

# Graphs

## Unweighted Graphs

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# Today we're going to cover

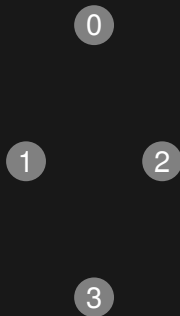
- ▶ Graph basics
- ▶ Graph representation (recap)
- ▶ Depth-first search
- ▶ Connected components
- ▶ DFS tree
- ▶ Bridges
- ▶ Strongly connected components
- ▶ Topological sort
- ▶ Breadth-first search
- ▶ Shortest paths in unweighted graphs

# What is a graph?

# What is a graph?

## ► Vertices

- Road intersections
- Computers
- Floors in a house
- Objects



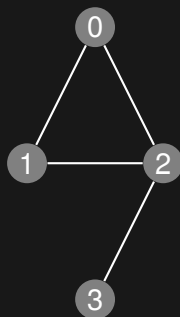
# What is a graph?

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- Floors in a house
- Objects

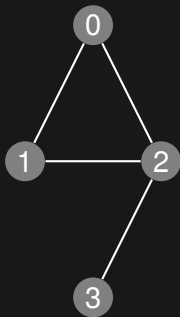
## ► Edges

- Roads
- Ethernet cables
- Stairs or elevators
- Relation between objects



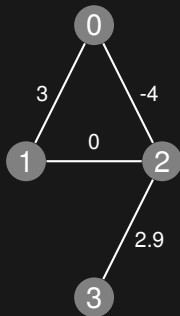
# Types of edges

- Unweighted



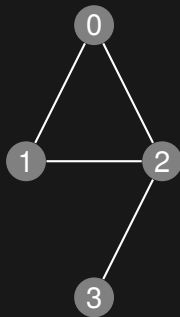
# Types of edges

- Unweighted or **Weighted**



# Types of edges

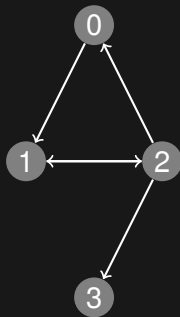
- ▶ Unweighted or Weighted
- ▶ Undirected





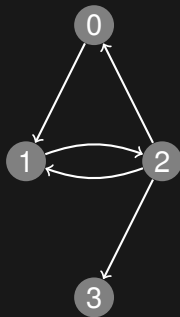
# Types of edges

- ▶ Unweighted or Weighted
- ▶ Undirected or **Directed**

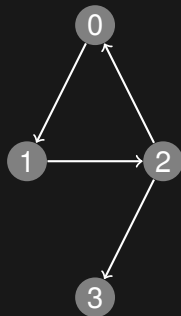


# Types of edges

- ▶ Unweighted or Weighted
- ▶ Undirected or **Directed**

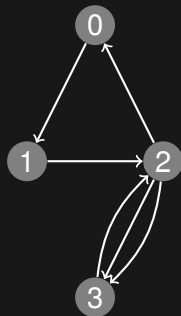


# Multigraphs



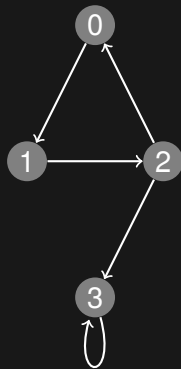
# Multigraphs

- Multiple edges



# Multigraphs

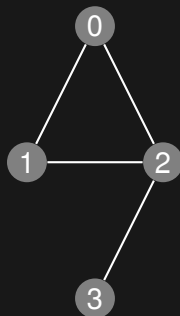
- ▶ Multiple edges
- ▶ Self-loops



# Adjacency list

0: 1, 2  
1: 0, 2  
2: 0, 1, 3  
3: 2

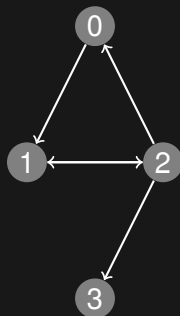
```
vector<int> adj[4];  
adj[0].push_back(1);  
adj[0].push_back(2);  
adj[1].push_back(0);  
adj[1].push_back(2);  
adj[2].push_back(0);  
adj[2].push_back(1);  
adj[2].push_back(3);  
adj[3].push_back(2);
```



# Adjacency list (directed)

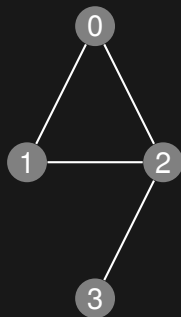
0: 1  
1: 2  
2: 0, 1, 3  
3:

```
vector<int> adj[4];  
adj[0].push_back(1);  
adj[1].push_back(2);  
adj[2].push_back(0);  
adj[2].push_back(1);  
adj[2].push_back(3);
```



# Vertex properties (undirected graph)

- ▶ Degree of a vertex
  - Number of adjacent edges
  - Number of adjacent vertices

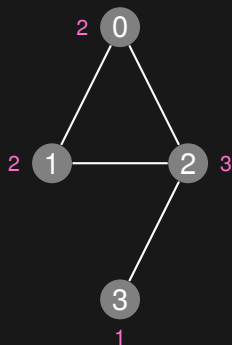




# Vertex properties (undirected graph)

## ► Degree of a vertex

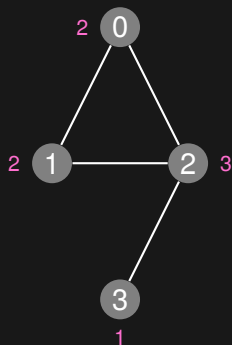
- Number of adjacent edges
- Number of adjacent vertices



# Vertex properties (undirected graph)

- ▶ Degree of a vertex
  - Number of adjacent edges
  - Number of adjacent vertices
- ▶ Handshaking lemma

$$\sum_{v \in V} \deg(v) = 2|V|$$

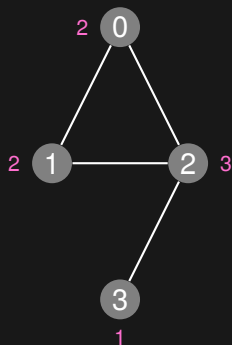


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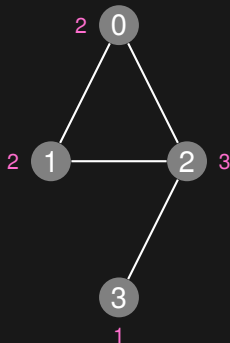
$$2 + 2 + 3 + 1 = 2 \times 4$$



# Vertex properties (undirected graph)

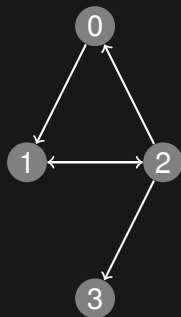
```
0: 1, 2  
1: 0, 2  
2: 0, 1, 3  
3: 2
```

```
adj[0].size() // 2  
adj[1].size() // 2  
adj[2].size() // 3  
adj[3].size() // 1
```



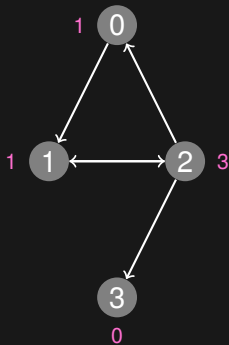
# Vertex properties (directed graph)

- ▶ Outdegree of a vertex
  - Number of outgoing edges



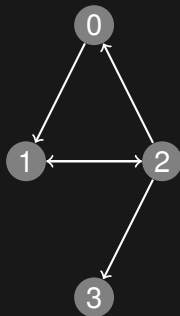
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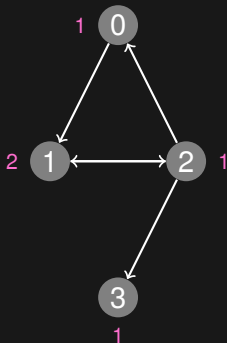
# Vertex properties (directed graph)

- ▶ Outdegree of a vertex
  - Number of outgoing edges
- ▶ Indegree of a vertex
  - Number of incoming edges



# Vertex properties (directed graph)

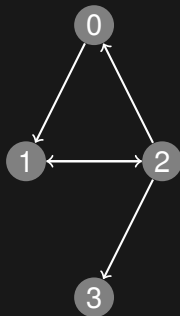
- ▶ Outdegree of a vertex
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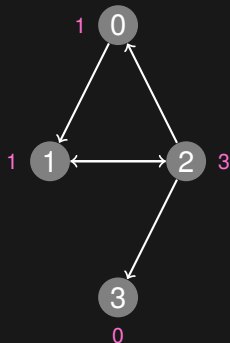
# Vertex properties (directed graph)

- ▶ Outdegree of a vertex
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# Vertex properties (directed graph)

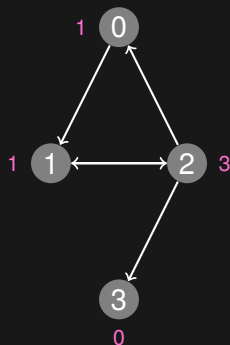
- ▶ Outdegree of a vertex
  - Number of outgoing edges
- ▶ Indegree of a vertex
  - Number of incoming edges



