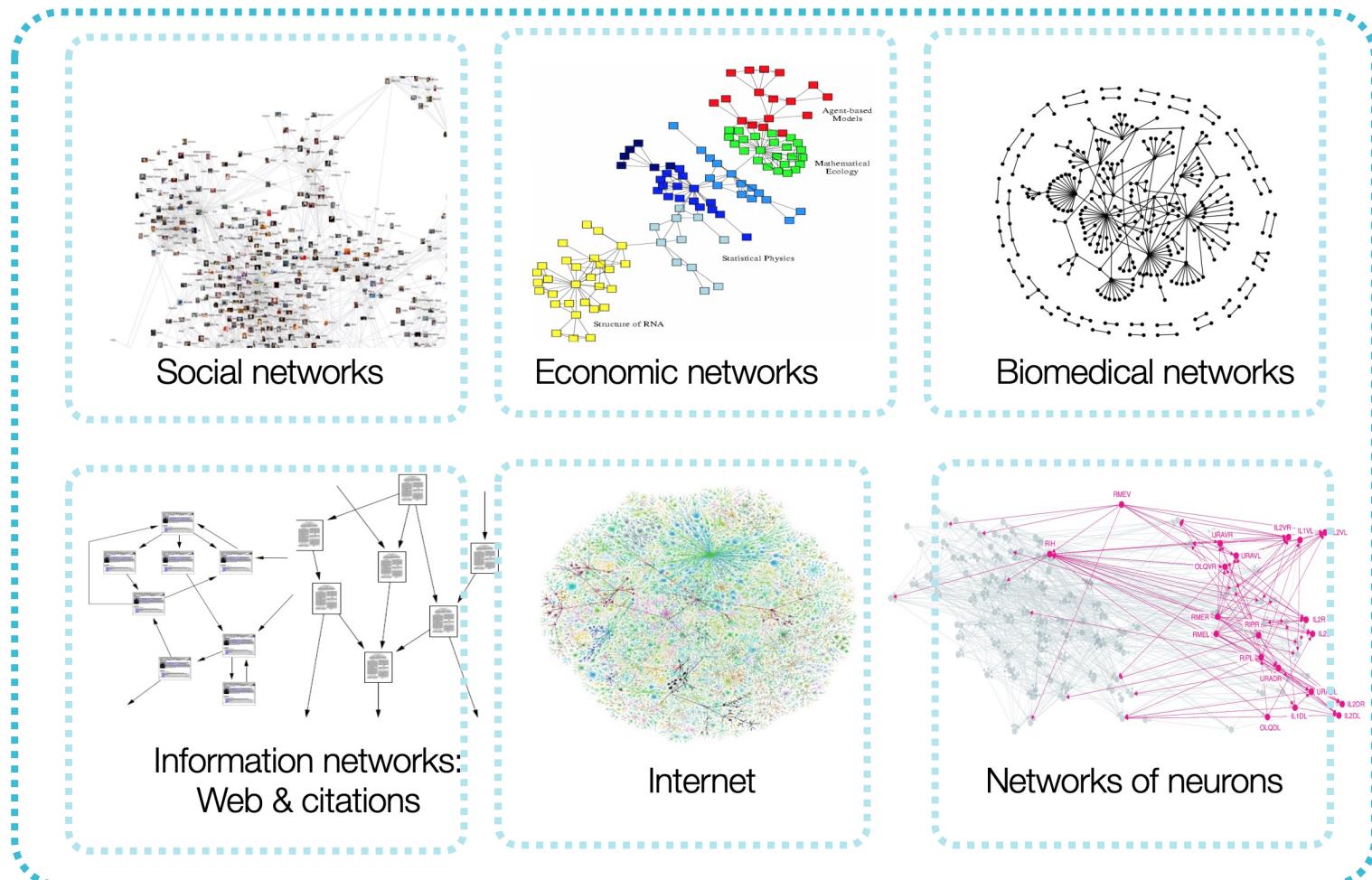
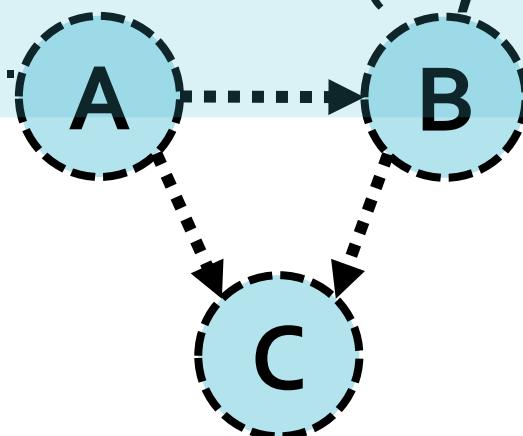


Image is taken from William Hamilton [COMP551: Graph Representation Learning](#)

# Graphs

- ❖ A Graph is built by two sets, i.e.:
  - Vertices (V) set also called nodes or points
  - Edges (E) set also called links, connections, or arcs

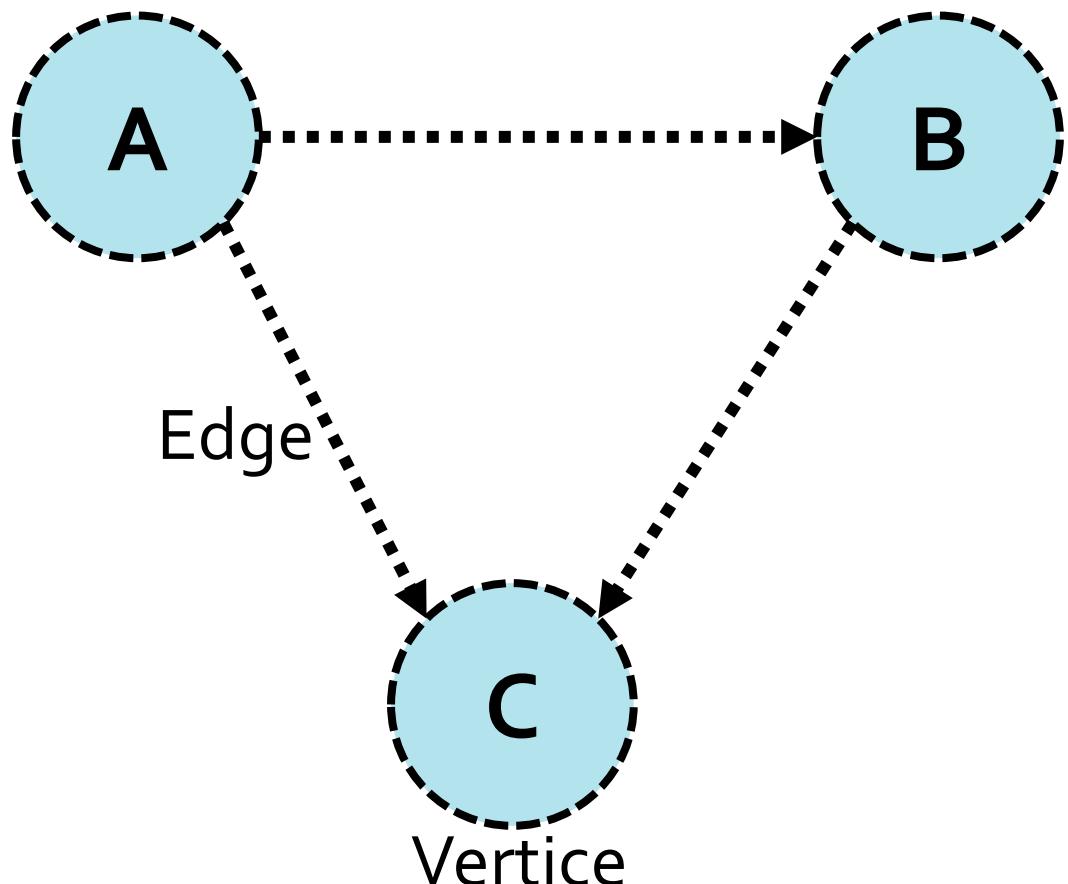
- ❖ Graphs are often shown:
  - ❖ Vertices set as boxes (or circles)
  - ❖ Edge set as lines or arcs between the boxes (or circles).
- ❖ There is an edge between  $v_1$  and  $v_2$  if  $(v_1, v_2)$  is an element of the Edge set.



# Graphs

## ❖ Example:

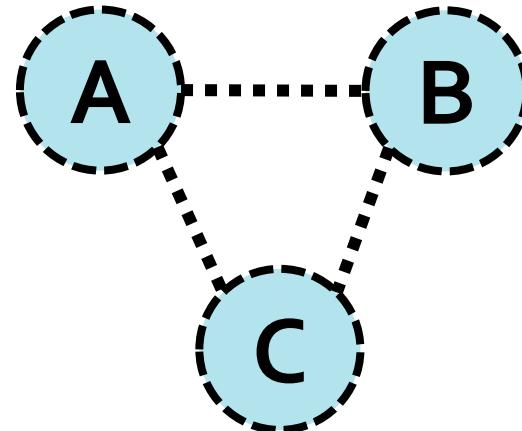
- The Vertices (nodes) set = {A,B,C}
- The Edges (links) set = {(A,B),(B,C),(A,C)}



# Graphs

## ❖ Undirected Graphs

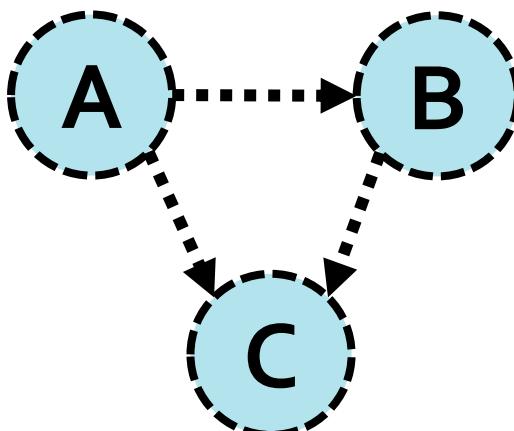
- is a graph in which **edges** does **not have orientations**.
- **order** of the vertices in the pairs in the edge set **does no matter**.
- We draw a straight line between the vertices.
- Adjacency relation is **symmetric** in undirected graph.



# Graphs

## ❖ Directed Graphs.

- is a graph in which **edges have orientations**.
- **order** of the vertices in the pairs in the edge set **matters**.
- We use arrows for the arcs between vertices.
- Adjacency relation is **asymmetric** in directed graph.



# Graphs

## ❖ How to represent **Graphs**?

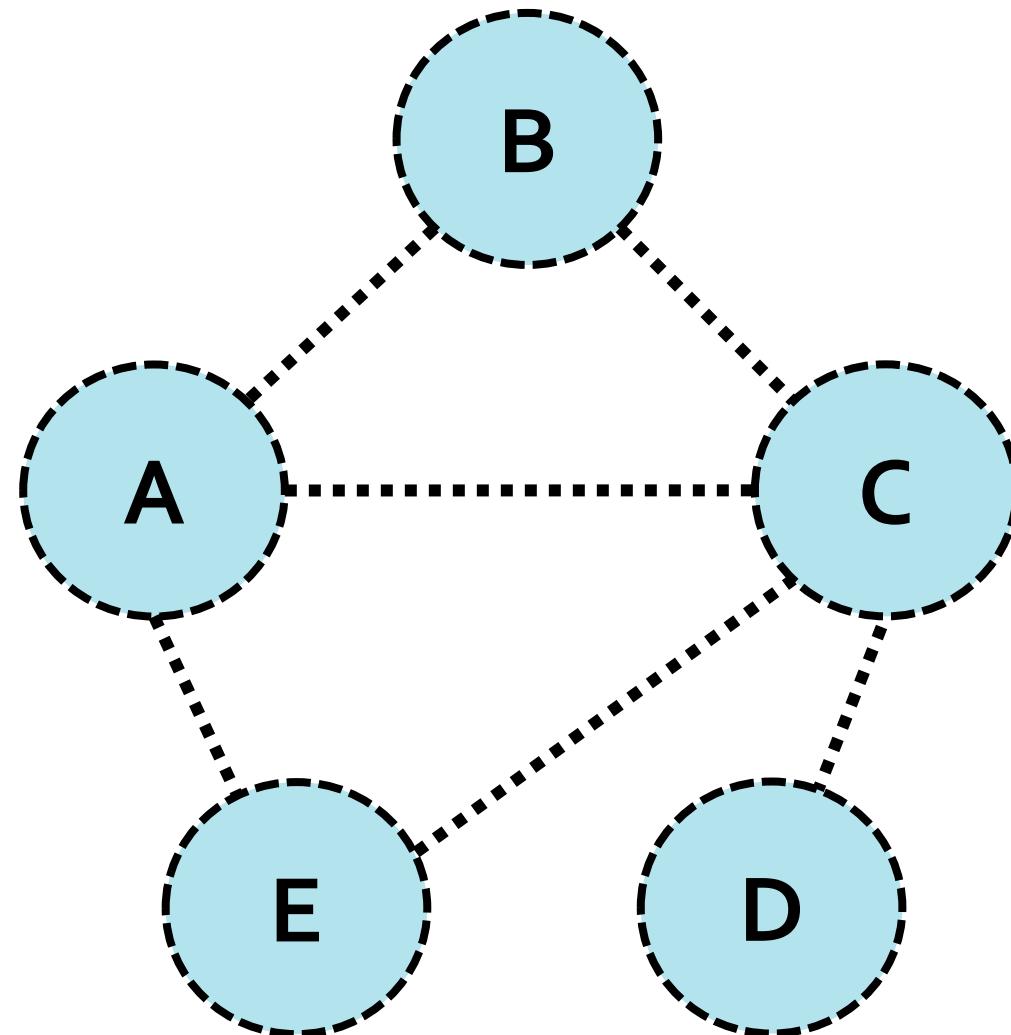
❖ There can be different methods. Two methods :

- List (called **Adjacency List**)
- Matrix (called **Adjacency Matrix**)

❖ Lets see examples of both.

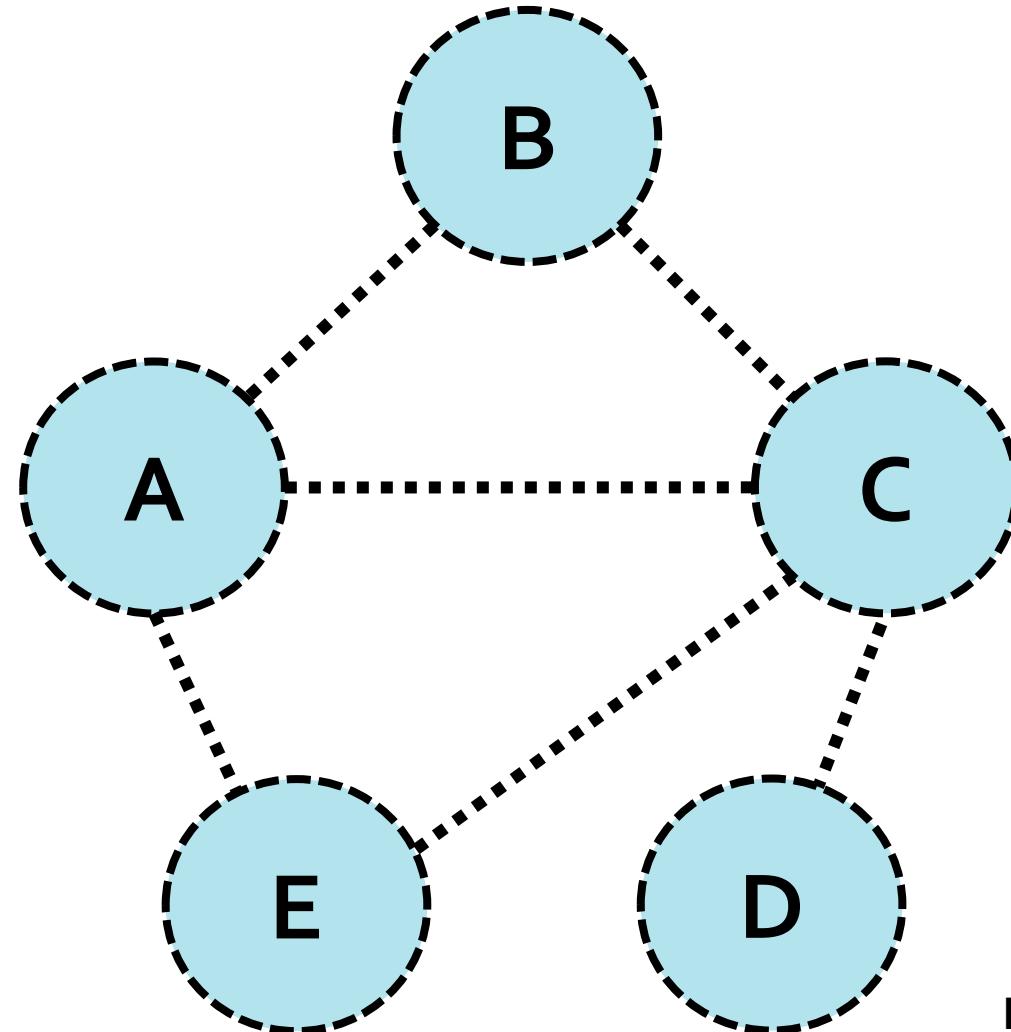
❖ First consider **undirected** Graphs.