# Adjacency list (directed)

```
0: 1  
1: 2  
2: 0, 1, 3  
3:
```

```
adj[0].size() // 1  
adj[1].size() // 1  
adj[2].size() // 3  
adj[3].size() // 0
```



# Paths

- Path / Walk / Trail:

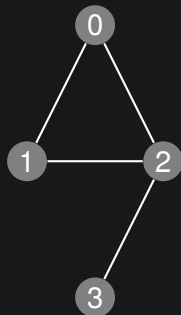
$$e_1 e_2 \dots e_k$$

such that

$$e_i \in E$$

$$e_i = e_j \Rightarrow i = j$$

$$\text{to}(e_i) = \text{from}(e_{i+1})$$



# Paths

- Path / Walk / Trail:

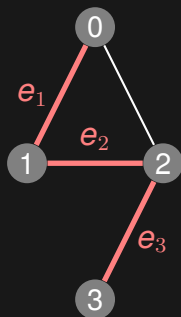
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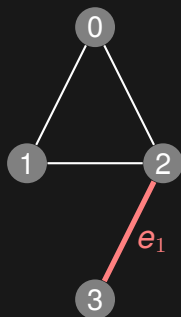
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- Path / Walk / Trail:

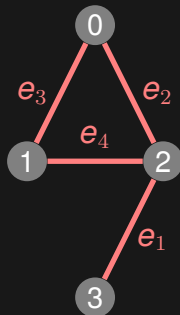
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# Cycles

- Cycle / Circuit / Tour:

$$e_1 e_2 \dots e_k$$

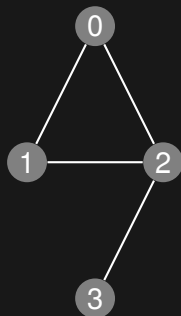
such that

$$e_i \in E$$

$$e_i = e_j \Rightarrow i = j$$

$$\text{to}(e_i) = \text{from}(e_{i+1})$$

$$\text{from}(e_1) = \text{to}(e_k)$$





# Cycles

- Cycle / Circuit / Tour:

$$e_1 e_2 \dots e_k$$

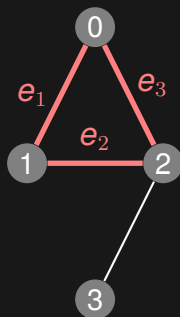
such that

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$$\text{to}(e_i) = \text{from}(e_{i+1})$$

$$\text{from}(e_1) = \text{to}(e_k)$$



# Cycles

- Cycle / Circuit / Tour:

$$e_1 e_2 \dots e_k$$

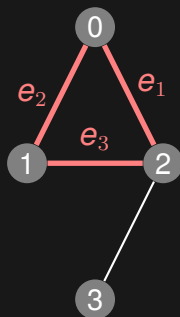
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$$\text{from}(e_1) = \text{to}(e_k)$$



# Cycles

- Cycle / Circuit / Tour:

$$e_1 e_2 \dots e_k$$

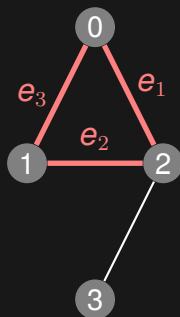
such that

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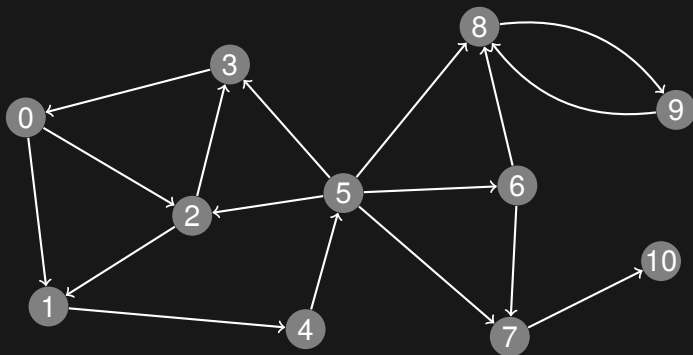
$$\text{from}(e_1) = \text{to}(e_k)$$



# Depth-first search

- ▶ Given a graph (either directed or undirected) and two vertices  $u$  and  $v$ , does there exist a path from  $u$  to  $v$ ?
- ▶ Depth-first search is an algorithm for finding such a path, if one exists
- ▶ It traverses the graph in depth-first order, starting from the initial vertex  $u$
- ▶ We don't actually have to specify a  $v$ , since we can just let it visit all reachable vertices from  $u$  (and still same time complexity)
- ▶ But what is the time complexity?
- ▶ Each vertex is visited once, and each edge is traversed once
- ▶  $O(n + m)$

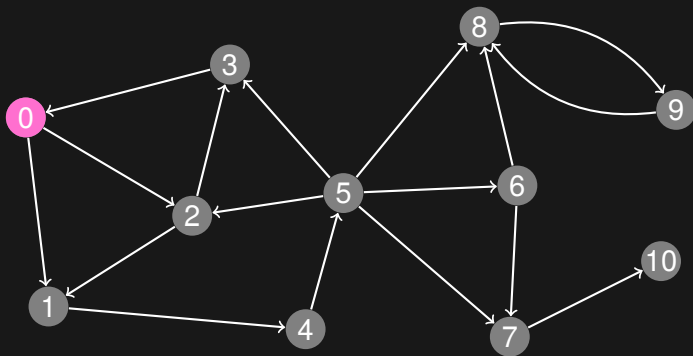
# Depth-first search



Stack: |

	0	1	2	3	4	5	6	7	8	9	10
marked	0	0	0	0	0	0	0	0	0	0	0

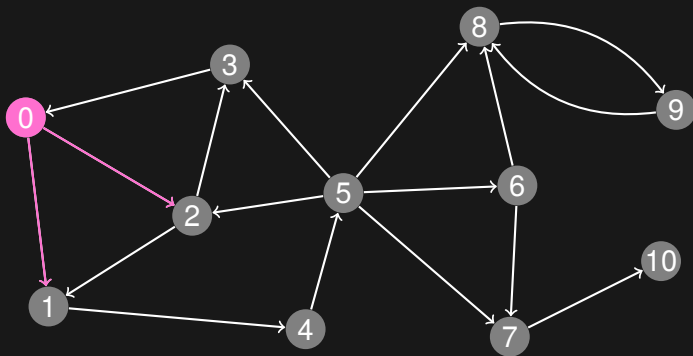
# Depth-first search



Stack: 0 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	0	0	0	0	0	0	0	0	0	0

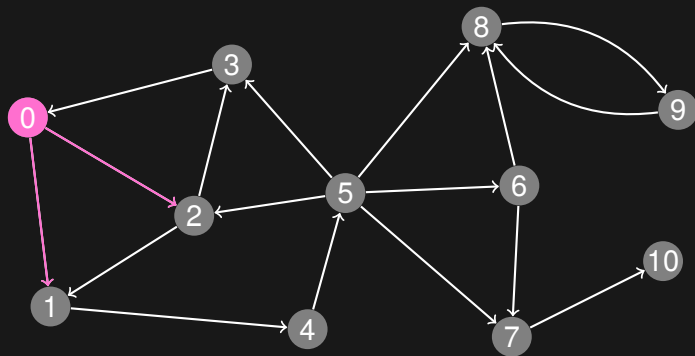
# Depth-first search



Stack: 0 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	0	0	0	0	0	0	0	0	0	0

# Depth-first search

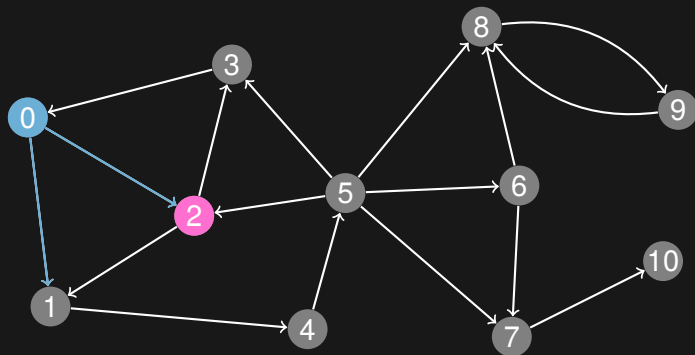


Stack: 0 | 2 1

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	0	0	0	0	0	0	0



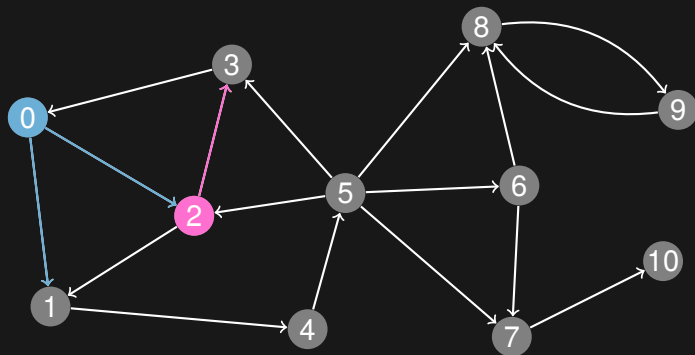
# Depth-first search



Stack: 2 | 1

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	0	0	0	0	0	0	0

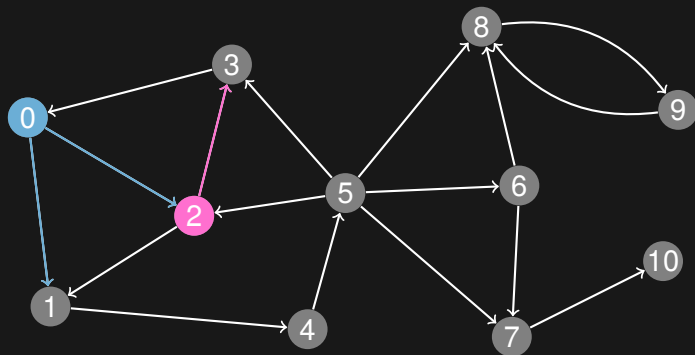
# Depth-first search



Stack: 2 | 1

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	0	0	0	0	0	0	0

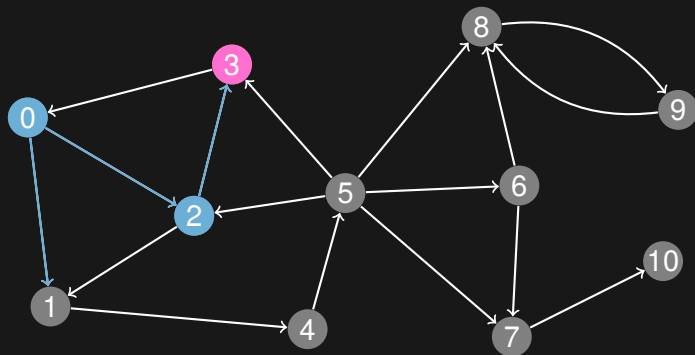
# Depth-first search



Stack: 2 | 3 1

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	0	0	0	0	0	0	0

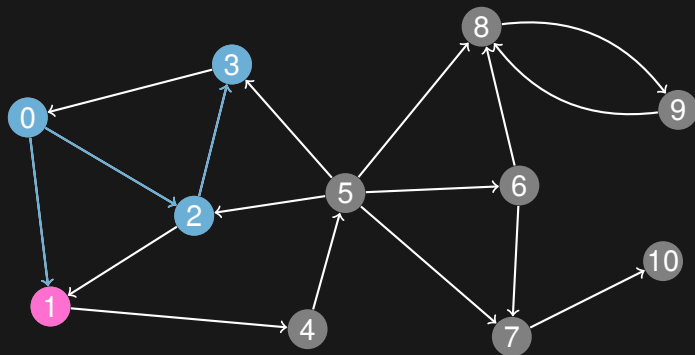
# Depth-first search



Stack: 3 | 1

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	0	0	0	0	0	0	0

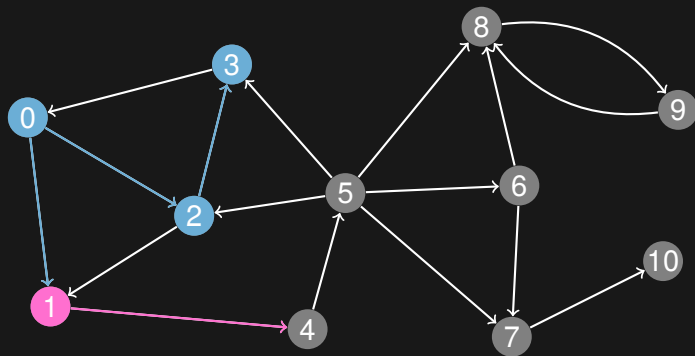
# Depth-first search



Stack: 1 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	0	0	0	0	0	0	0

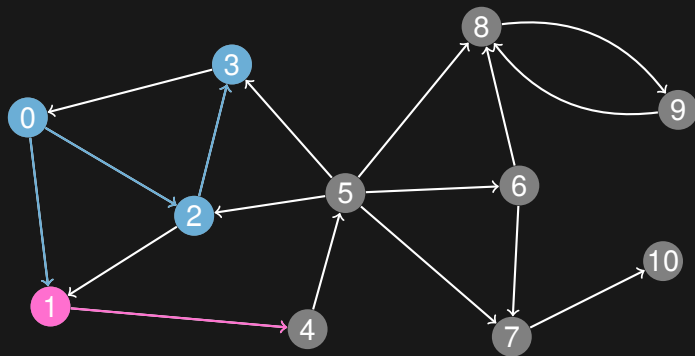
# Depth-first search



Stack: 1 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	0	0	0	0	0	0	0

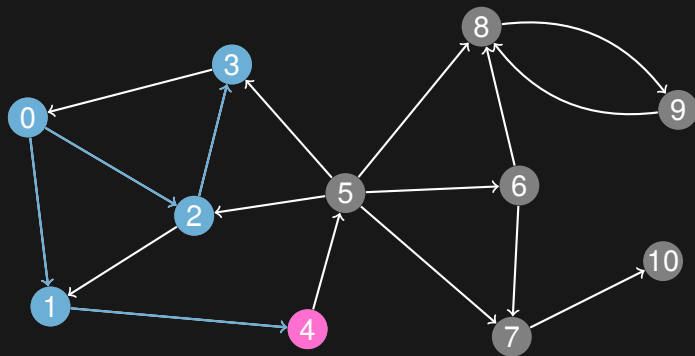
# Depth-first search



Stack: 1 | 4

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	0	0	0	0	0	0

# Depth-first search

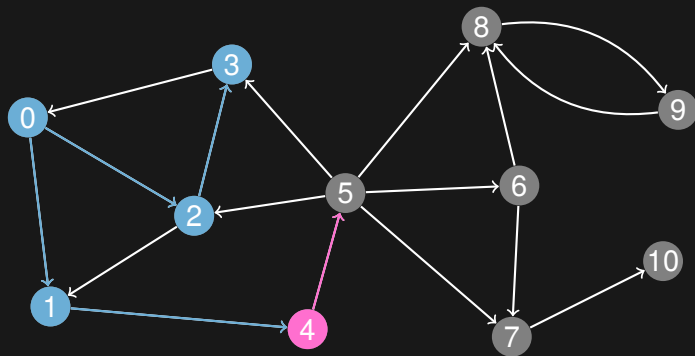