# Undirected Graphs



Adjacency List

# Undirected Graphs



## Adjacency List

A: B, C, E

B: A, C

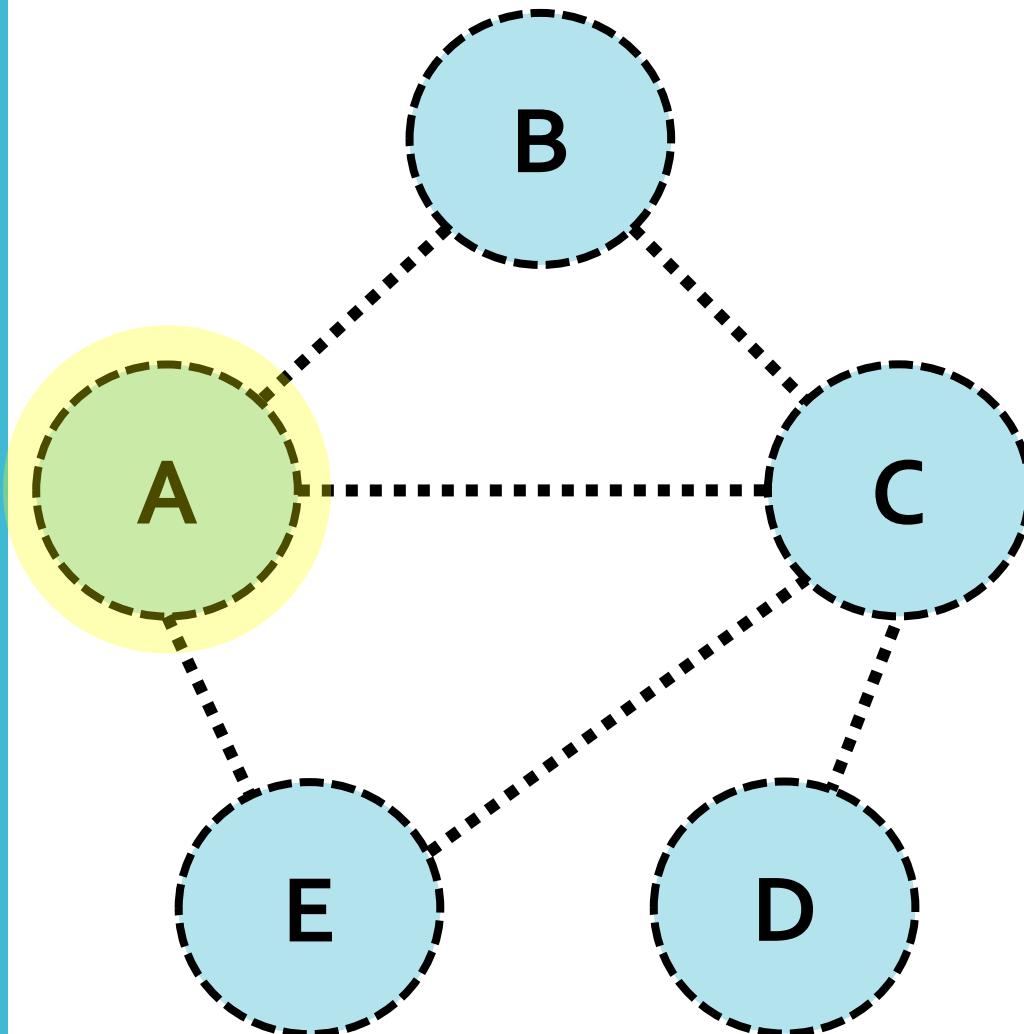
C: A, B, D, E

D: C

E: A, C

**Note:** adjacency list uses  $O(|E|)$  space

# Undirected Graphs



## Adjacency List

A: B, C, E

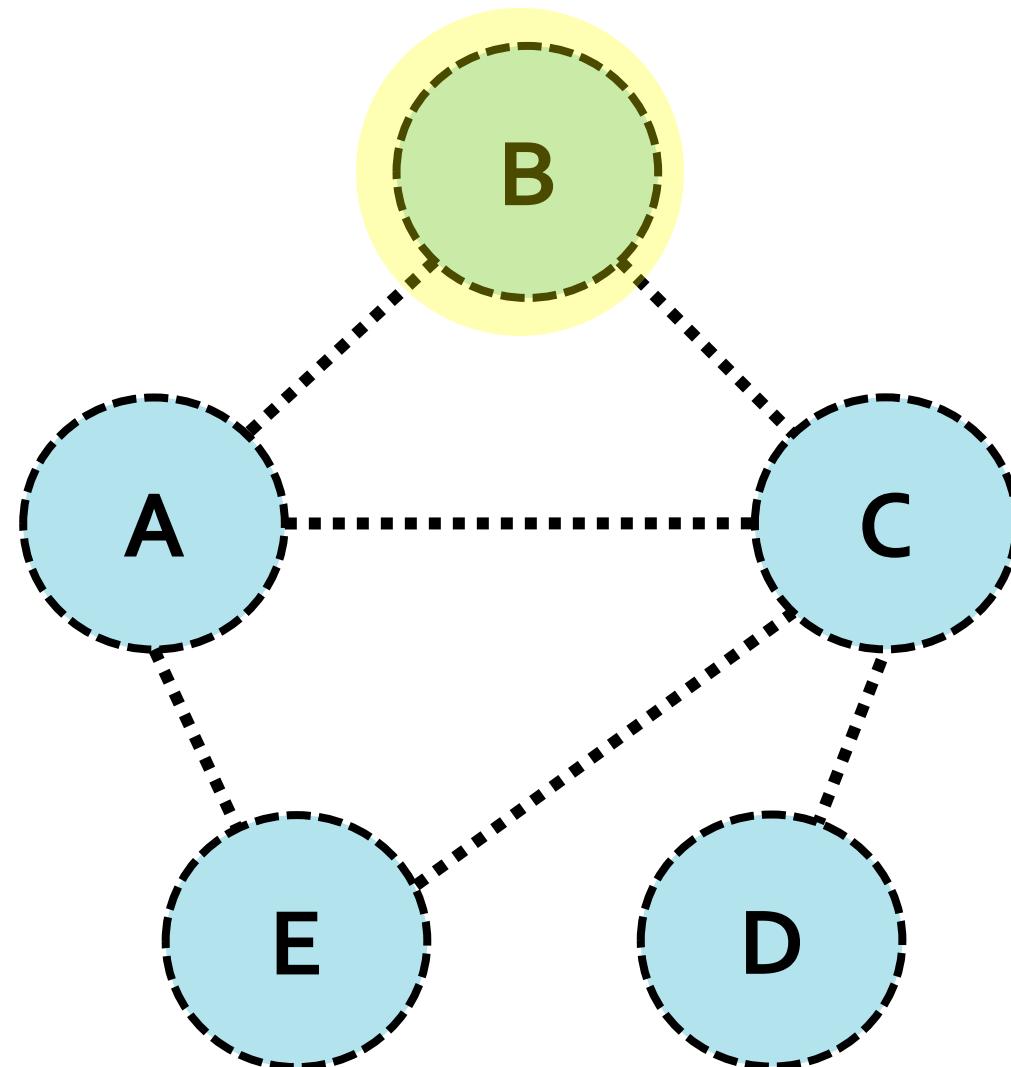
B: A, C

C: A, B, D, E

D: C

E: A, C

# Undirected Graphs



## Adjacency List

A: B, C, E

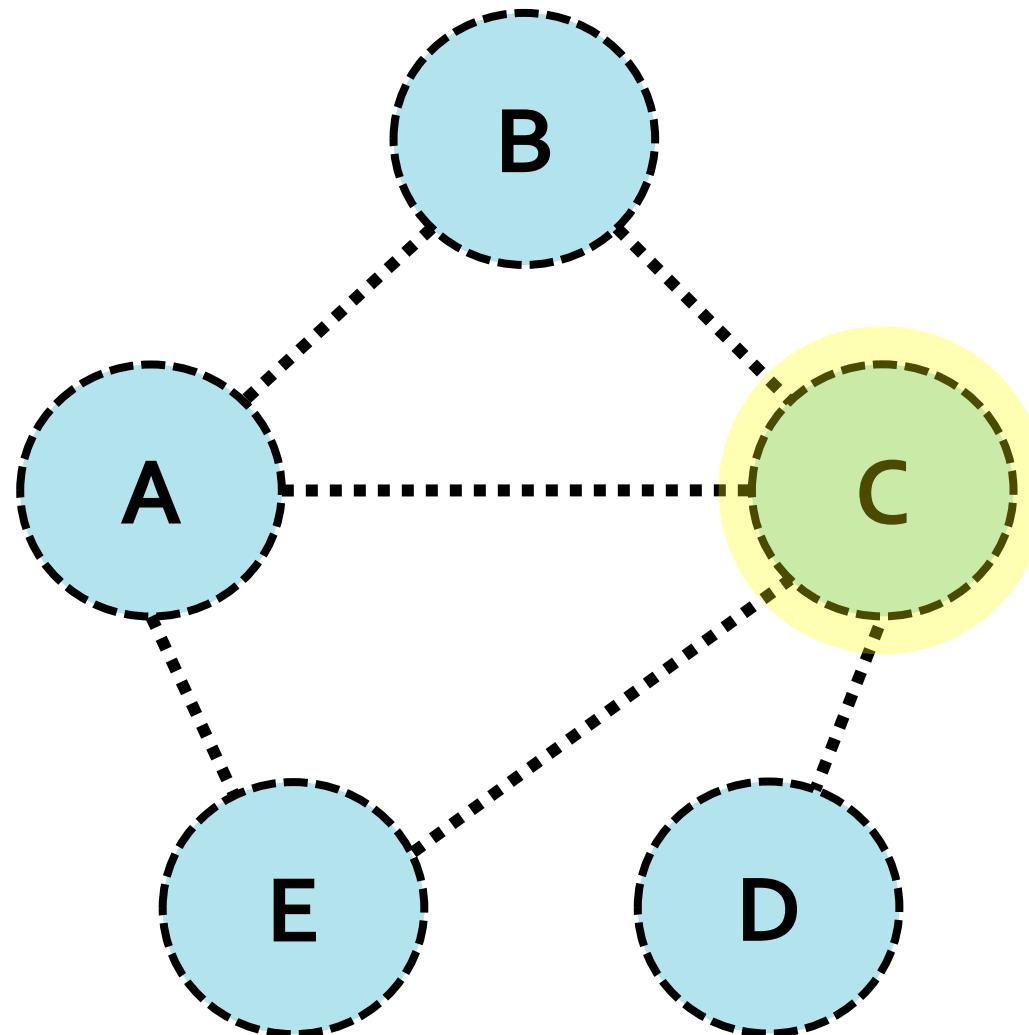
B: A, C

C: A, B, D, E

D: C

E: A, C

# Undirected Graphs



## Adjacency List

A: B, C, E

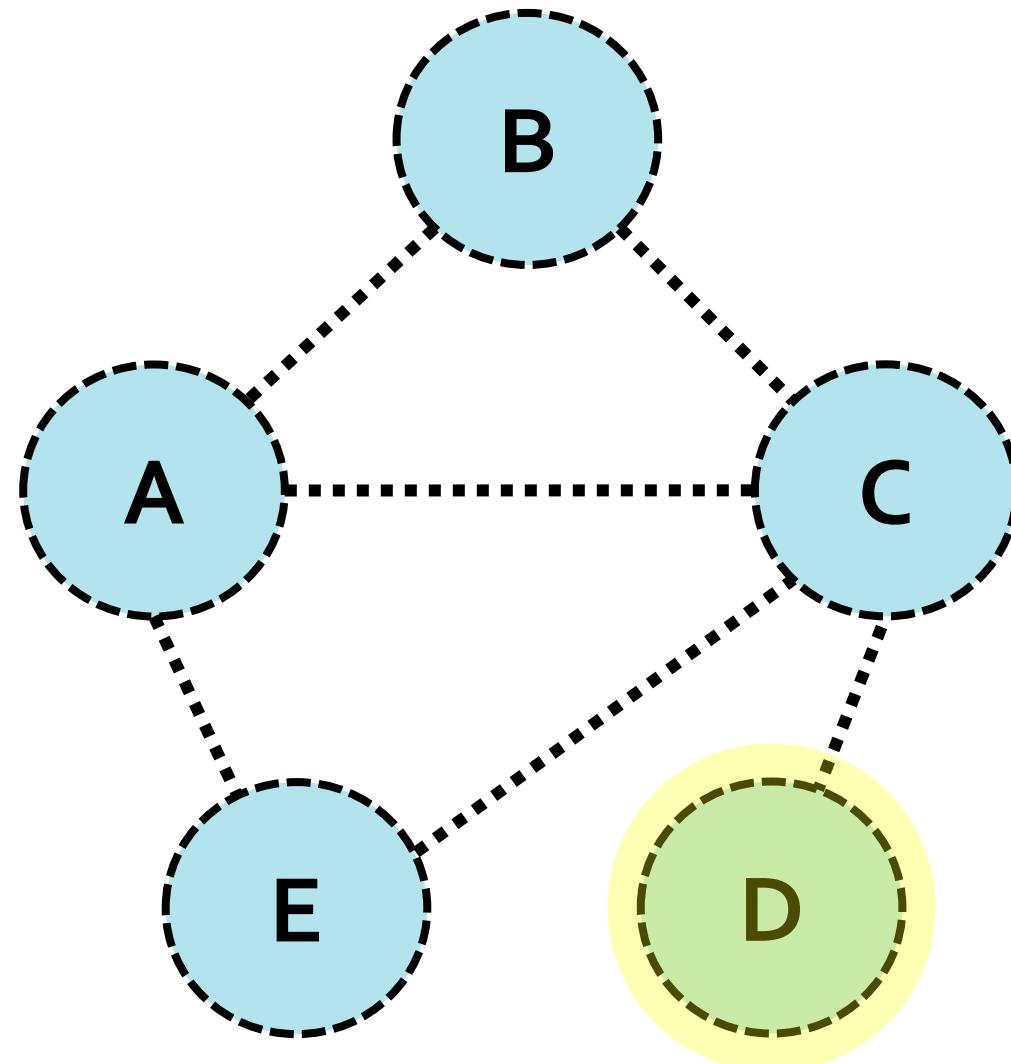
B: A, C

C: A, B, D, E

D: C

E: A, C

# Undirected Graphs



## Adjacency List

A: B, C, E

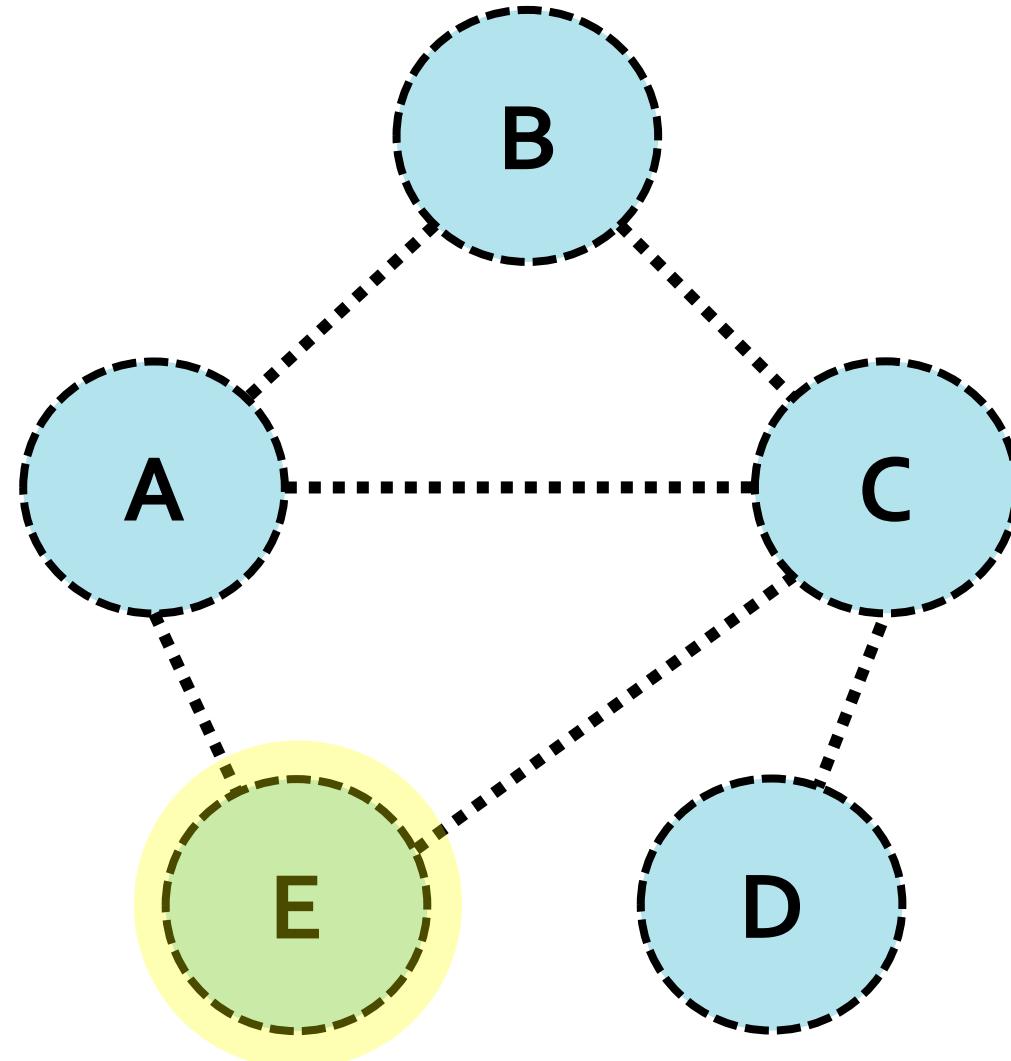
B: A, C

C: A, B, D, E

D: C

E: A, C

# Undirected Graphs



## Adjacency List

A: B, C, E

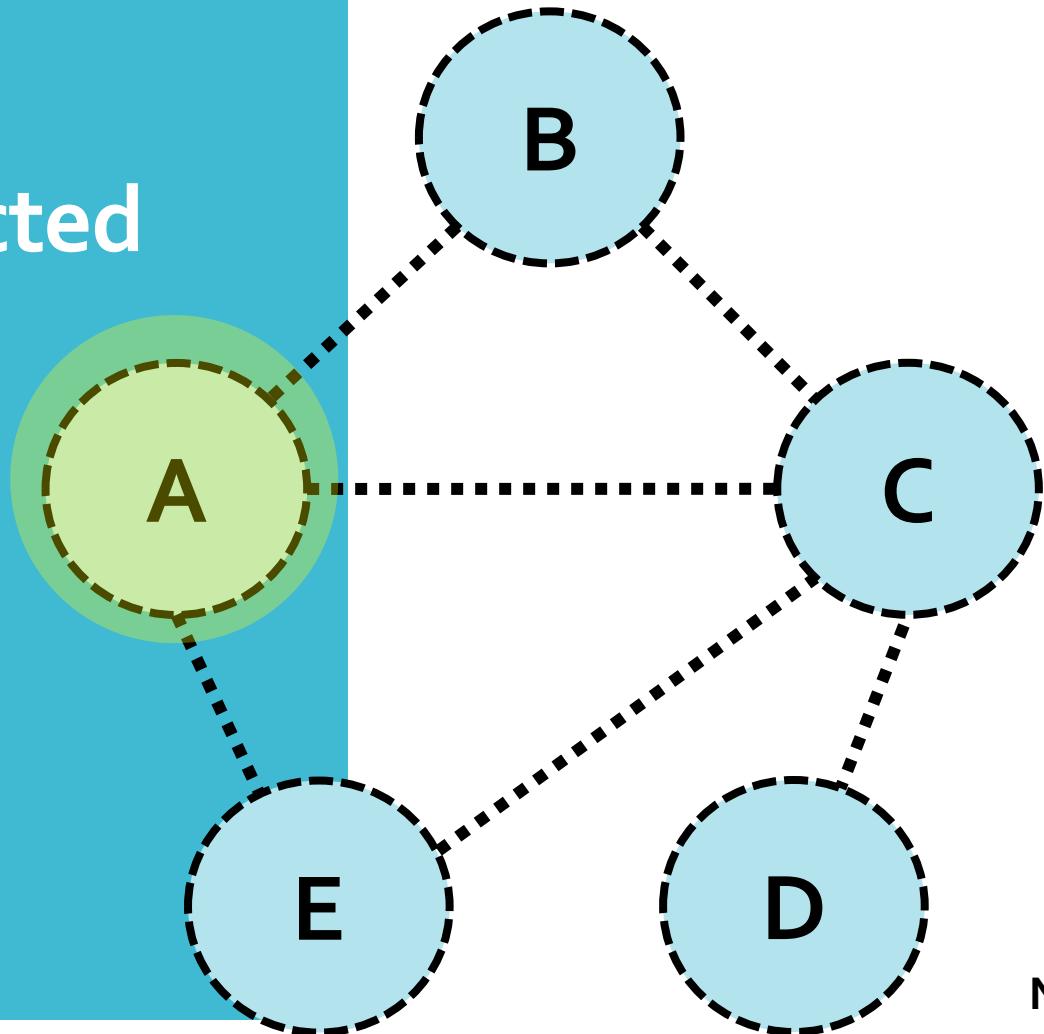
B: A, C

C: A, B, D, E

D: C

E: A, C

# Undirected Graphs



# Adjacency Matrix

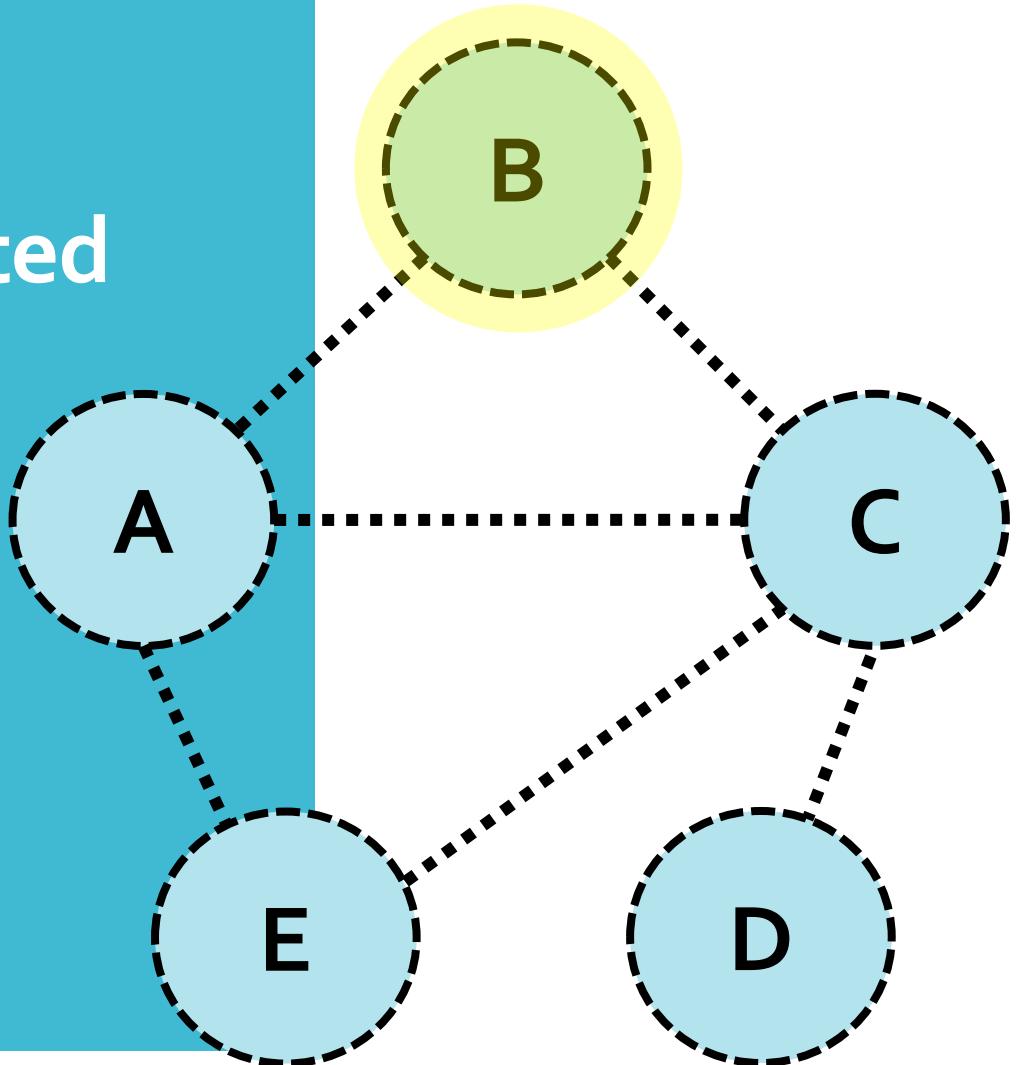
**from**

to

	A	B	C	D	E
A	0	1	1	0	1
B	1	0	1	0	0
C	1	1	0	1	1
D	0	0	1	0	0
E	1	0	1	0	0

**Note:** matrix can be implemented as a 2D Python List with  $|V| \times |V|$  cells

## Undirected Graphs

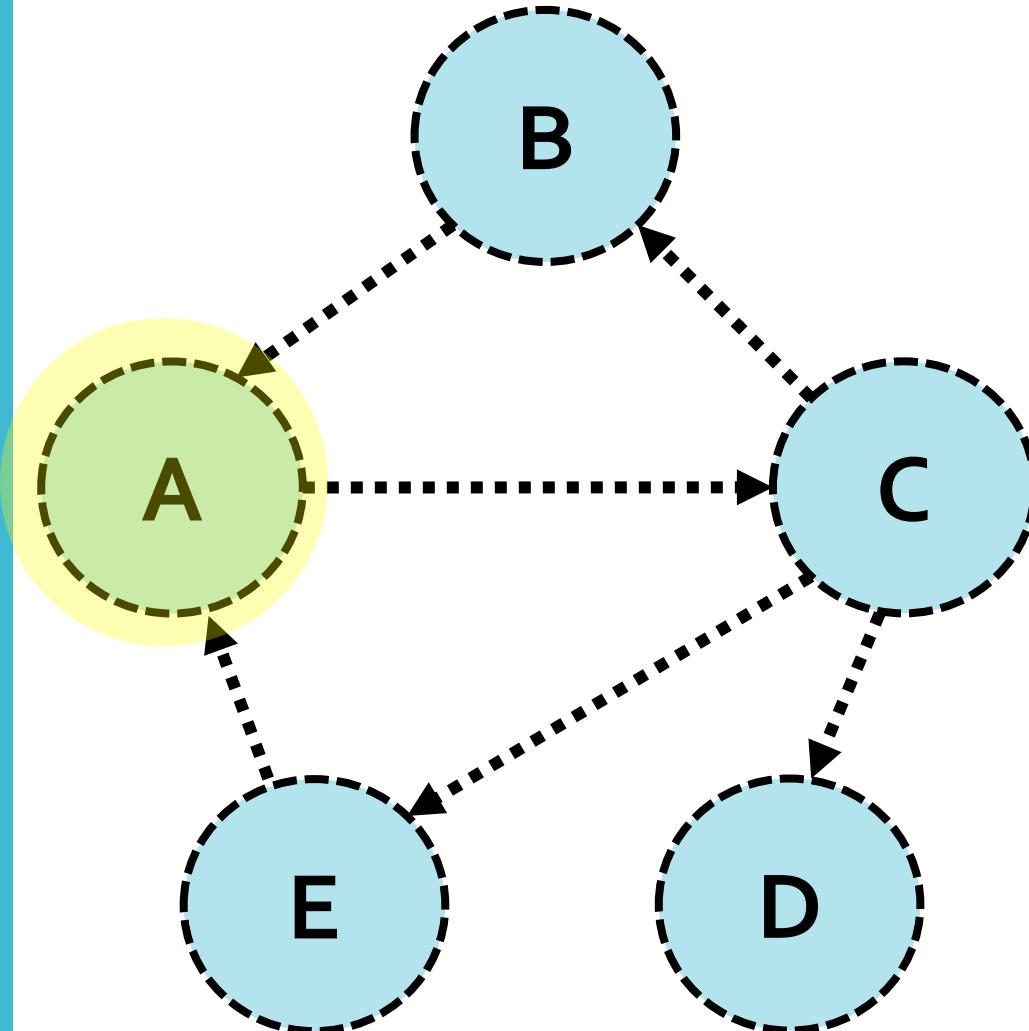


## Adjacency Matrix

from                      to

	A	B	C	D	E
A	0	1	1	0	1
B	1	0	1	0	0
C	1	1	0	1	1
D	0	0	1	0	0
E	1	0	1	0	0

# Directed Graphs



## Adjacency List

A: C

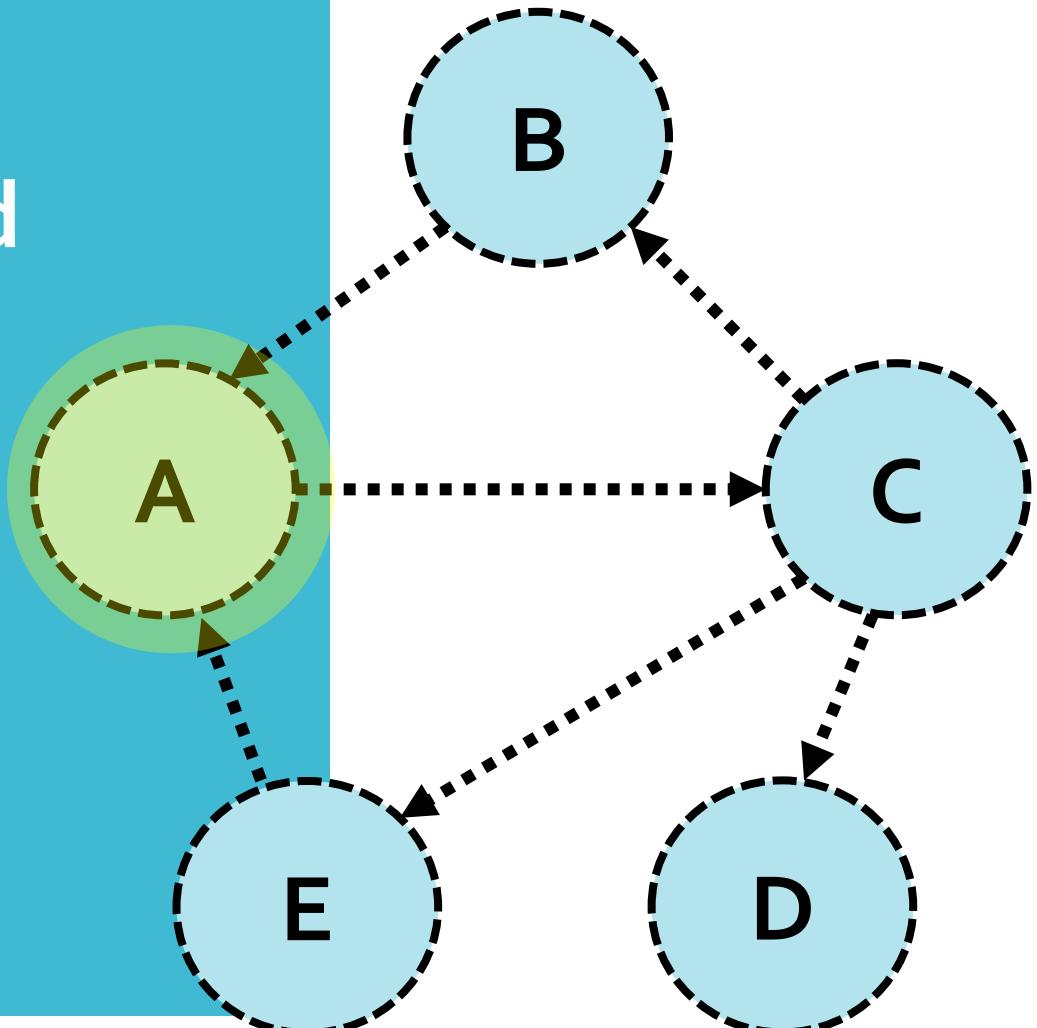
B: A

C: B, D, E

D:

E: A

## Directed Graphs

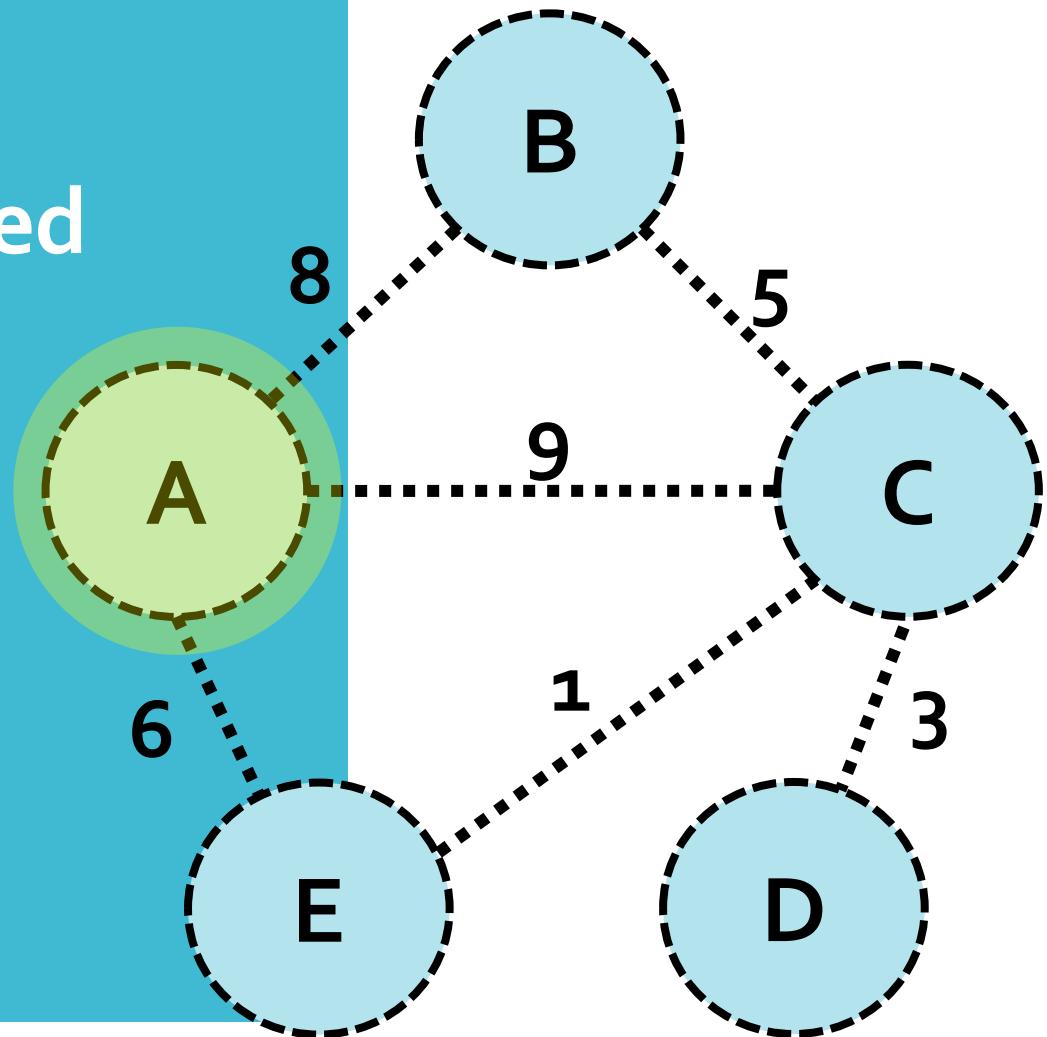


## Adjacency Matrix

from                      to

	A	B	C	D	E
A	0	0	1	0	0
B	1	0	0	0	0
C	0	1	0	1	1
D	0	0	0	0	0
E	1	0	0	0	0

## Weighted Graphs



These can be used to represent  
also other types of Graphs

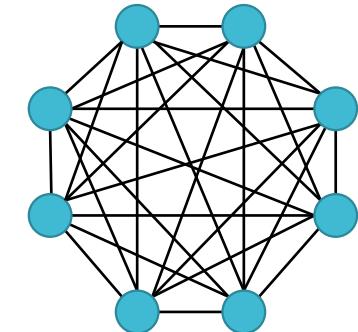
to  
from

	A	B	C	D	E
A	0	8	9	0	6
B	8	0	5	0	0
C	9	5	0	3	1
D	0	0	3	0	0
E	6	0	1	0	0

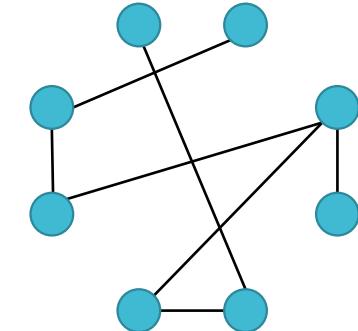
# Quiz

❖ Consider the **following graphs**. Which Graph representation better suits to each of them?

- a) **Dense Graph:**  
graph where  $|E| = |V|^2$



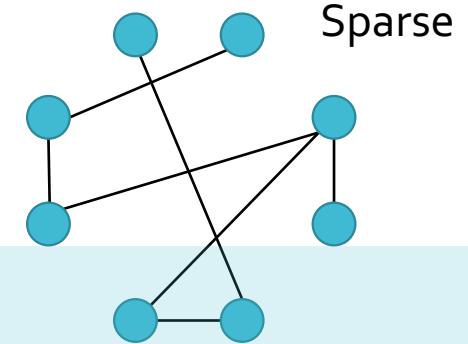
- b) **Sparse Graph:**  
graph where  $|E| = |V|$



# Answer

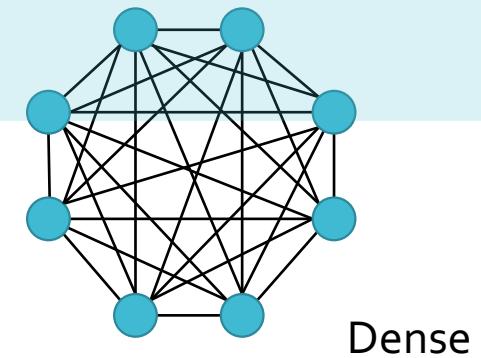
## ❖ Adjacency List

- **Pro:** Faster (& uses less space) for Sparse graphs.
- **Con:** Slower for Dense graphs



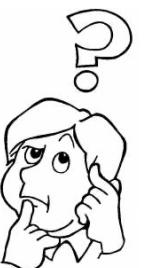
## ❖ Adjacency Matrix

- **Pro:** Faster for Dense graphs.
- **Con:** Uses more space.



## Answer

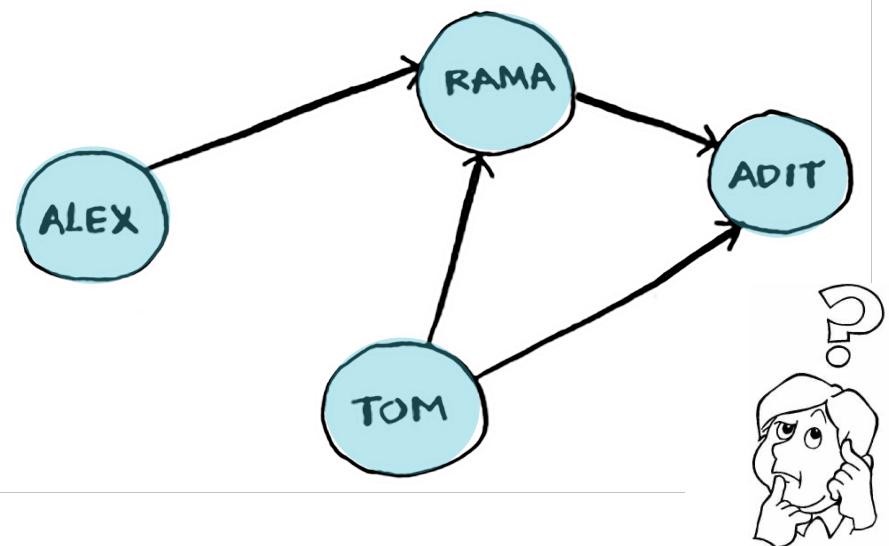
- ❖ **Adjacency Matrix** takes up  $|V|^2$  space, regardless of how dense the graph is.
- ❖ Matrix for a graph with **10,000 vertices** will take up at least **100,000,000 Bytes!**



## Quiz

- ❖ Suppose the following Graph is given representing **who owes money to whom**.
- ❖ Implement a **Graph class** with Python **dictionary** and create the graph
- ❖ The class should have at least the following methods:

- `add_edge()`
- `print_graph()`
- `Print_edges()`

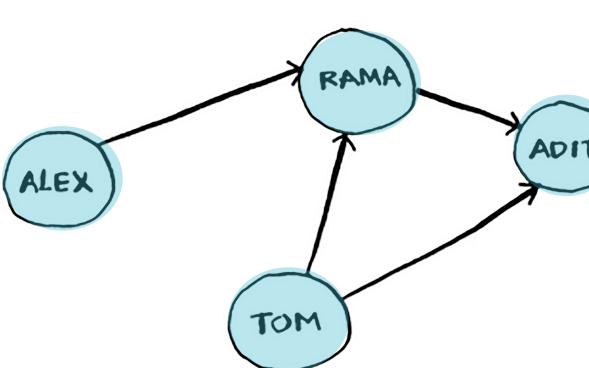




❖ This is an implement of a **Graph** class.

```
class Graph:  
    graph = dict()  
  
    def add_edge(self, node, neighbour):  
        if node not in self.graph:  
            self.graph[node] = [neighbour]  
        else:  
            self.graph[node].append(neighbour)  
  
    def print_graph(self):  
        print(self.graph)
```

Answer

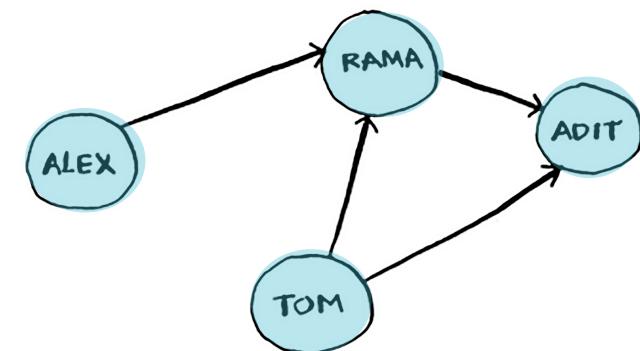


❖ This is an implement of a **Graph** class.



Answer

```
my_graph = Graph()  
  
my_graph.add_edge('Alex', 'Rama')  
my_graph.add_edge('Rama', 'Adit')  
my_graph.add_edge('Tom', 'Rama')  
my_graph.add_edge('Tom', 'Adit')  
  
my_graph.print_graph()
```



[Output:]

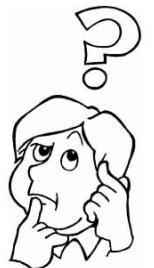
```
{'Alex': ['Rama'], 'Rama': ['Adit'], 'Tom': ['Rama', 'Adit']}
```

# Answer

```
def print_edges(self):  
    for node in self.graph:  
        for neighbour in self.graph[node]:  
            print("(", node, ", ", neighbour, ")")  
  
my_graph.print_edges()
```

## [Output:]

```
( Alex , Rama )  
( Rama , Adit )  
( Tom , Rama )  
( Tom , Adit )
```

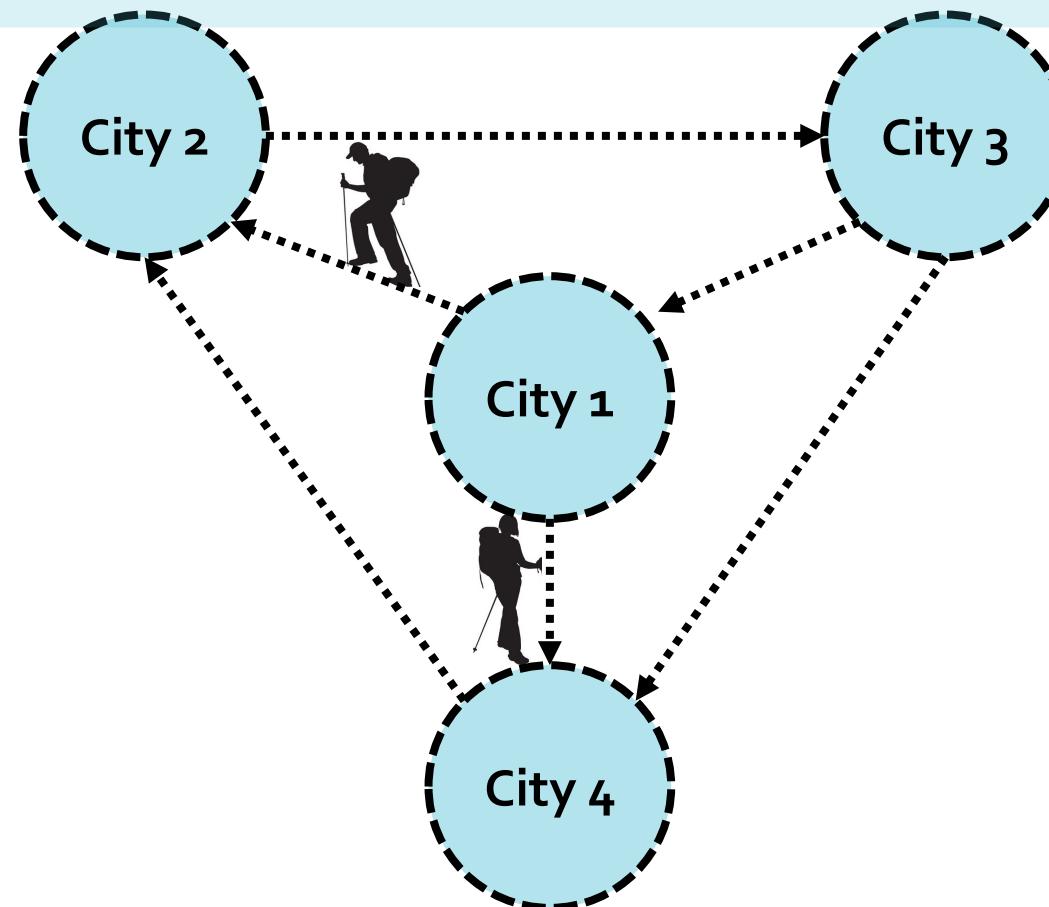


# Graph Traverse

- ❖ **Graph Traverse:** is a process for navigating through a graph by examining the vertices (nodes) and edges (links).
- ❖ Graph traversal algorithms are key for solving many **fundamental problems** that can be modelled by graphs.
- ❖ There are different traverse algorithms. Here are the two popular algorithms:
  - **Breadth-First Search (BFS)**
  - **Depth-First Search (DFS)**

# BFS

- ❖ **Breadth-First Search (BFS)** is a traverse algorithm that proceeds in rounds and subdivides the vertices into levels.
- ❖ **Breadth-First Search (BFS)** looks like an algorithm that sends out many explorers in all directions. They explore the graph in coordinated fashion.



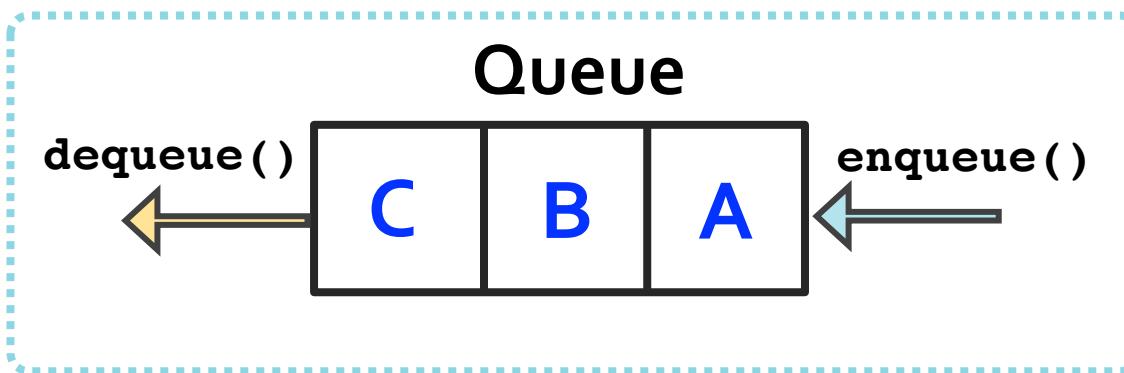
# BFS

- ❖ Breadth-First Search (BFS) algorithm:
  - ❖ Starts at a node (vertex): **Level 0 checked!**
  - ❖ In the 1<sup>st</sup> round, visits the nodes adjacent to the start node: **Level 1 checked!**
  - ❖ In the 2<sup>nd</sup> round, sends explorers two steps away from the starting node and not previously assigned to a level: **Level 3 checked!**
  - ❖ Continues this process until no new node are found. When a node is visited, it is basically printed out.

## Quiz

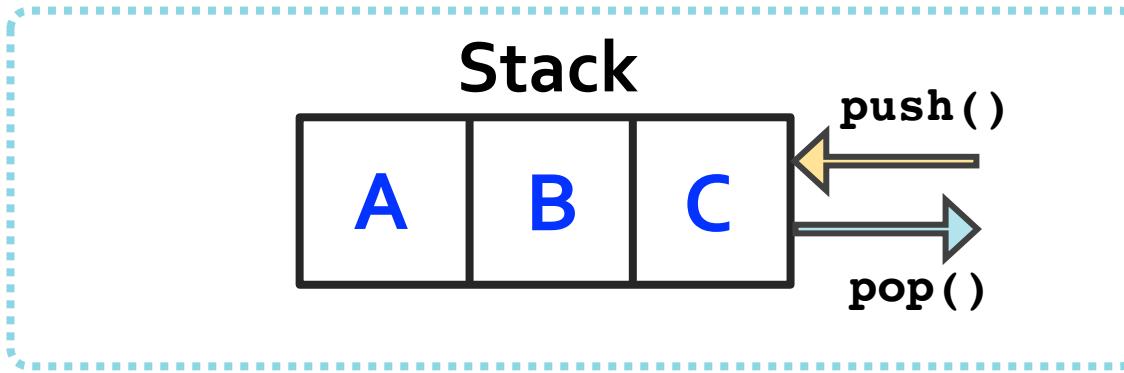
❖ Based on that idea, which **data structure** do you think can be used for implementing BFS?