Stack: 4 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	0	0	0	0	0	0



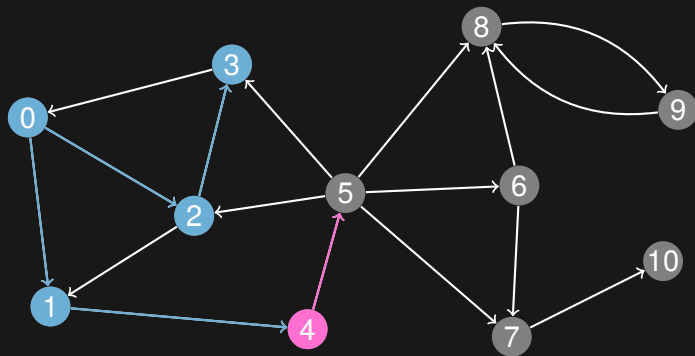
# Depth-first search



Stack: 4 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	0	0	0	0	0	0

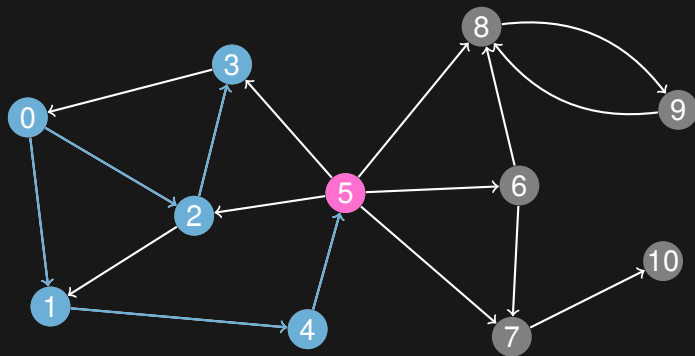
# Depth-first search



Stack: 4 | 5

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	0	0	0	0	0

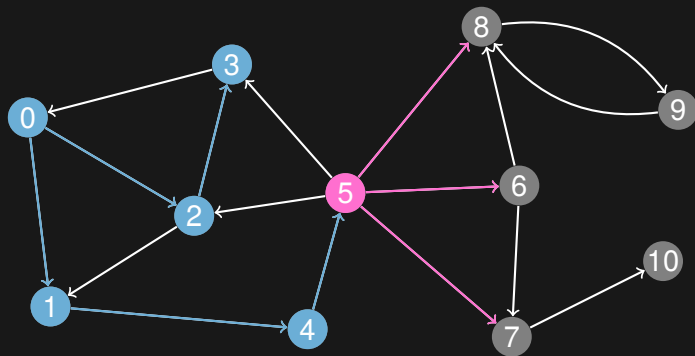
# Depth-first search



Stack: 5 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	0	0	0	0	0

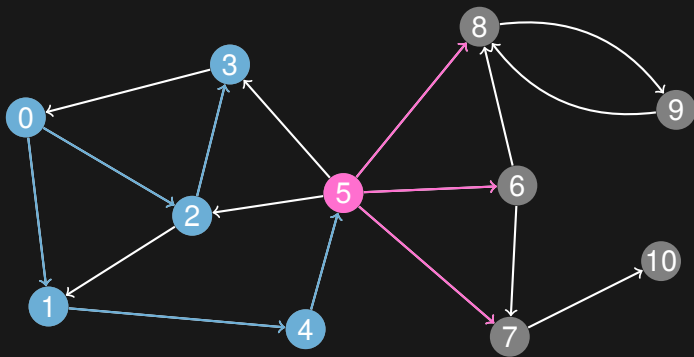
# Depth-first search



Stack: 5 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	0	0	0	0	0

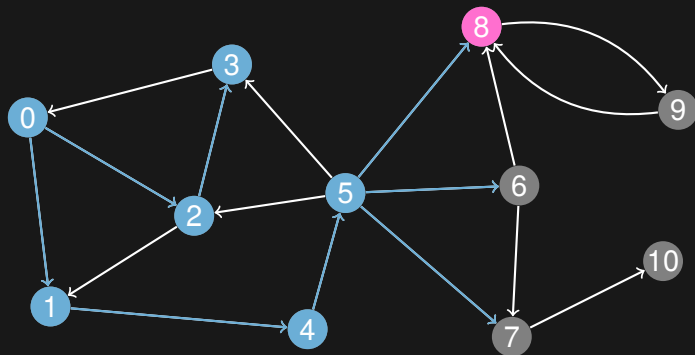
# Depth-first search



Stack: 5 | 8 6 7

[illegible]

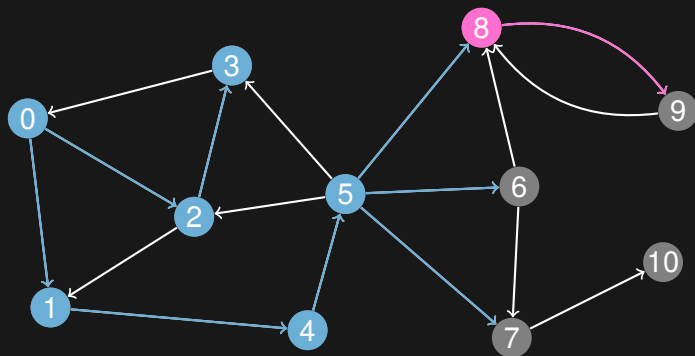
# Depth-first search



Stack: 8 | 6 7

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	0

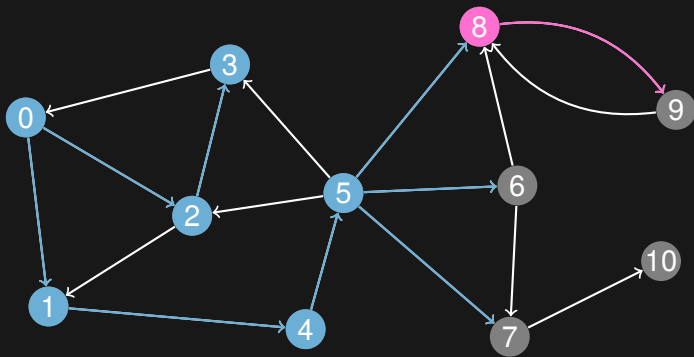
# Depth-first search



Stack: 8 | 6 7

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	0

# Depth-first search

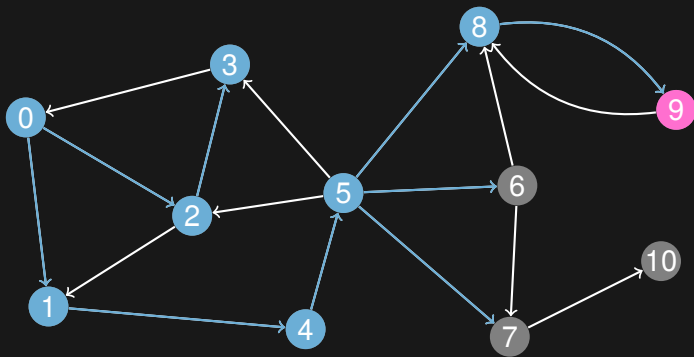


Stack: 8 | 9 6 7

[illegible]



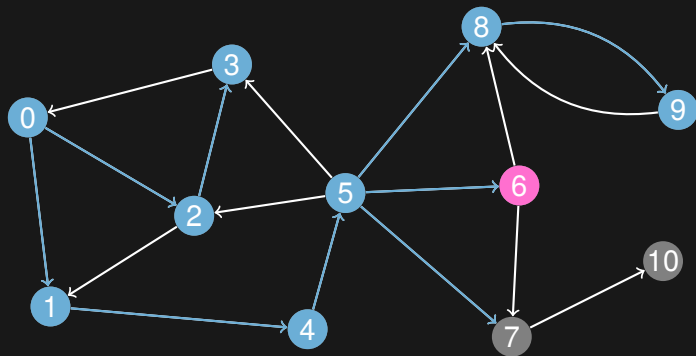
# Depth-first search



Stack: 9 | 6 7

[illegible]

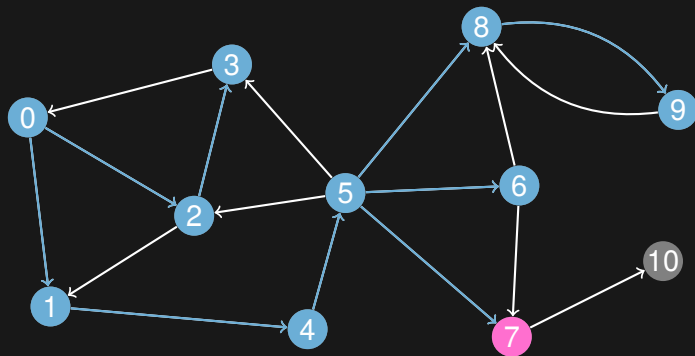
# Depth-first search



Stack: 6 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	0

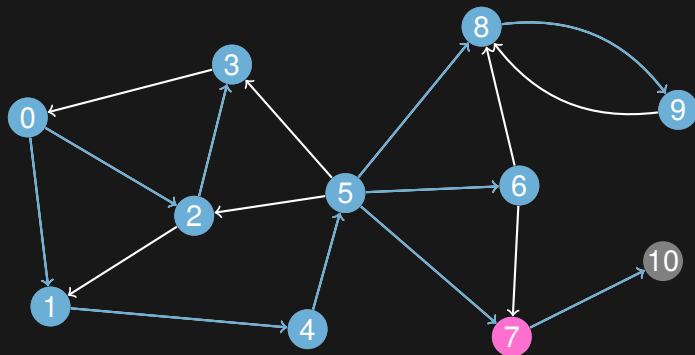
# Depth-first search



Stack: 7 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	0

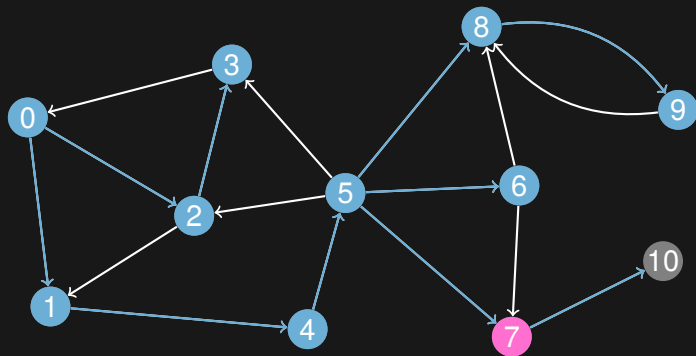
# Depth-first search



Stack: 7 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	0

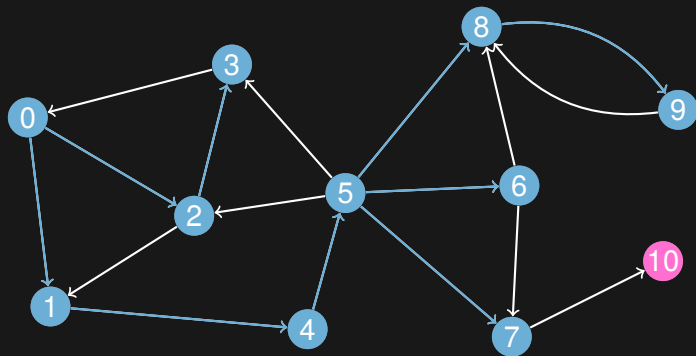
# Depth-first search



Stack: 7 | 10

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	1

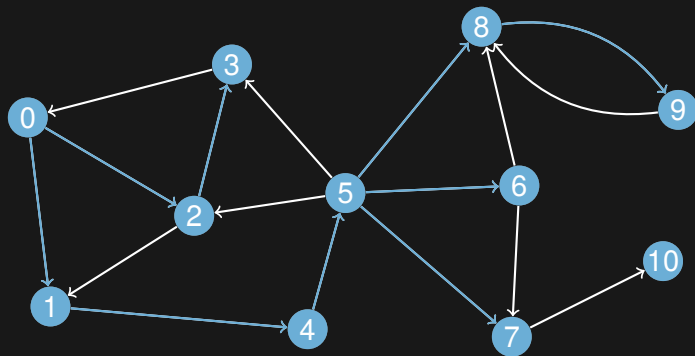
# Depth-first search



Stack: 10 |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	1

# Depth-first search



Stack: |

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	1

# Depth-first search

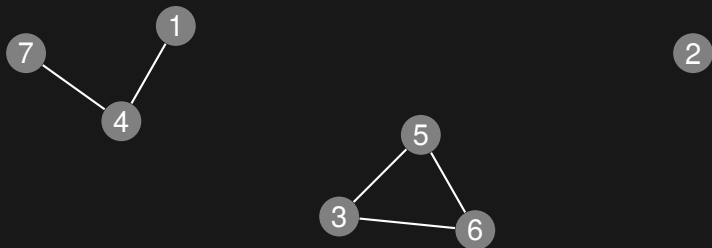
```
vector<int> adj[1000];  
vector<bool> visited(1000, false);  
  
void dfs(int u) {  
    if (visited[u]) {  
        return;  
    }  
  
    visited[u] = true;  
  
    for (int i = 0; i < adj[u].size(); i++) {  
        int v = adj[u][i];  
        dfs(v);  
    }  
}
```



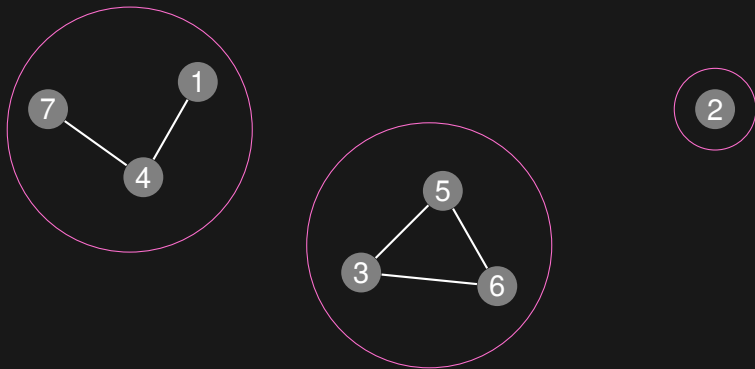
# Connected components

- ▶ An undirected graph can be partitioned into connected components
- ▶ A connected component is a maximal subset of the vertices such that each pair of vertices is reachable from each other
- ▶ We've already seen this in a couple of problems, but we've been using Union-Find to keep track of the components

# Connected components



# Connected components



# Connected components

- ▶ Also possible to find these components using depth-first search
- ▶ Pick some vertex we don't know anything about, and do a depth-first search out from it
- ▶ All vertices reachable from that starting vertex are in the same component
- ▶ Repeat this process until you have all the components
- ▶ Time complexity is  $O(n + m)$

# Connected components

```
vector<int> adj[1000];
vector<int> component(1000, -1);

void find_component(int cur_comp, int u) {
    if (component[u] != -1) {
        return;
    }

    component[u] = cur_comp;

    for (int i = 0; i < adj[u].size(); i++) {
        int v = adj[u][i];
        find_component(cur_comp, v);
    }
}

int components = 0;
for (int u = 0; u < n; u++) {
    if (component[u] == -1) {
        find_component(components, u);
        components++;
    }
}
```

# Depth-first search tree

- ▶ When we do a depth-first search from a certain vertex, the path that we take forms a tree
- ▶ When we go from a vertex to another vertex that we haven't visited before, the edge that we take is called a *forward edge*
- ▶ When we go from a vertex to another vertex that we've already visited before, the edge that we take is called a *backward edge*
- ▶ To be more specific: the forward edges form a tree
- ▶ *see example*

# Depth-first search tree

- ▶ This tree of forward edges, along with the backward edges, can be analyzed to get a lot of information about the original graph
- ▶ For example: a backward edge represents a cycle in the original graph
- ▶ If there are no backward edges, then there are no cycles in the original graph (i.e. the graph is acyclic)

# Analyzing the DFS tree

- ▶ Let's take a closer look at the depth-first search tree
- ▶ First, let's number each of the vertices in the order that we visit them in the depth-first search
- ▶ For each vertex, we want to know the smallest number of a vertex that we visited when exploring the subtree rooted at the current vertex
- ▶ Why? We'll see in a bit..
- ▶ *see example*



# Analyzing the DFS tree

```
const int n = 1000;
vector<int> adj[n];
vector<int> low(n), num(n, -1);
int curnum = 0;

void analyze(int u, int p) {
    low[u] = num[u] = curnum++;
    for (int i = 0; i < adj[u].size(); i++) {
        int v = adj[u][i];
        if (v == p) continue;
        if (num[v] == -1) {
            analyze(v, u);
            low[u] = min(low[u], low[v]);
        } else {
            low[u] = min(low[u], num[v]);
        }
    }
}

for (int u = 0; u < n; u++) {
    if (num[u] == -1) {
        analyze(u, -1);
    }
}
```

# Analyzing the DFS tree

- ▶ Time complexity of this is just  $O(n + m)$ , since this is basically just one depth-first search
- ▶ Now, as promised, let's see some applications of this

# Bridges

- ▶ We have an undirected graph
- ▶ Without loss of generality, assume it is connected (i.e. one big connected component)
- ▶ Find an edge, so that if you remove that edge the graph is no longer connected
- ▶ Naive algorithm: Try removing edges, one at a time, and finding the connected components of the resulting graph
- ▶ That's pretty inefficient,  $O(m(n + m))$

# Bridges

- ▶ Let's take a look at the values that we computed in the DFS tree
- ▶ We see that a forward edge  $(u, v)$  is a bridge if and only if  $\text{low}[v] > \text{num}[u]$
- ▶ Simple to extend our analyze function to return all bridges
- ▶ Again, this is just  $O(n + m)$

# Bridges

```
const int n = 1000;
vector<int> adj[n];
vector<int> low(n), num(n, -1);
int curnum = 0;

vector<pair<int, int> > bridges;

void find_bridges(int u, int p) {
    low[u] = num[u] = curnum++;
    for (int i = 0; i < adj[u].size(); i++) {
        int v = adj[u][i];
        if (v == p) continue;
        if (num[v] == -1) {
            find_bridges(v, u);
            low[u] = min(low[u], low[v]);
        } else {
            low[u] = min(low[u], num[v]);
        }

        if (low[v] > num[u]) {
            bridges.push_back(make_pair(u, v));
        }
    }
}

for (int u = 0; u < n; u++) {
    if (num[u] == -1) {
        find_bridges(u, -1);
    }
}
```

# Strongly connected components

- ▶ We know how to find connected components in undirected graphs
- ▶ But what about directed graphs?
- ▶ Such components behave a bit differently in directed graphs, especially since if  $v$  is reachable from  $u$ , it doesn't mean that  $u$  is reachable from  $v$
- ▶ The definition remains the same, though
- ▶ A strongly connected component is a maximal subset of the vertices such that each pair of vertices is reachable from each other

# Strongly connected components

- ▶ The connected components algorithm won't work here
- ▶ Instead we can use the depth-first search tree of the graph to find these components
- ▶ *see example*

# Strongly connected components

```
vector<int> adj[100];  
vector<int> low(100), num(100, -1);  
vector<bool> incomp(100, false);  
int curnum = 0;
```

```
stack<int> comp;
```

```
void scc(int u) {
```

```
    // scc code...
```

```
}
```

```
for (int i = 0; i < n; i++) {
```

```
    if (num[i] == -1) {
```

```
        scc(i);
```

```
    }
```

```
}
```



# Strongly connected components

```
void scc(int u) {  
  
    comp.push(u);  
    incomp[u] = true;  
  
    low[u] = num[u] = curnum++;  
    for (int i = 0; i < adj[u].size(); i++) {  
        int v = adj[u][i];  
        if (num[v] == -1) {  
            scc(v);  
            low[u] = min(low[u], low[v]);  
        } else if (incomp[v]) {  
            low[u] = min(low[u], num[v]);  
        }  
    }  
  
    if (num[u] == low[u]) {  
        printf("comp: ");  
        while (true) {  
            int cur = comp.top();  
            comp.pop();  
            incomp[cur] = false;  
            printf("%d, ", cur);  
            if (cur == u) {  
                break;  
            }  
        }  
  
        printf("\n");  
    }  
}
```