(i)



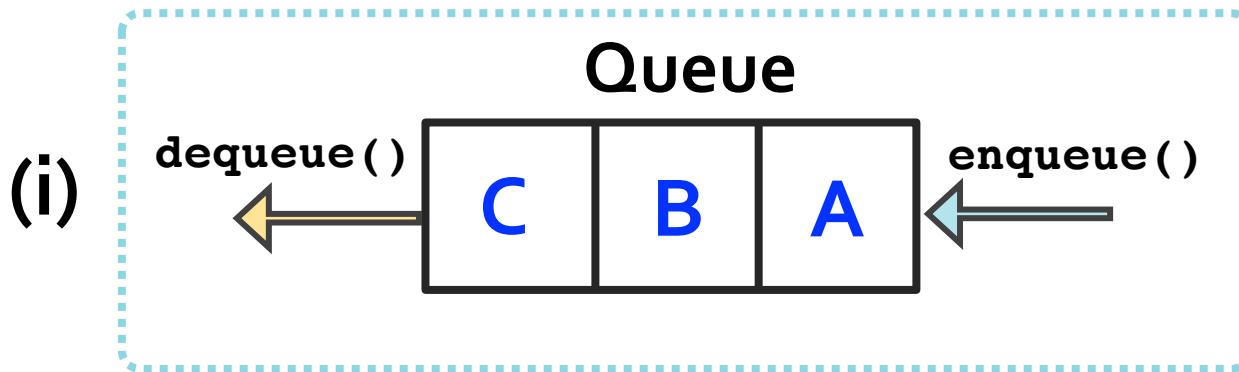
?

(ii)

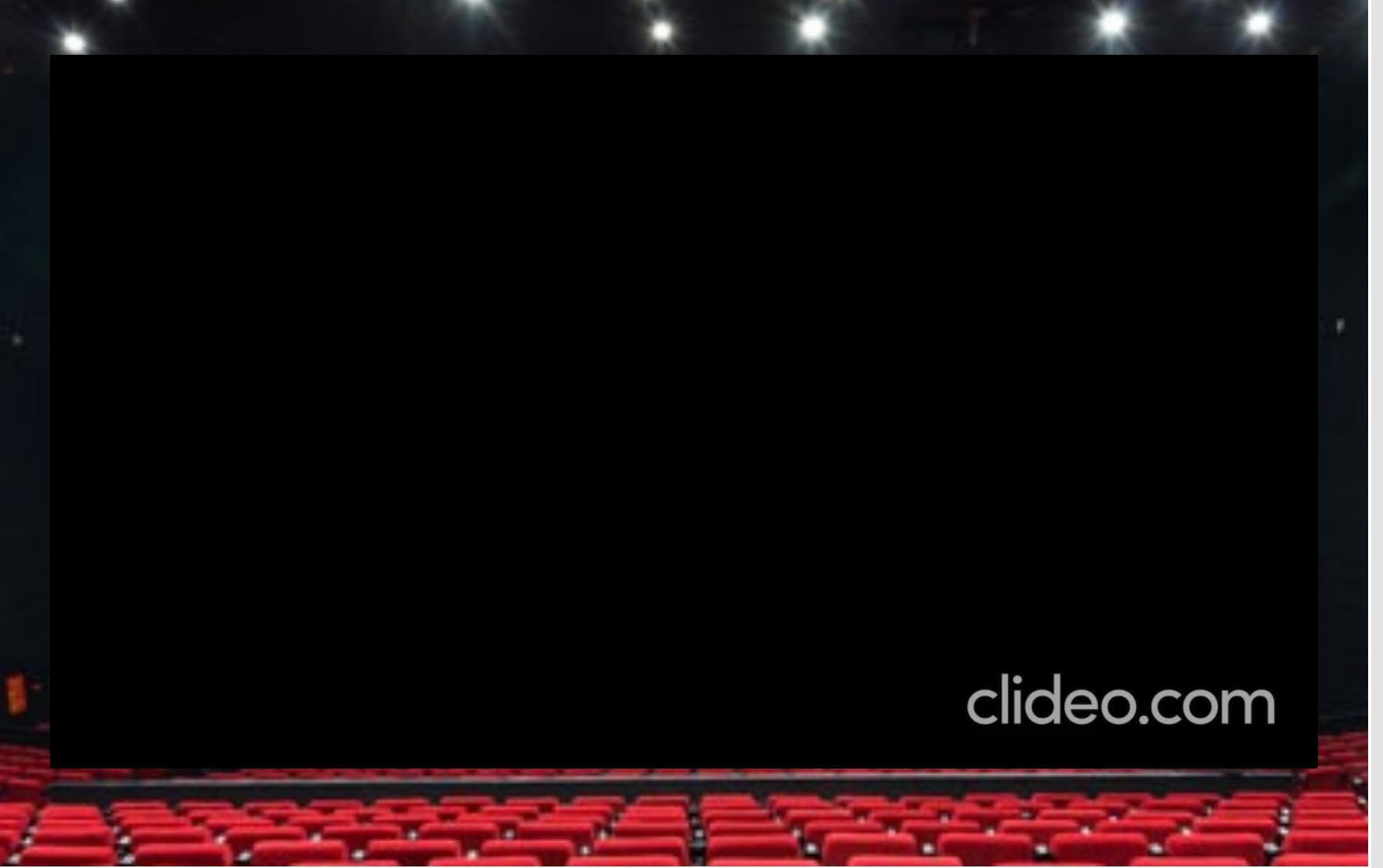


## Answer

❖ **Queue** can be used to implement **BFS**.



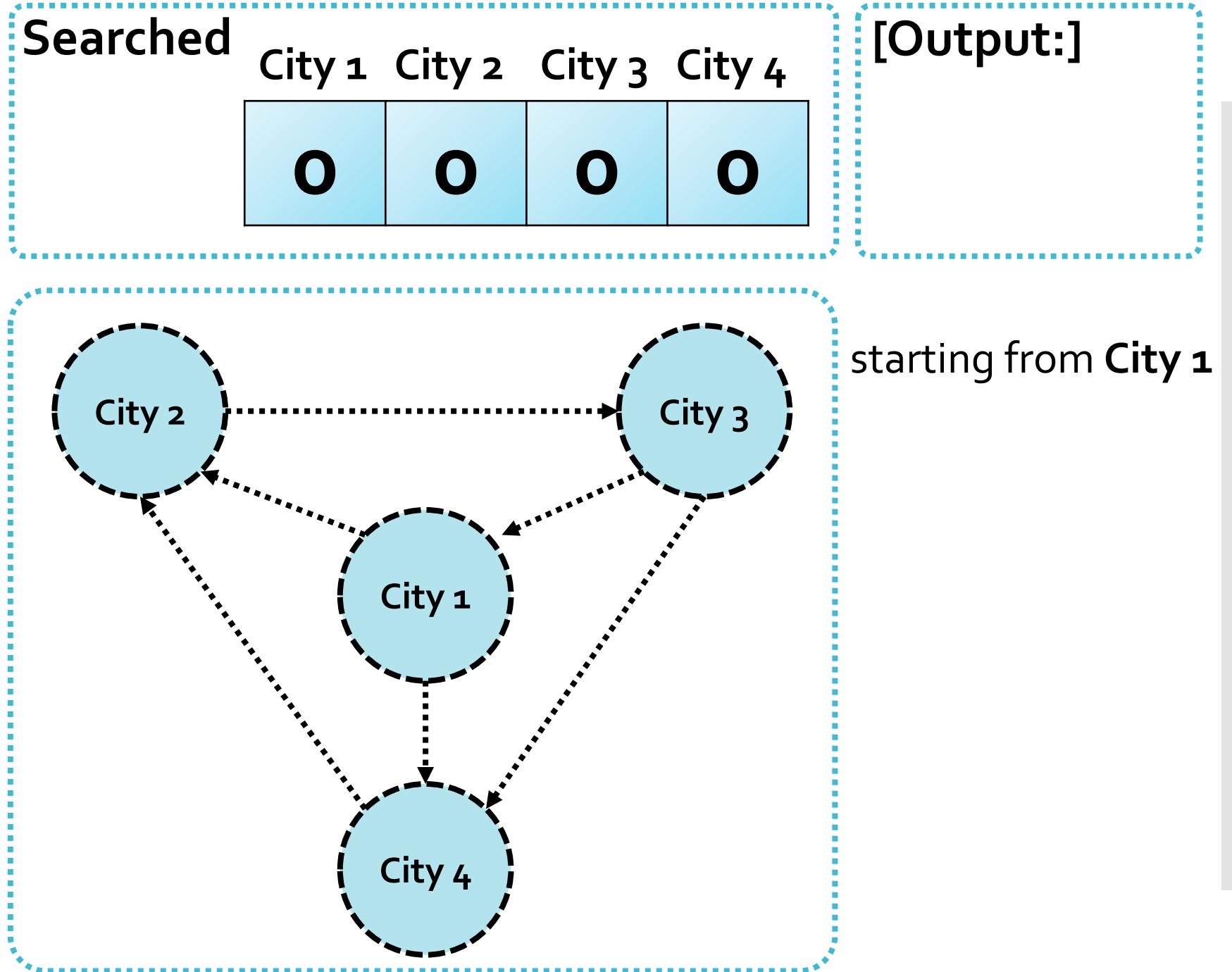
BFS



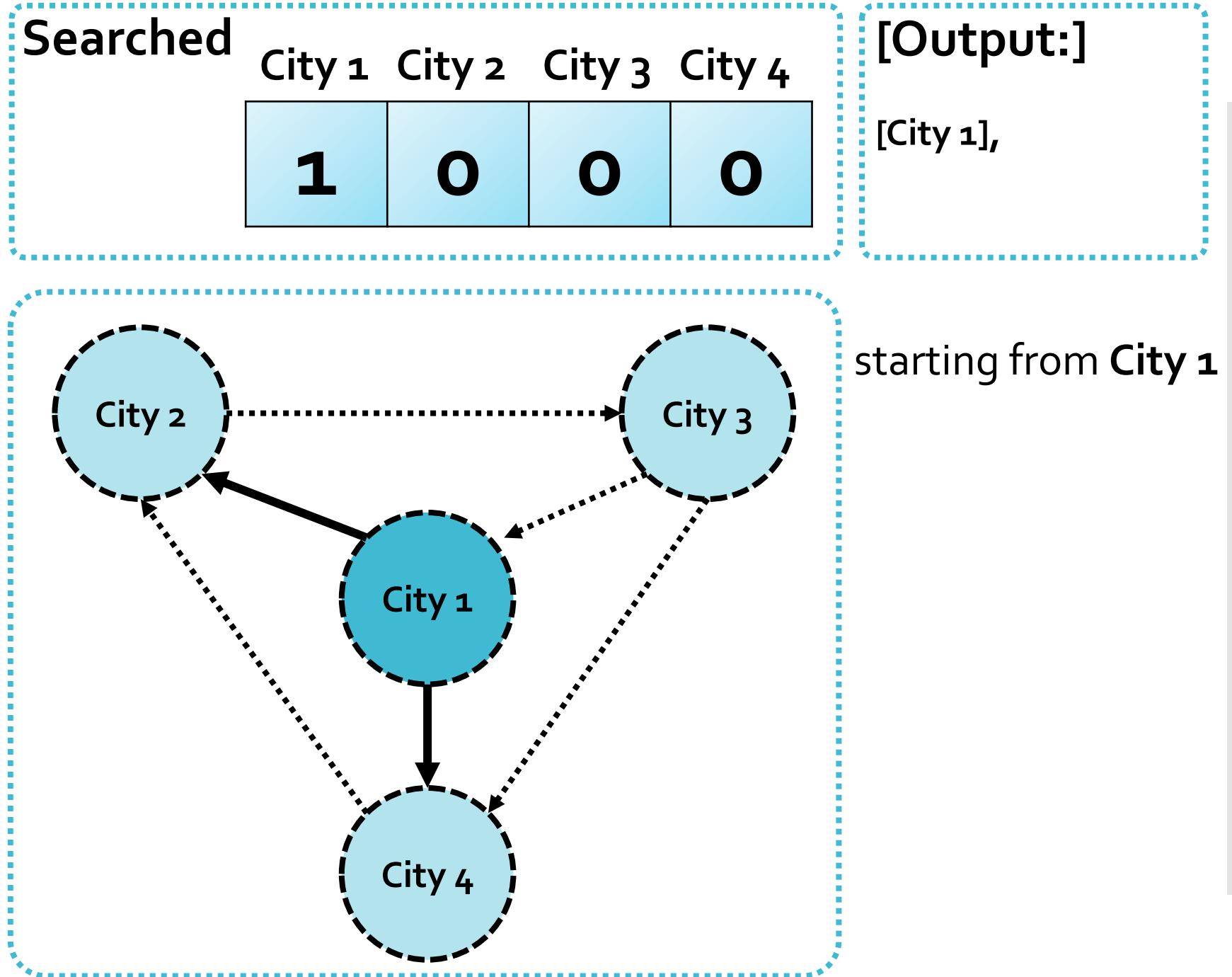
clideo.com

ref: [youtu.be/oDqjPvD54Ss](https://youtu.be/oDqjPvD54Ss)

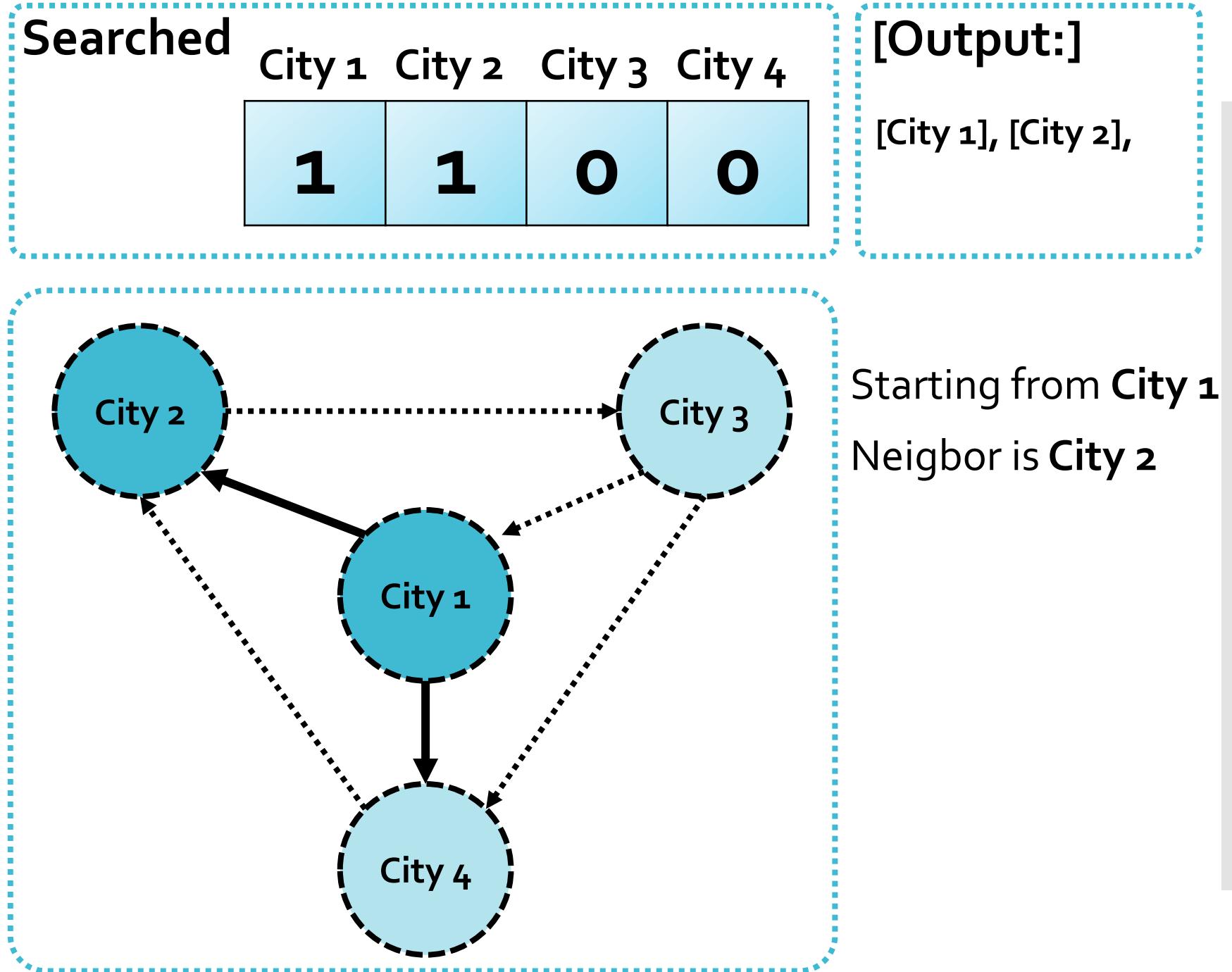
# Breadth First Search (BFS)