# Strongly connected components

- ▶ Time complexity?
- ▶ Basically just the DFS analyze function (which was  $O(n + m)$ ), with one additional loop to construct the component
- ▶ But each vertex is only in one component...
- ▶ Time complexity still just  $O(n + m)$

# Example problem: Come and Go

- ▶ <http://uva.onlinejudge.org/external/118/11838.html>

# Topological sort

- ▶ We have  $n$  tasks
- ▶ Each task  $i$  has a list of tasks that must be finished before we can start task  $i$
- ▶ Find an order in which we can process the tasks
- ▶ Can be represented as a directed graph
  - Each task is a vertex in the graph
  - If task  $j$  should be finished before task  $i$ , then we add a directed edge from vertex  $i$  to vertex  $j$
- ▶ Notice that this can't be solved if the graph contains a cycle
- ▶ A modified depth-first search can be used to find an ordering in  $O(n + m)$  time, or determine that one does not exist

# Topological sort

```
vector<int> adj[1000];
vector<bool> visited(1000, false);
vector<int> order;

void topsort(int u) {
    if (visited[u]) {
        return;
    }

    visited[u] = true;
    for (int i = 0; i < adj[u].size(); i++) {
        int v = adj[u][i];
        topsort(v);
    }

    order.push_back(u);
}

for (int u = 0; u < n; u++) {
    topsort(u);
}
```

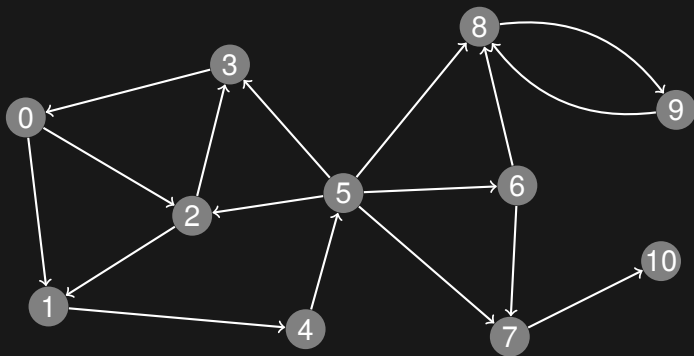
# Example problem: Ordering Tasks

- ▶ <http://uva.onlinejudge.org/external/103/10305.html>

# Breadth-first search

- ▶ There's another search algorithm called Breadth-first search
- ▶ Only difference is the order in which it visits the vertices
- ▶ It goes in order of increasing distance from the source vertex

# Breadth-first search

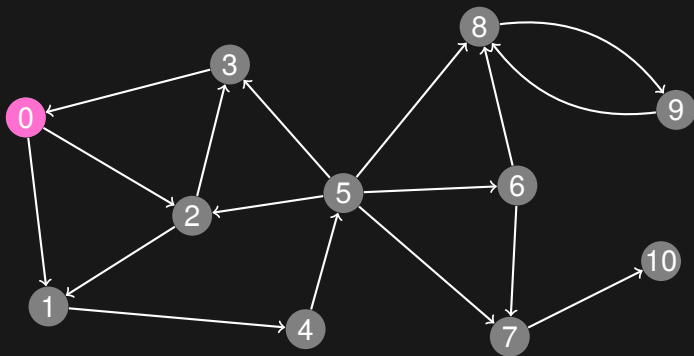


Queue:

	0	1	2	3	4	5	6	7	8	9	10
marked	0	0	0	0	0	0	0	0	0	0	0



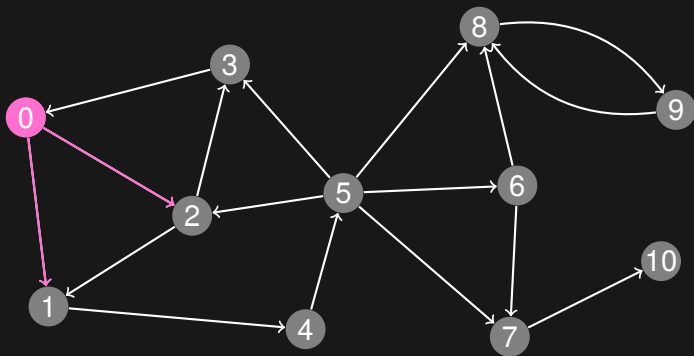
# Breadth-first search



Queue: 0

	0	1	2	3	4	5	6	7	8	9	10
marked	1	0	0	0	0	0	0	0	0	0	0

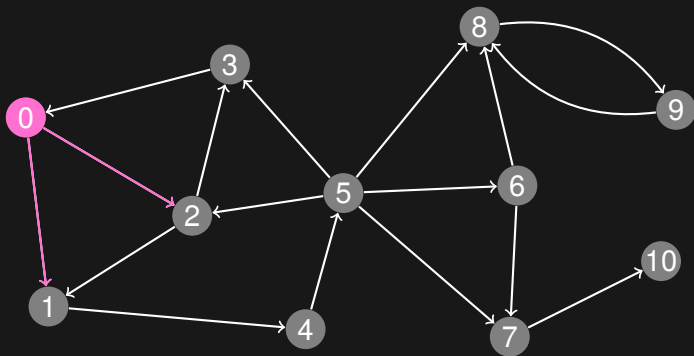
# Breadth-first search



Queue: 0

	0	1	2	3	4	5	6	7	8	9	10
marked	1	0	0	0	0	0	0	0	0	0	0

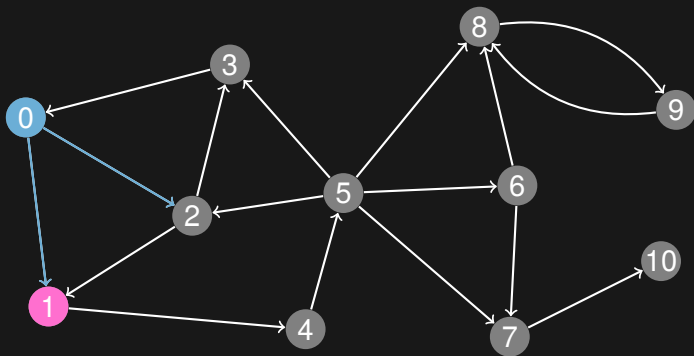
# Breadth-first search



Queue:    0 1 2

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	0	0	0	0	0	0	0

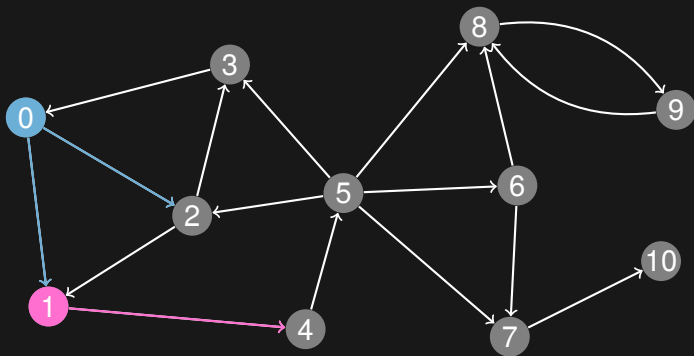
# Breadth-first search



Queue:    1 2

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	0	0	0	0	0	0	0

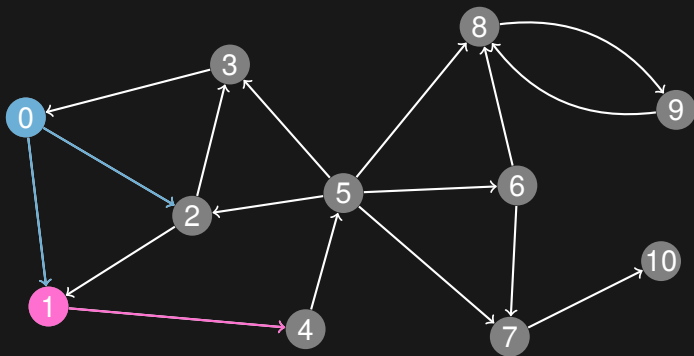
# Breadth-first search



Queue:    1 2

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	0	0	0	0	0	0	0

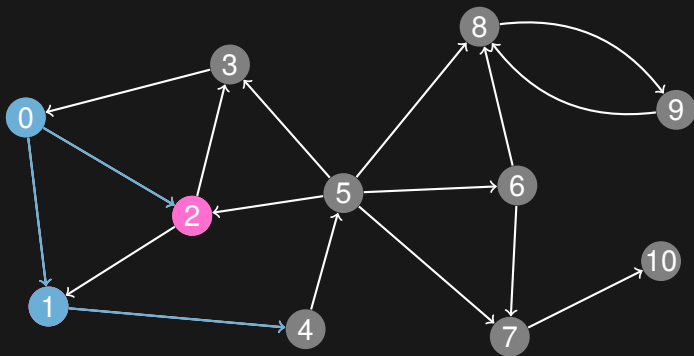
# Breadth-first search



Queue:    1 2 4

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	1	0	0	0	0	0	0

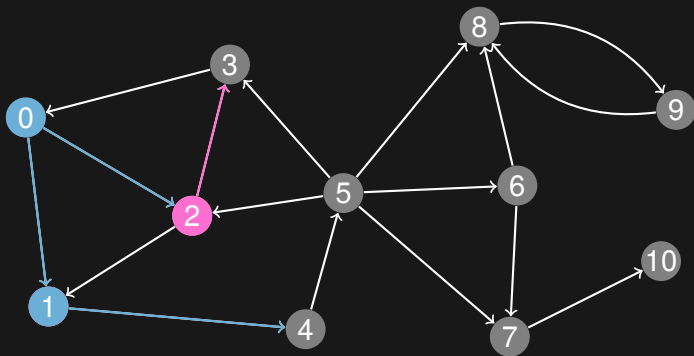
# Breadth-first search



Queue:    2 4

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	1	0	0	0	0	0	0

# Breadth-first search

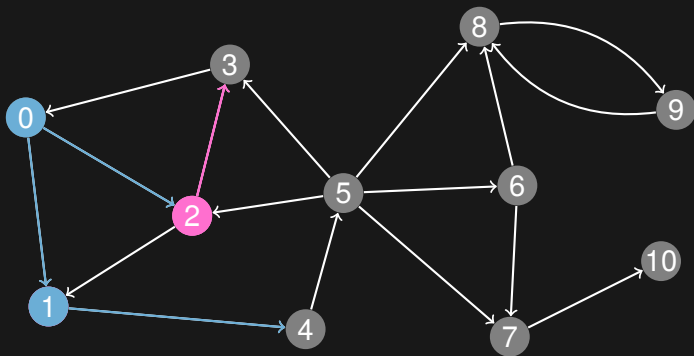


Queue:    2 4

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	0	1	0	0	0	0	0	0



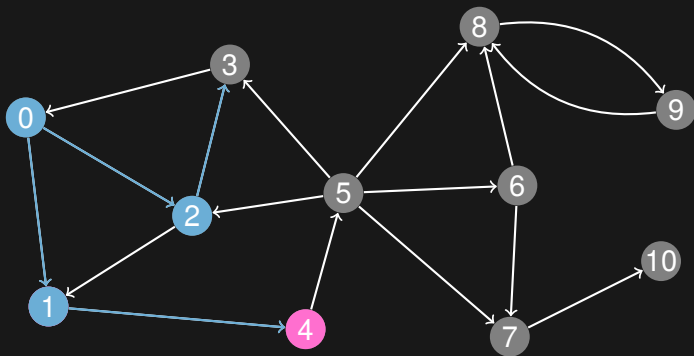
# Breadth-first search



Queue:    2 4 3

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	0	0	0	0	0	0

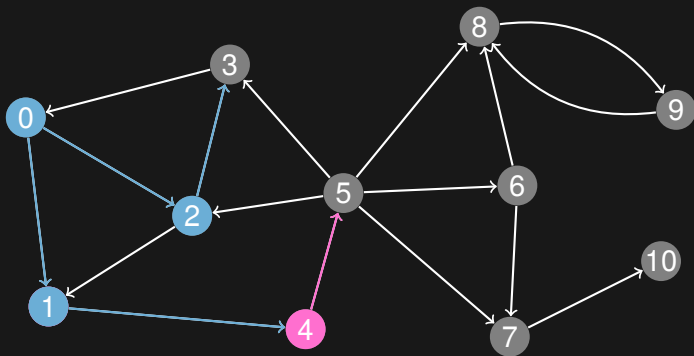
# Breadth-first search



Queue: 4 3

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	0	0	0	0	0	0

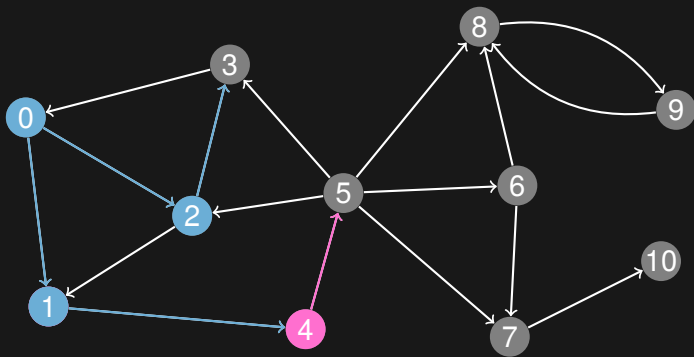
# Breadth-first search



Queue: 4 3

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	0	0	0	0	0	0

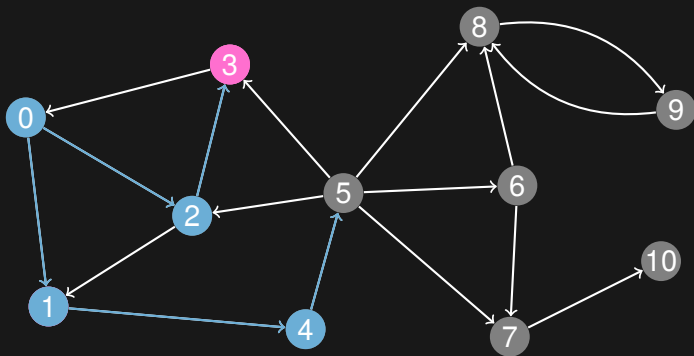
# Breadth-first search



Queue: 4 3 5

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	0	0	0	0	0

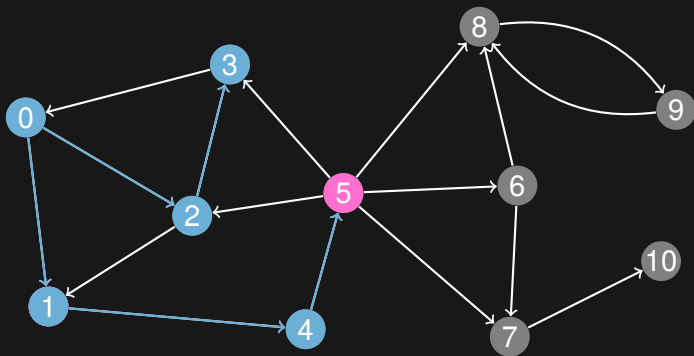
# Breadth-first search



Queue:    3 5

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	0	0	0	0	0

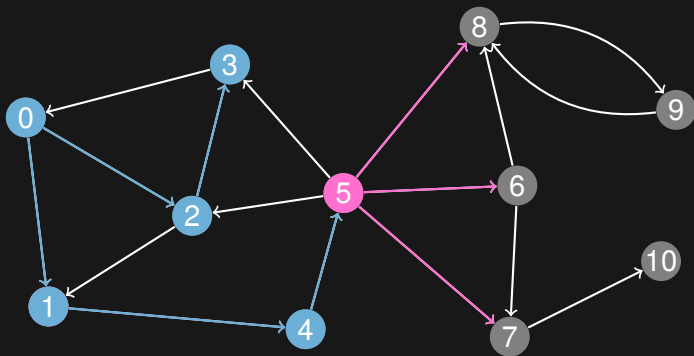
# Breadth-first search



Queue: 5

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	0	0	0	0	0

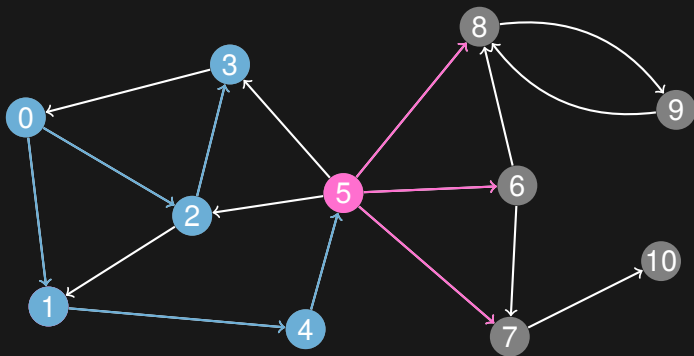
# Breadth-first search



Queue: 5

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	0	0	0	0	0

# Breadth-first search

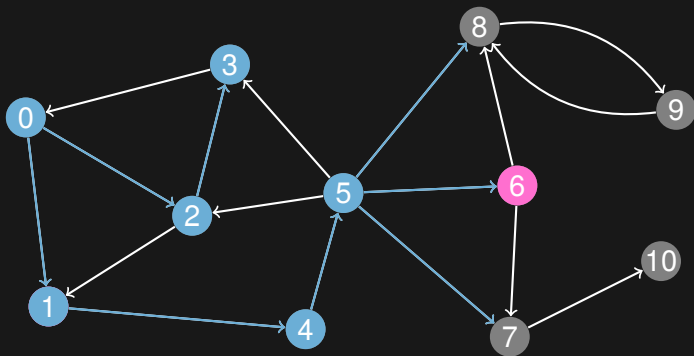


Queue:    5 6 7 8

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	0