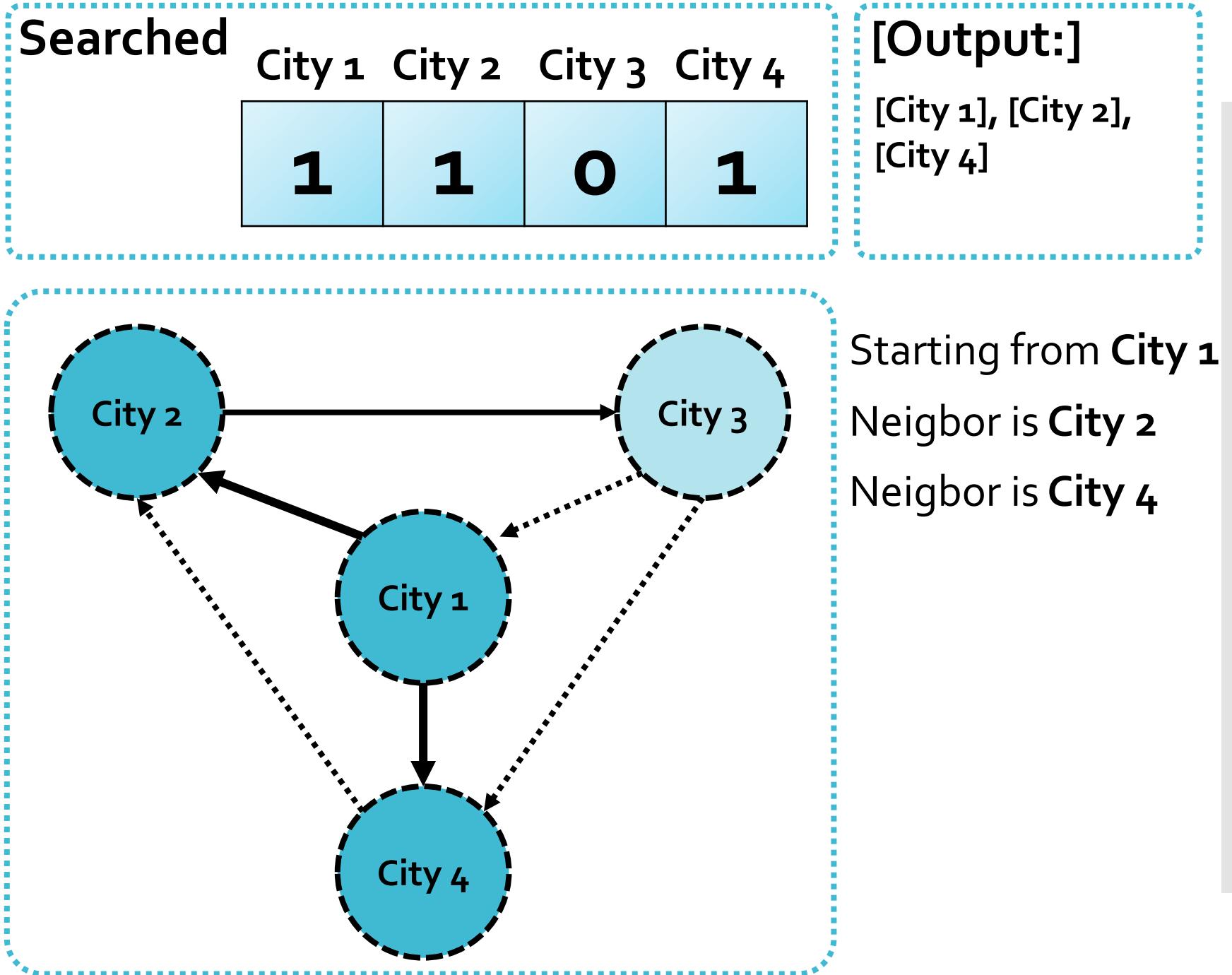
# Breadth First Search (BFS)



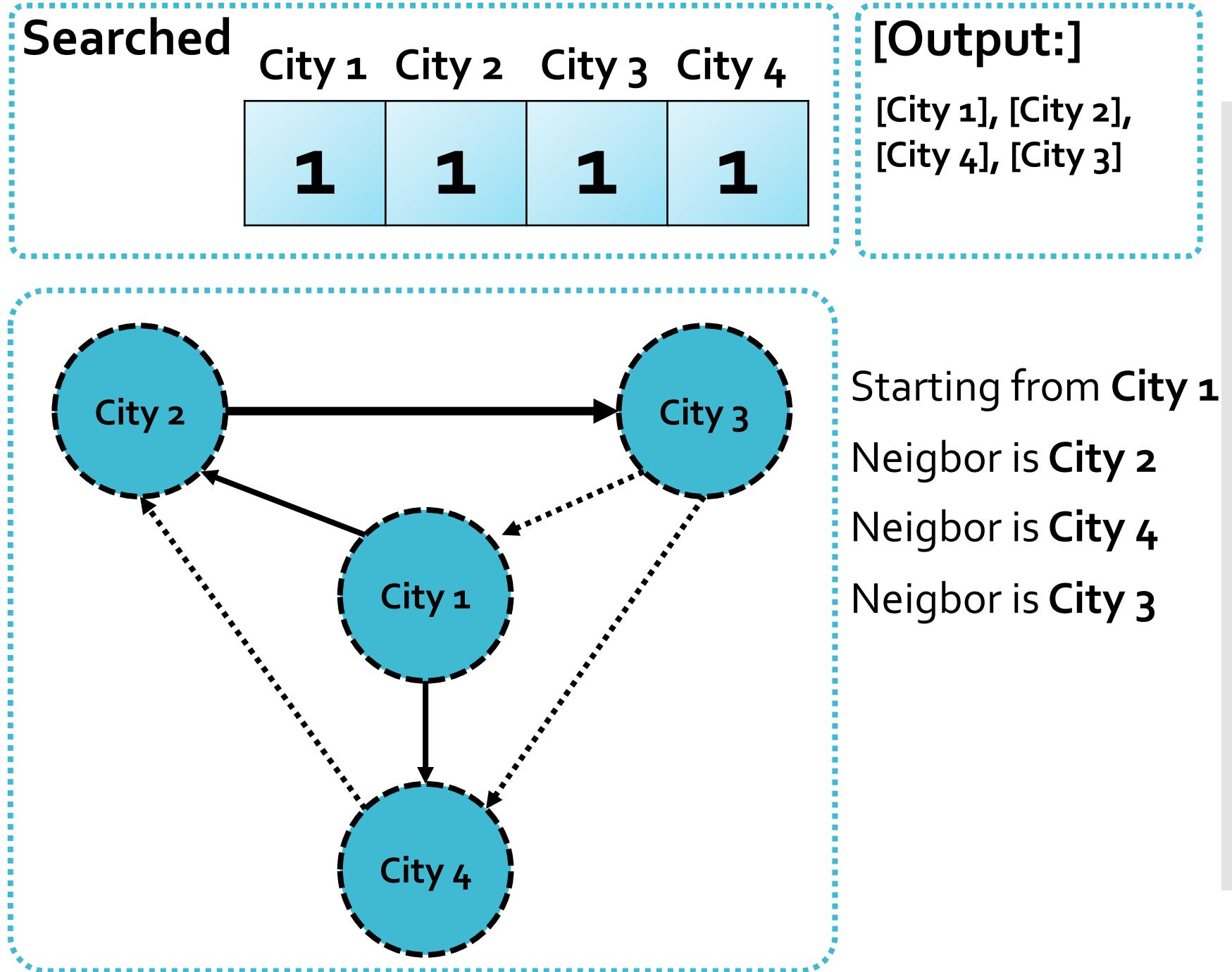
# Breadth First Search (BFS)



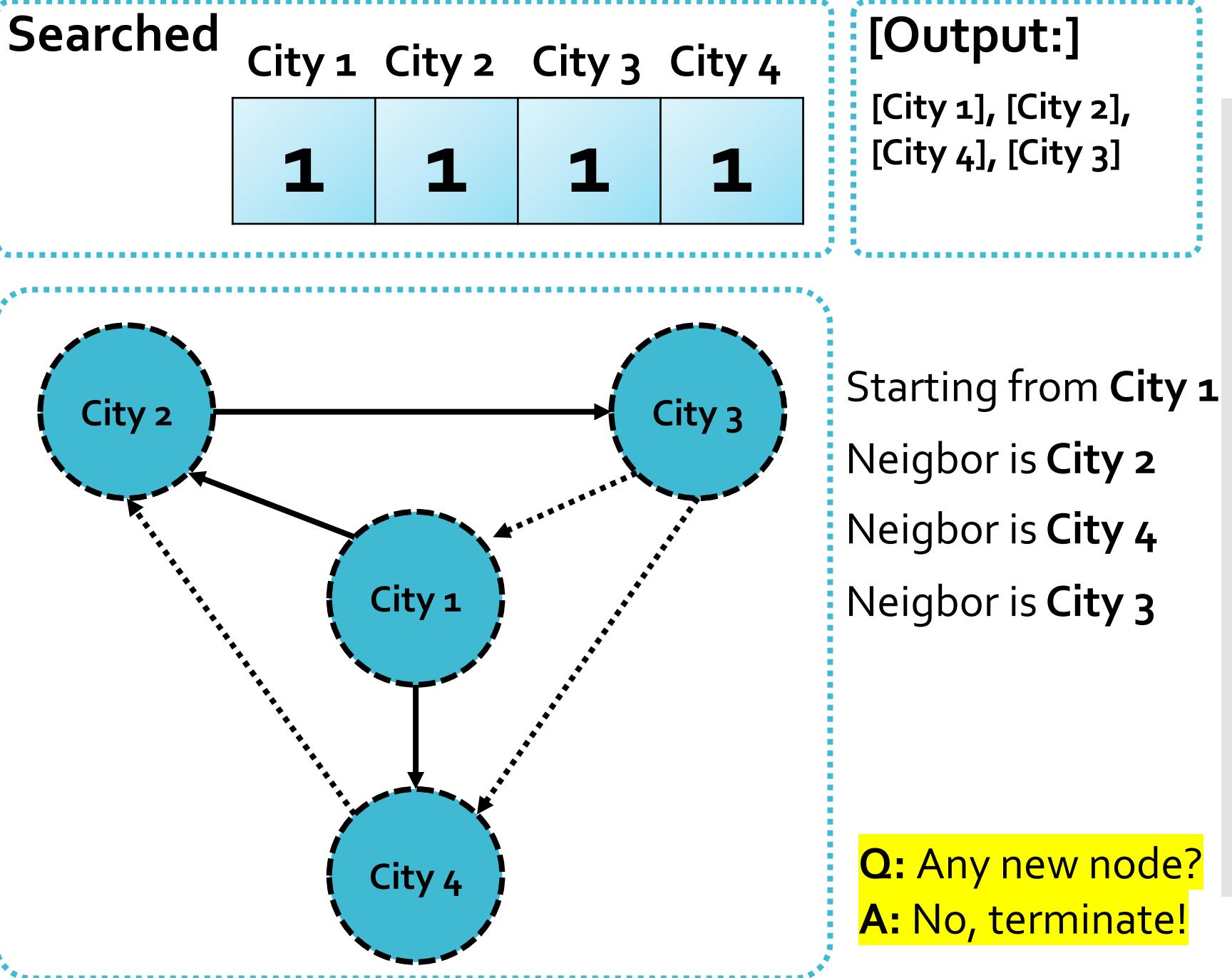
# Breadth First Search (BFS)



# Breadth First Search (BFS)



# Breadth First Search (BFS)



## ❖ Implementation of Breadth First Search (BFS)

Graph

```
class Graph:  
    graph = dict()  
  
    def breadth_first_search(self, node):  
  
        searched = [node]  
        search_queue = [node]  
  
        while search_queue:  
            node = search_queue.pop(0)  
  
            print("[", node, end=" ], ")  
  
            if node in self.graph:  
                for neighbour in self.graph[node]:  
                    if neighbour not in searched:  
                        searched.append(neighbour)  
                        search_queue.append(neighbour)
```

## ❖ Implementation of Breadth First Search (BFS)

Graph

```
class Graph:  
    graph = dict()  
  
    def breadth_first_search(self, node):  
  
        searched = [node]  
        search_queue = [node]  
  
        while search_queue:  
            node = search_queue.pop(0)  
  
            print("[", node, end=" ], ")  
  
            if node in self.graph:  
                for neighbour in self.graph[node]:  
                    if neighbour not in searched:  
                        searched.append(neighbour)  
                        search_queue.append(neighbour)
```

## ❖ Testing the Breadth First Search (BFS)

Graph

```
my_graph = Graph()

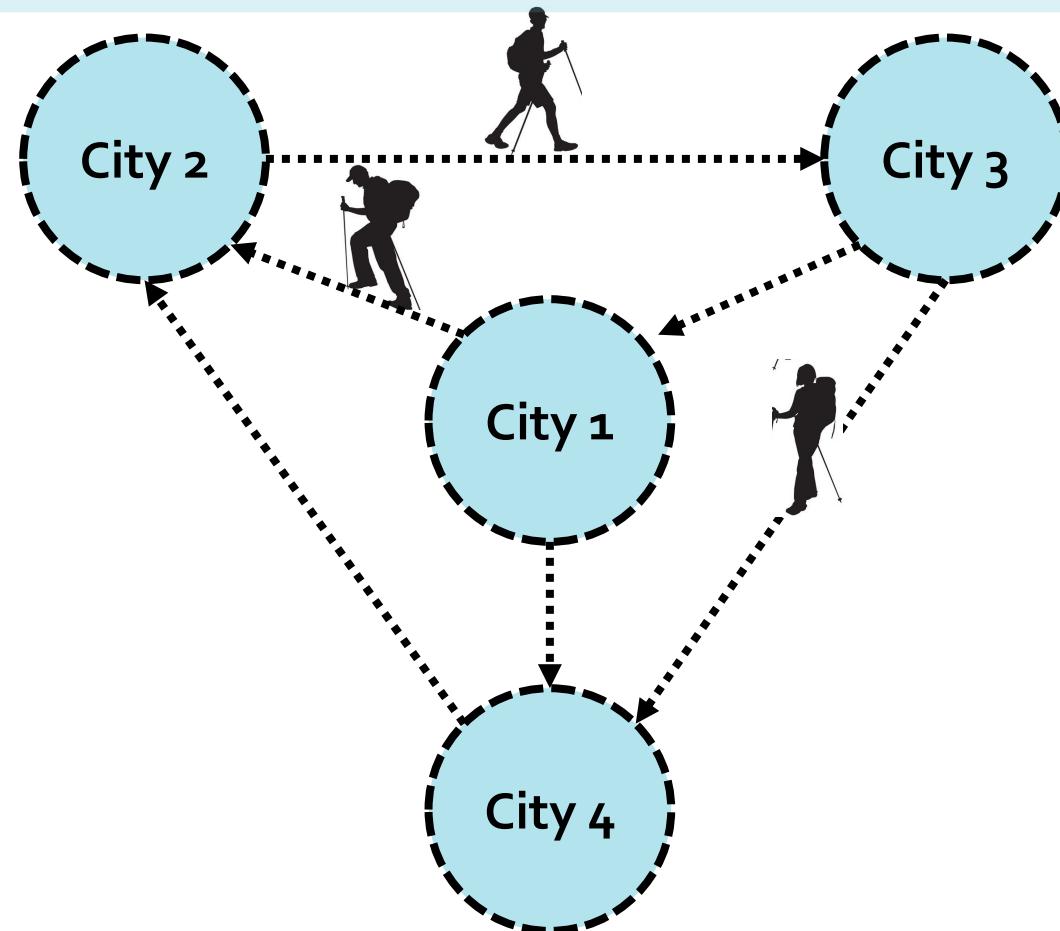
my_graph.add_edge('City 1', 'City 2')
my_graph.add_edge('City 1', 'City 4')
my_graph.add_edge('City 2', 'City 3')
my_graph.add_edge('City 3', 'City 1')
my_graph.add_edge('City 3', 'City 4')
my_graph.add_edge('City 4', 'City 2')

my_graph.breadth_first_search('City 1')

[ City 1 ], [ City 2 ], [ City 4 ], [ City 3 ]
```

# DFS

- ❖ Depth-First Search (DFS) is another traverse algorithm that looks like an algorithm of backtracking.
- ❖ DFS sends out an explorer to a route **as deep as possible** in the graph, and then backtracks and sends it to another route.



# DFS

- ❖ The **idea behind DFS** is to start from a node and move to the adjacent unvisited node and **continue** until there is no unvisited adjacent node.
- ❖ Then **backtrack** and check for other unvisited nodes and traverse them.
- ❖ When a node is visited, it is basically **printed out**.

# DFS

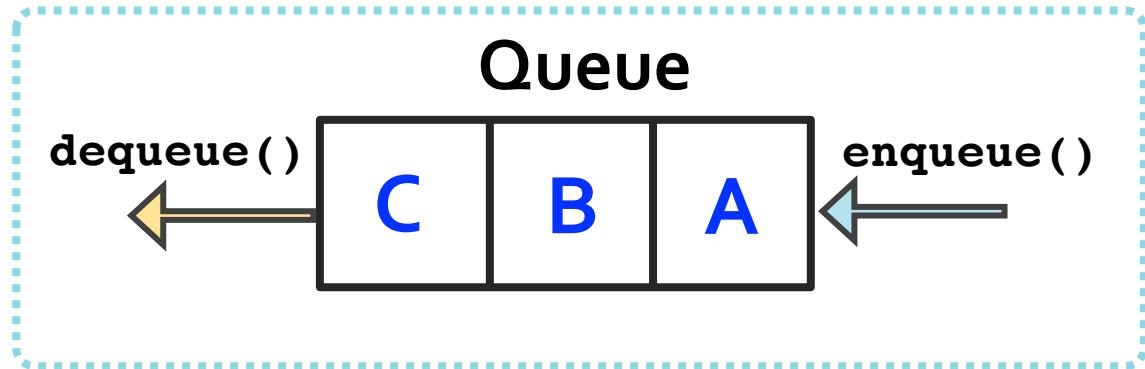
- ❖ **Depth-First Search (DFS) algorithm:**

- ❖ Starts at a node (vertex):

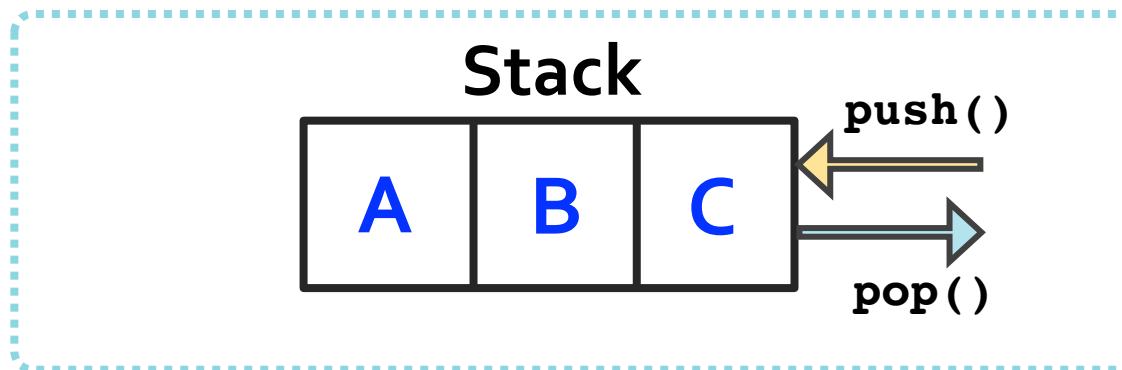
- Check the current node, note it as checked.
    - Traverse the adjacent unmarked node and call the recursive function with each of them.

## Quiz

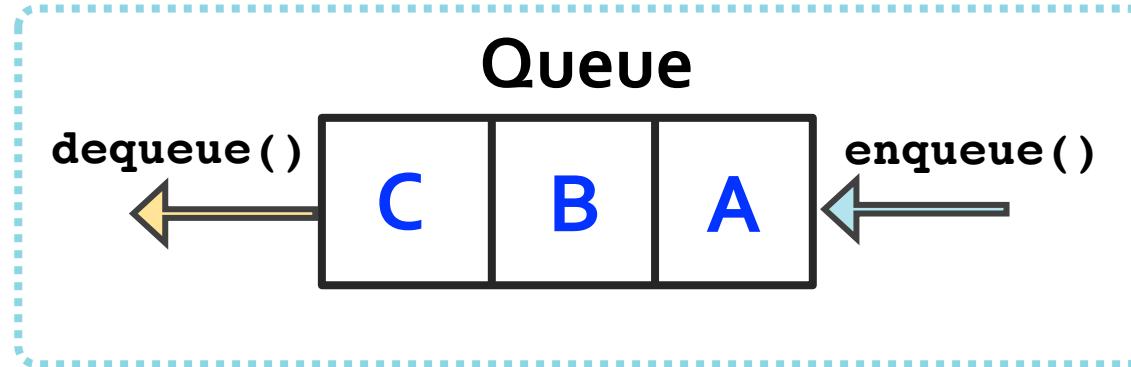
- ❖ Queue can be used to implement BFS.



- ❖ What about DFS?

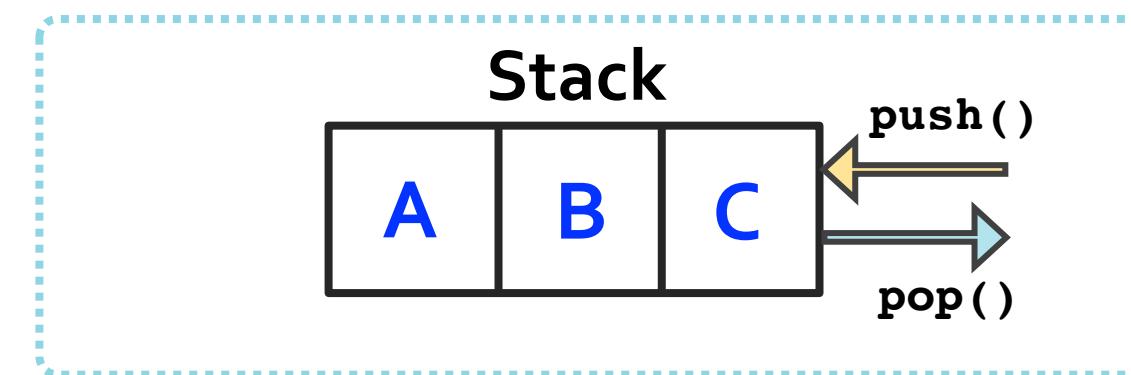


❖ **Queue** can be used to implement **BFS**.



Answer

❖ **Stacks** can be used to implement **DFS**.

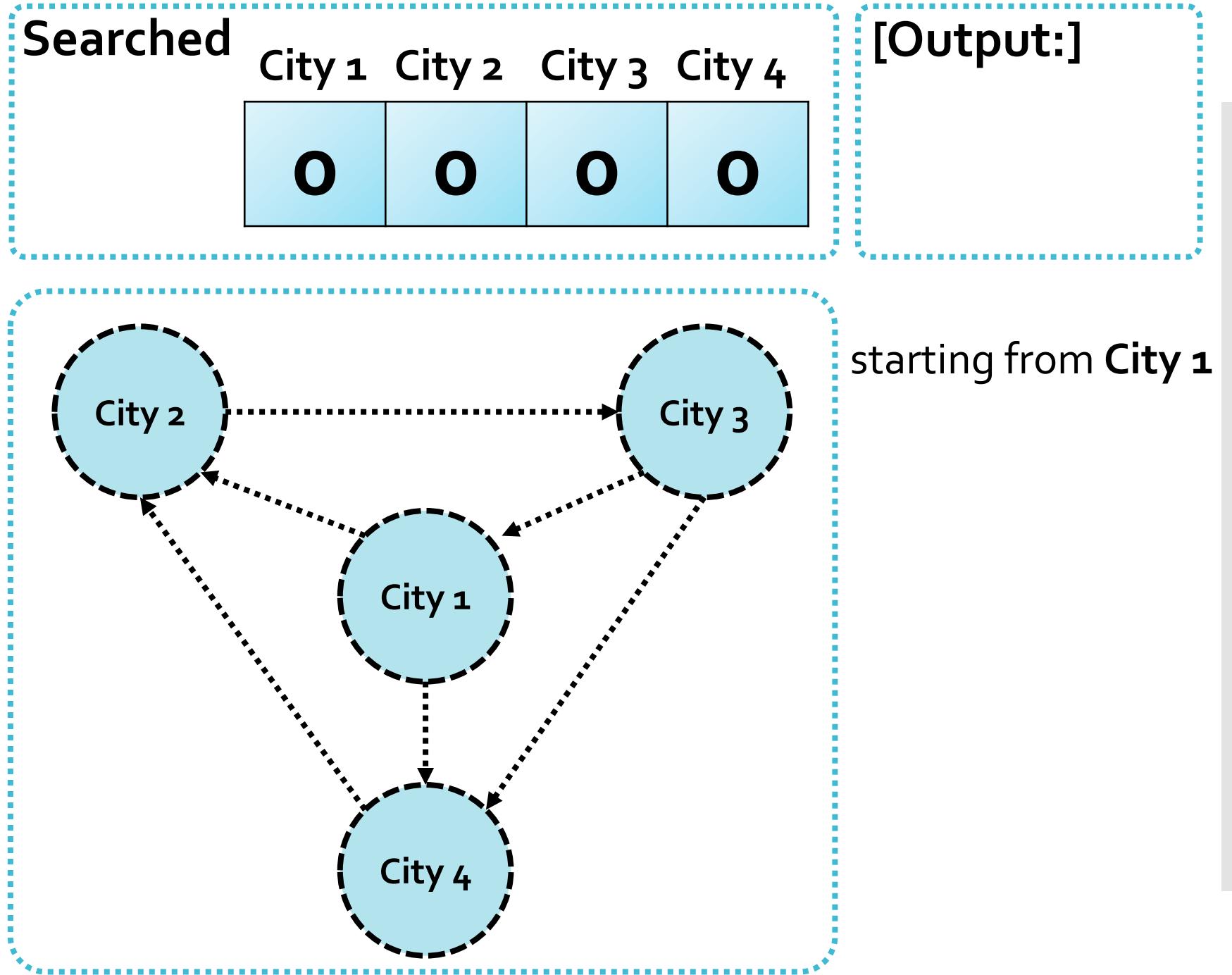


BFS vs DFS

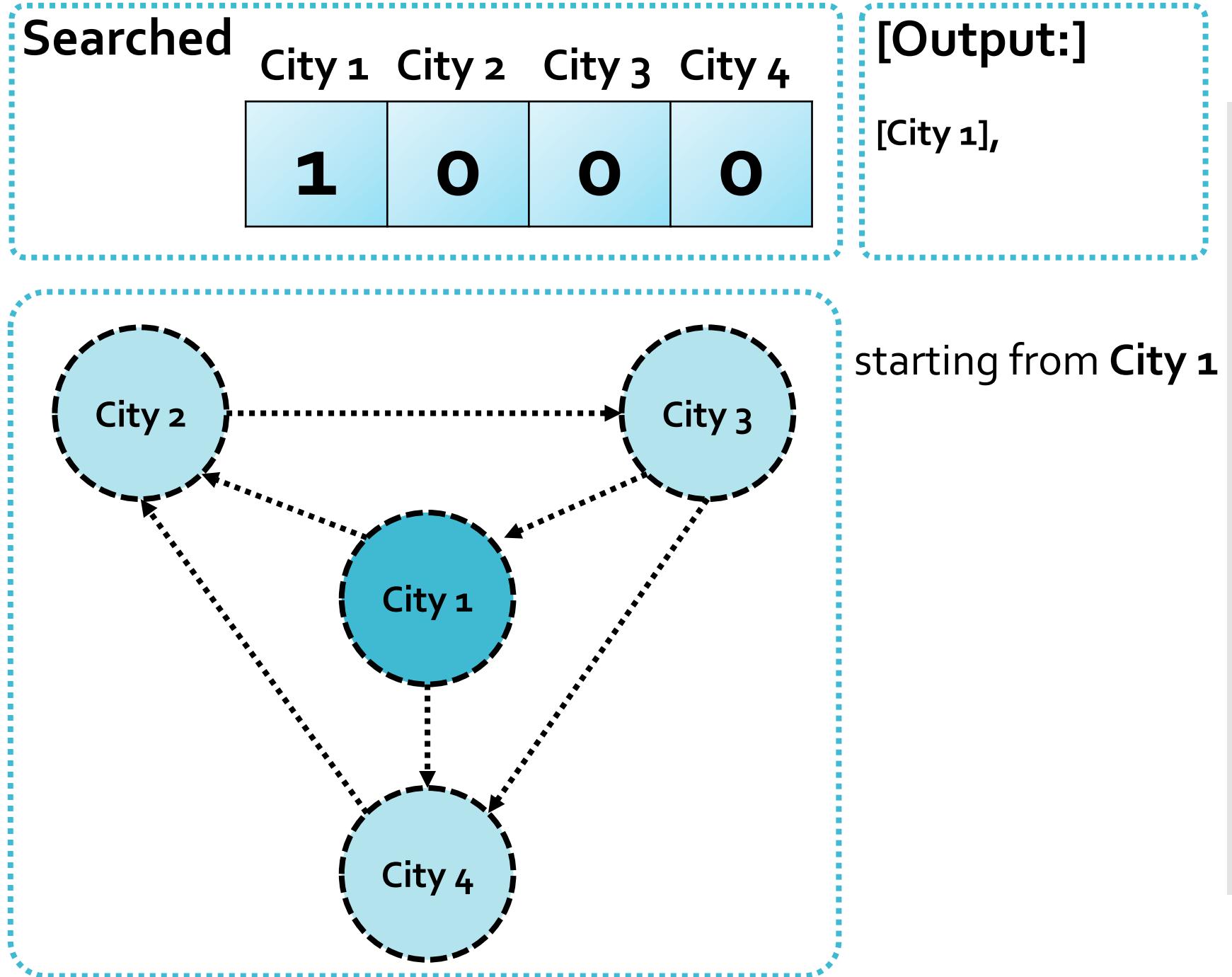


*ref: Youtube*

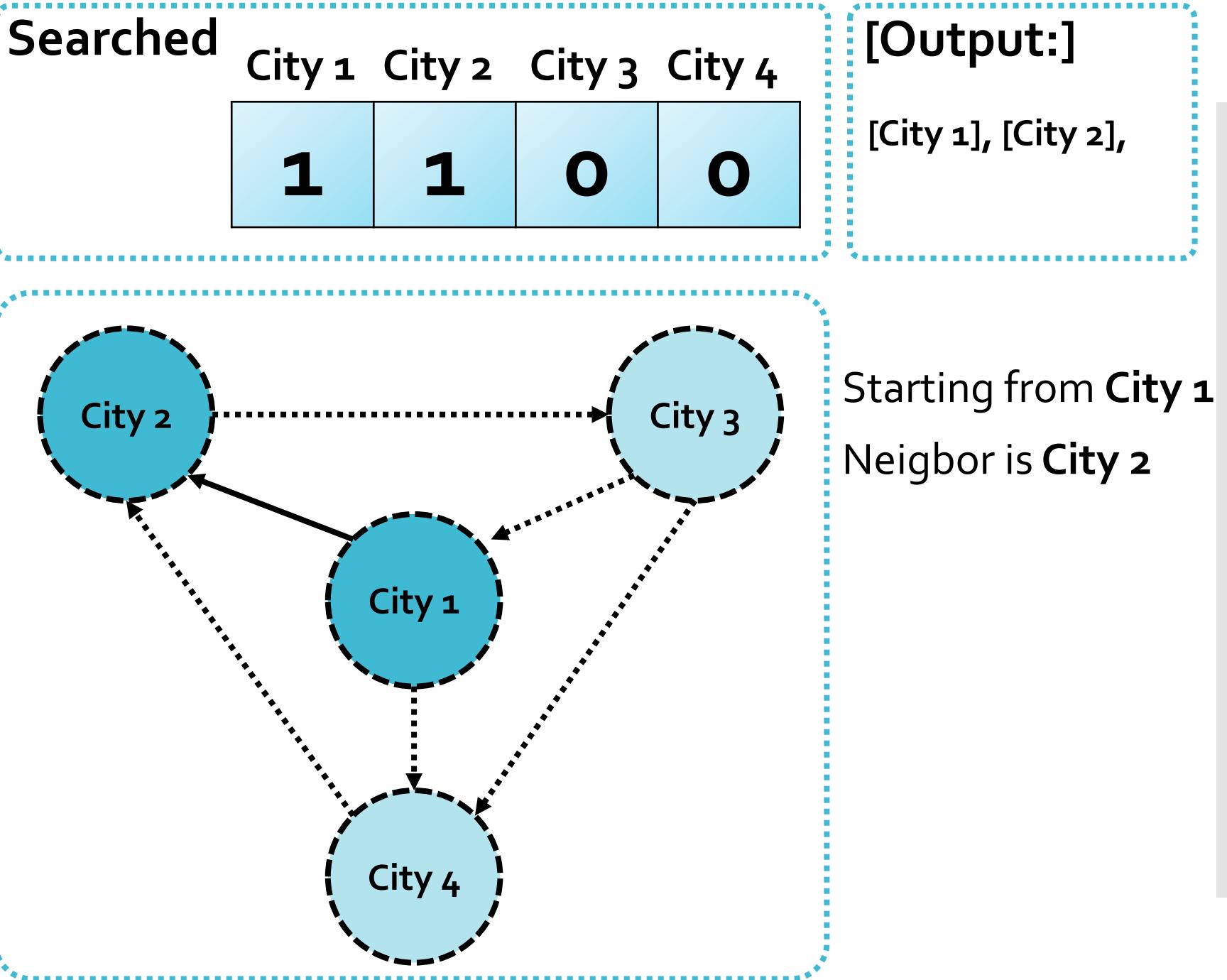
# Depth First Search (DFS)



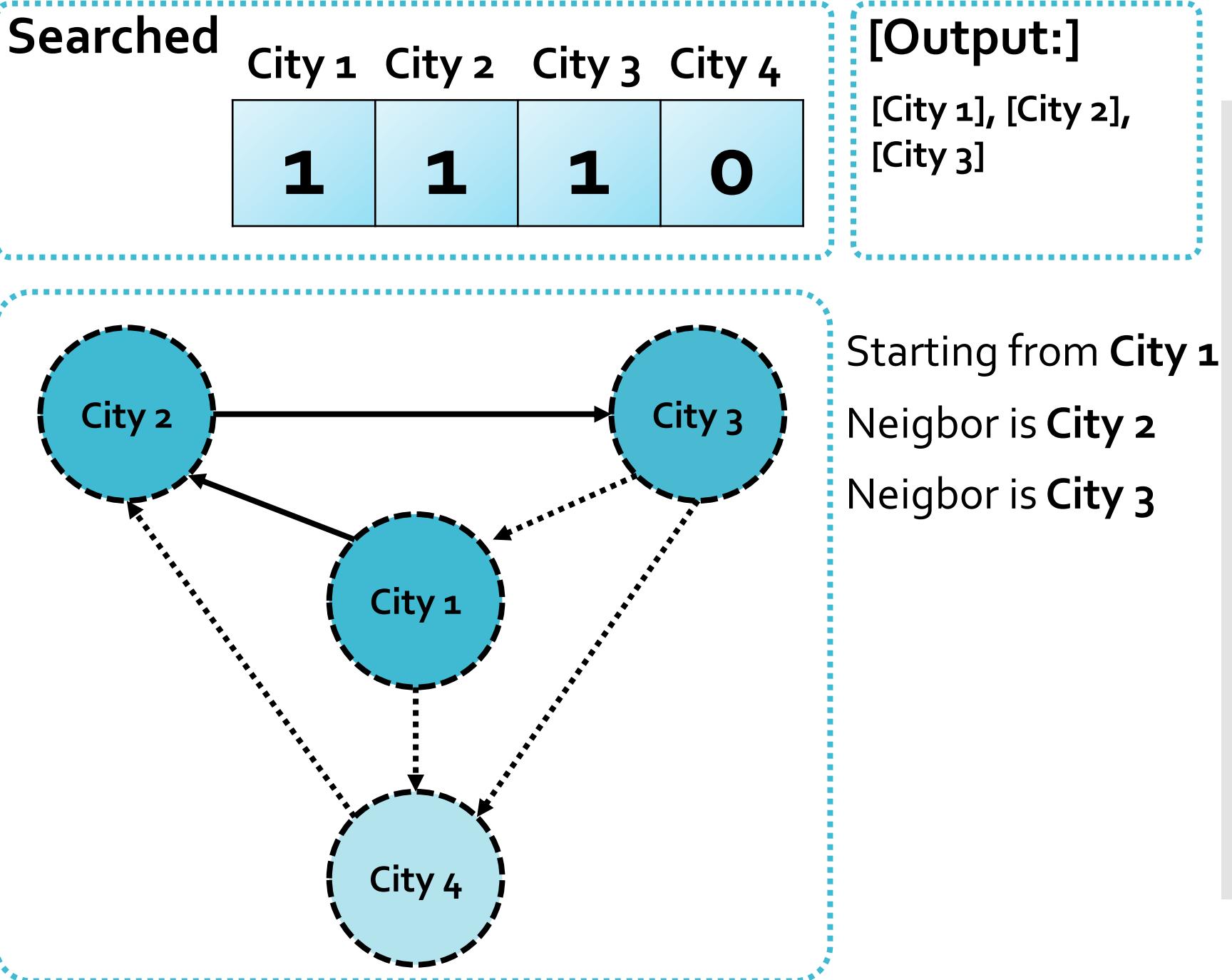
# Depth First Search (DFS)



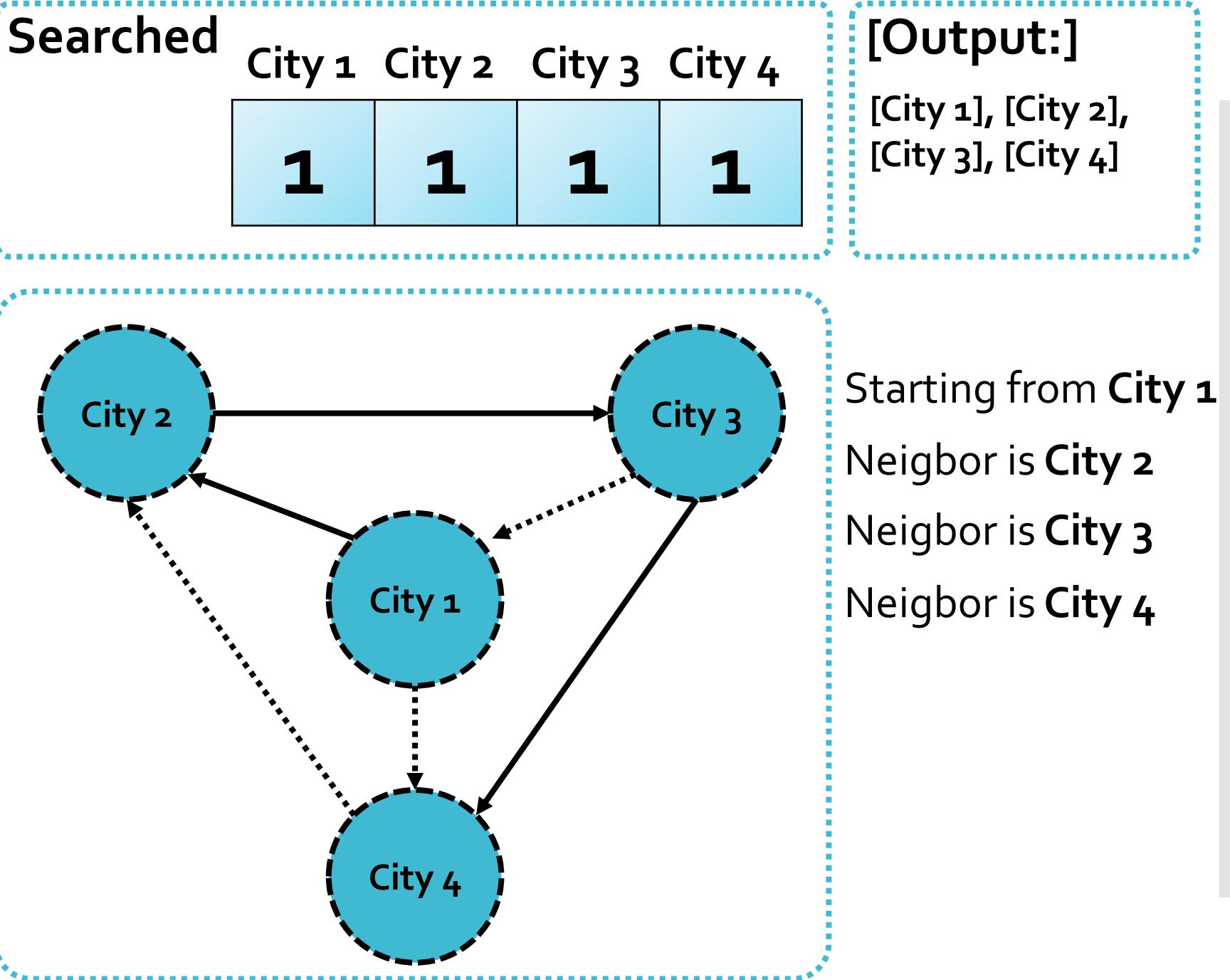
# Depth First Search (DFS)



# Depth First Search (DFS)

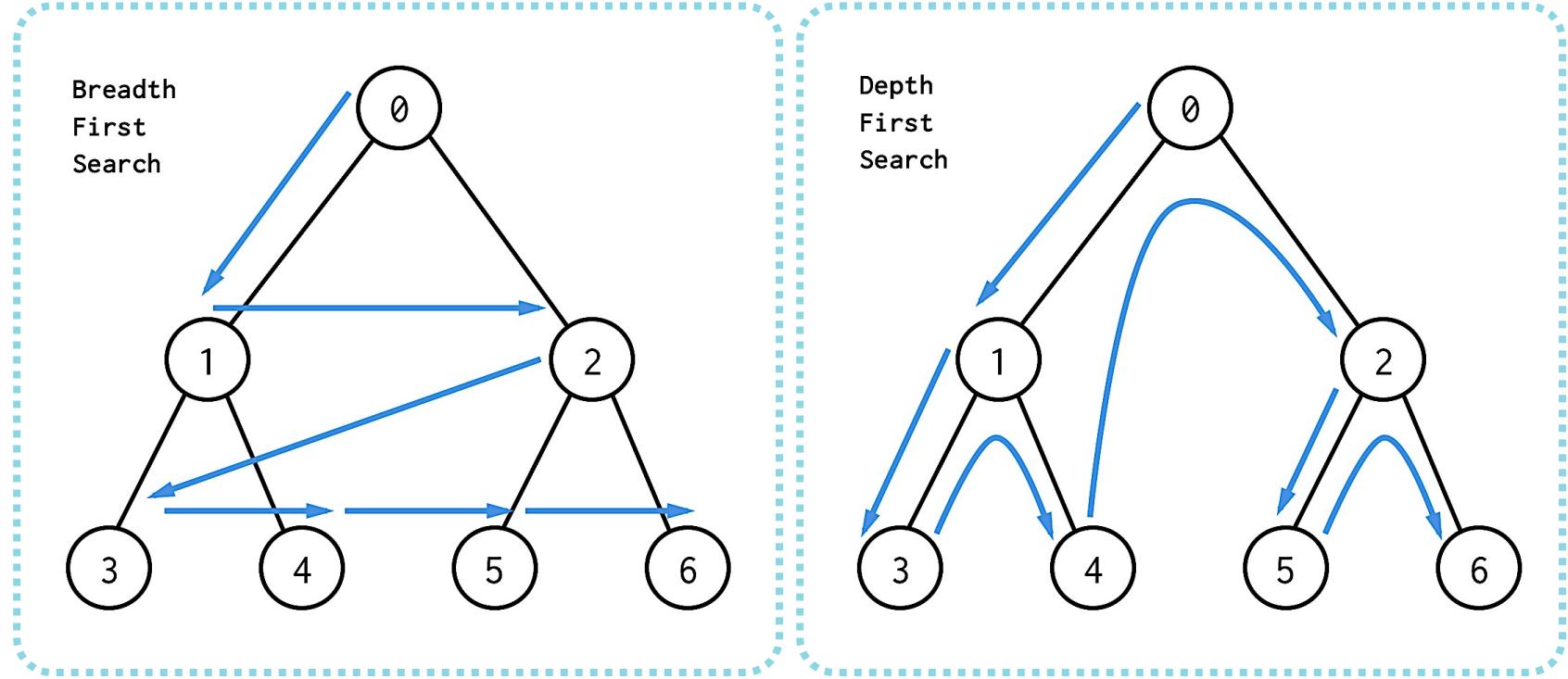


# Depth First Search (DFS)



# BFS vs DFS

## ❖ Breadth First Search (BFS) vs Depth First Search (DFS)



## ❖ Implementation of Depth First Search (DFS)

Graph

```
class Graph:  
    graph = dict()  
    searched = []  
  
    def depth_first_search(self, node):  
        if node not in self.searched:  
  
            print("[", node, end=" ],")  
  
            self.searched.append(node)  
            if node in self.graph:  
                for neighbour in self.graph[node]:  
                    self.depth_first_search(neighbour)
```

## ❖ Testing the Depth First Search (DFS)

Graph

```
my_graph = Graph()

my_graph.add_edge('City 1', 'City 2')
my_graph.add_edge('City 1', 'City 4')
my_graph.add_edge('City 2', 'City 3')
my_graph.add_edge('City 3', 'City 1')
my_graph.add_edge('City 3', 'City 4')
my_graph.add_edge('City 4', 'City 2')

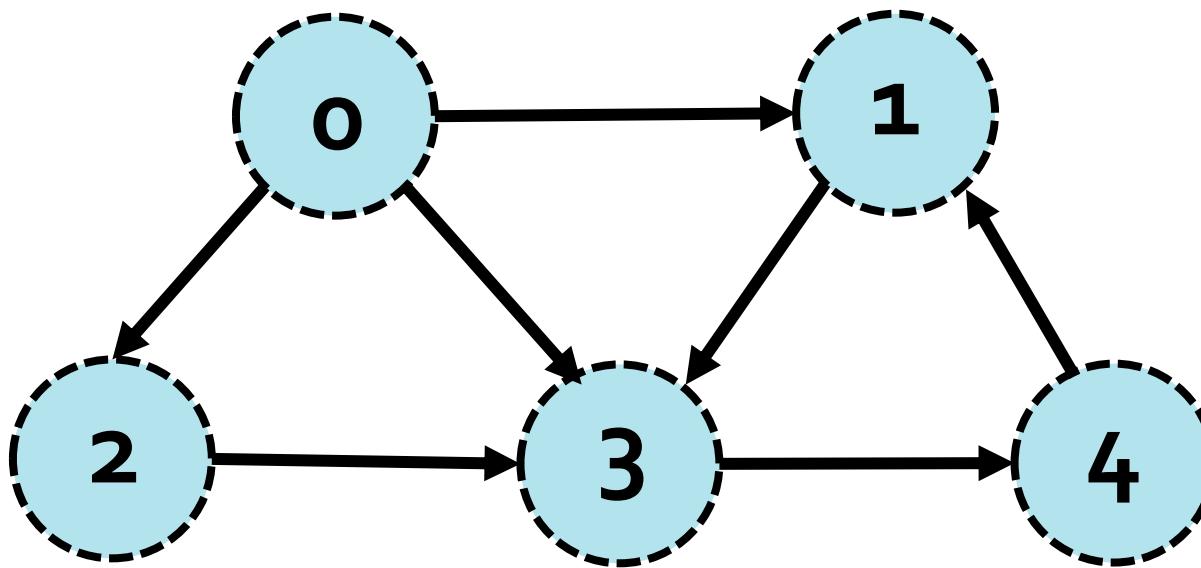
my_graph.depth_first_search('City 1')
```

[Output:]

```
[ City 1 ],[ City 2 ],[ City 3 ],[ City 4 ]
```

## Quiz

- ❖ Write a function **build\_my\_graph()** that:
- a) creates the following Graph.
  - b) runs Breadth First Search (BFS) starting from o.
  - c) runs Depth First Search (DFS) starting from o.



## Answer

```
def build_my_graph():
    my_graph = Graph()

    my_graph.add_edge(0, 1)
    my_graph.add_edge(0, 2)
    my_graph.add_edge(0, 3)
    my_graph.add_edge(1, 3)
    my_graph.add_edge(2, 3)
    my_graph.add_edge(3, 4)
    my_graph.add_edge(4, 1)

print('Breadth First Search (BFS):')
my_graph.breadth_first_search(0)
print('Depth First Search (DFS):')
my_graph.depth_first_search(0)
```

## Answer

```
build_my_graph()
```

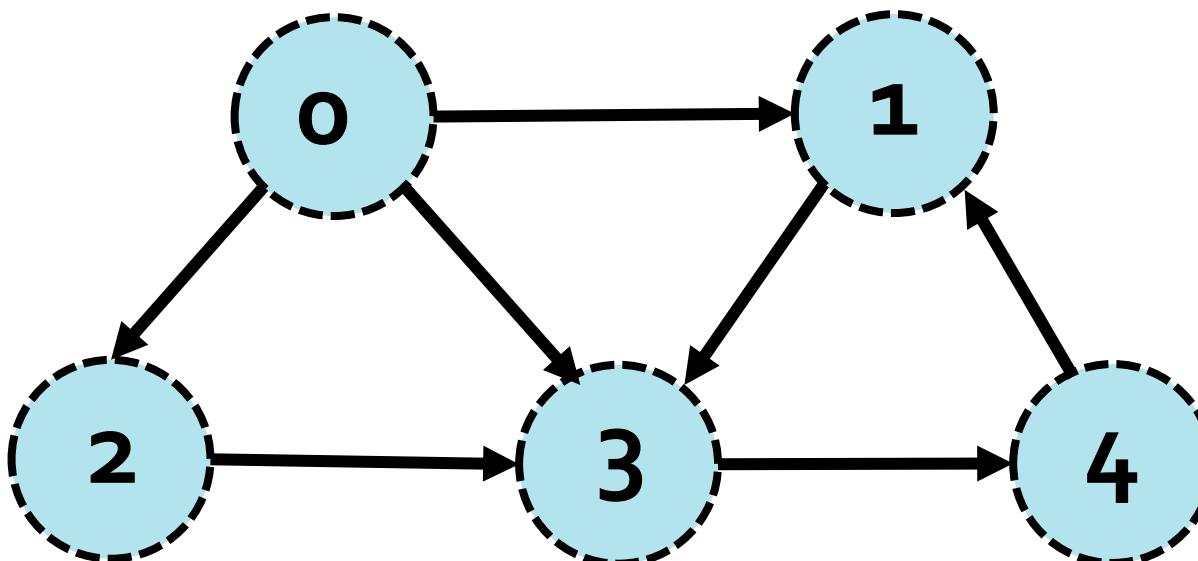
[Output:]

Breadth First Search (BFS):

```
[ 0 ], [ 1 ], [ 2 ], [ 3 ], [ 4 ]
```

Depth First Search (DFS):

```
[ 0 ], [ 1 ], [ 3 ], [ 4 ], [ 2 ]
```



## Next Lesson

- ❖ More Shortest Paths Algorithms (Dijkstra)