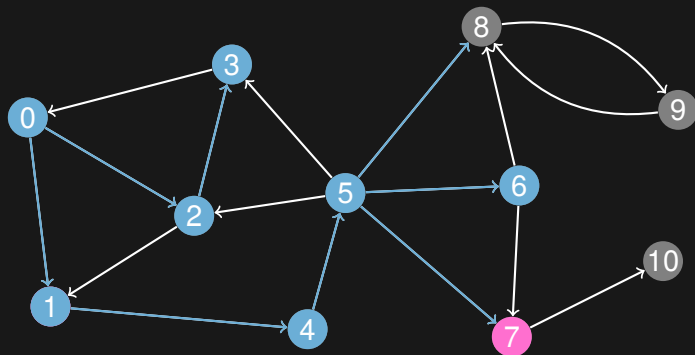
# Breadth-first search



Queue:    6 7 8

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	0

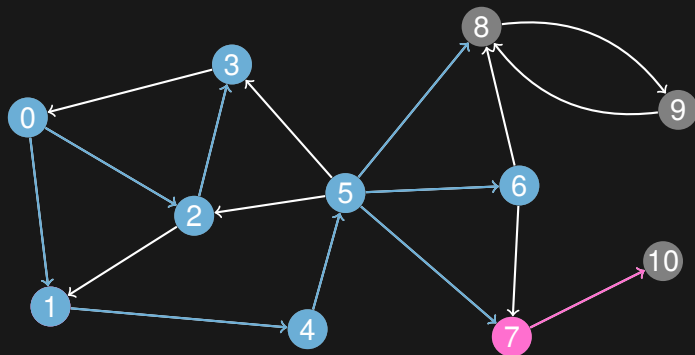
# Breadth-first search



Queue: 7 8

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	0

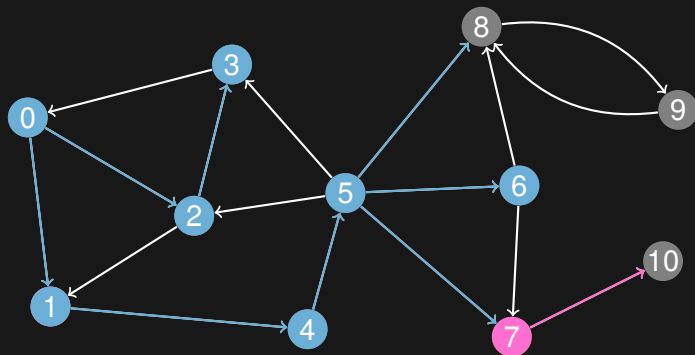
# Breadth-first search



Queue: 7 8

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	0

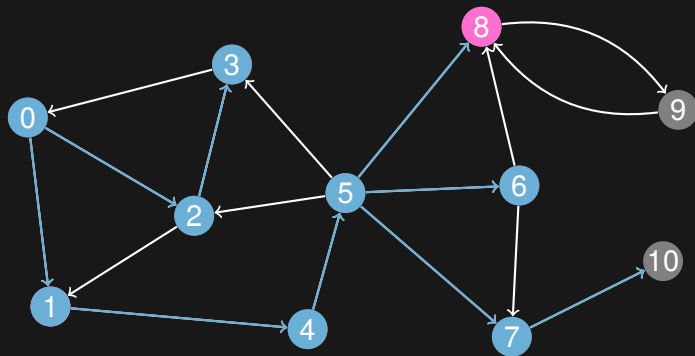
# Breadth-first search



Queue: 7 8 10

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	1

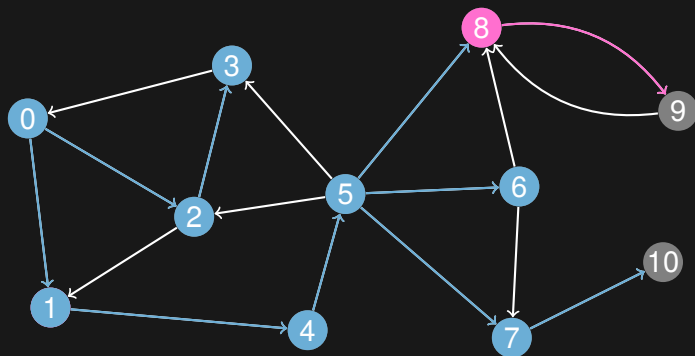
# Breadth-first search



Queue:    8 10

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	1

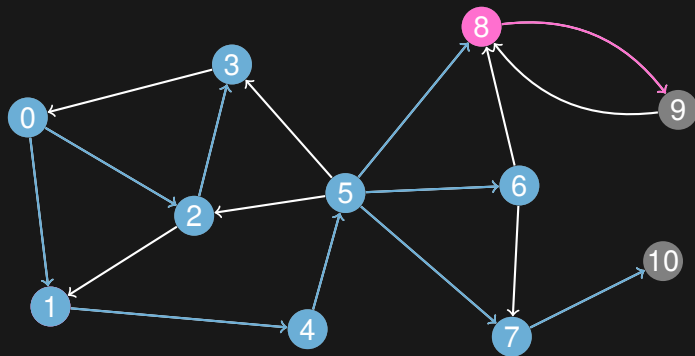
# Breadth-first search



Queue:    8 10

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	0	1

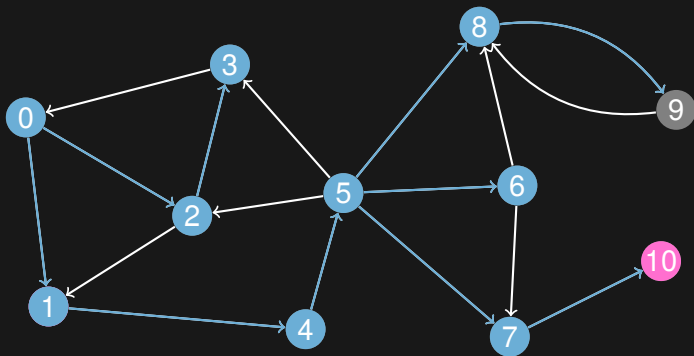
# Breadth-first search



Queue: 8 10 9

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	1

# Breadth-first search

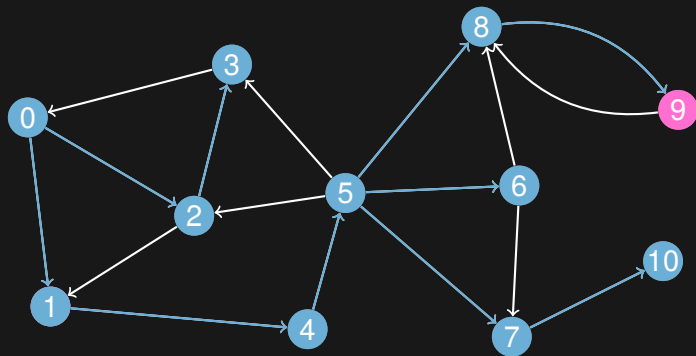


Queue:    10 9

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	1



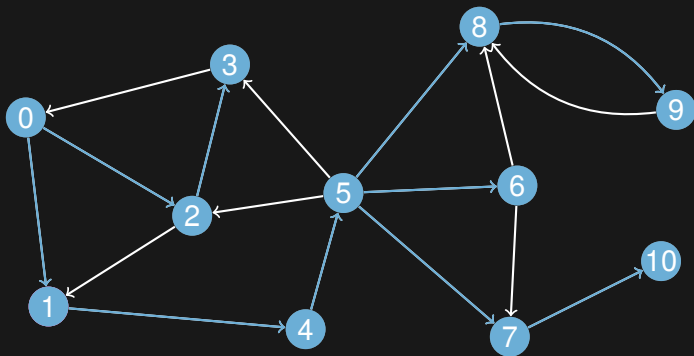
# Breadth-first search



Queue: 9

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	1

# Breadth-first search



Queue:

	0	1	2	3	4	5	6	7	8	9	10
marked	1	1	1	1	1	1	1	1	1	1	1

# Breadth-first search

```
vector<int> adj[1000];
vector<bool> visited(1000, false);

queue<int> Q;
Q.push(start);
visited[start] = true;

while (!Q.empty()) {
    int u = Q.front(); Q.pop();

    for (int i = 0; i < adj[u].size(); i++) {
        int v = adj[u][i];
        if (!visited[v]) {
            Q.push(v);
            visited[v] = true;
        }
    }
}
```

# Shortest path in unweighted graphs

- ▶ We have an unweighted graph, and want to find the shortest path from  $A$  to  $B$
- ▶ That is, we want to find a path from  $A$  to  $B$  with the minimum number of edges
- ▶ Breadth-first search goes through the vertices in increasing order of distance from the start vertex
- ▶ Just do a single breadth-first search from  $A$ , until we find  $B$
- ▶ Or let the search continue through the whole graph, and then we have the shortest paths from  $A$  to all other vertices
- ▶ Shortest path from  $A$  to all other vertices:  $O(n + m)$

# Shortest path in unweighted graphs

```
vector<int> adj[1000];  
vector<bool> dist(1000, -1);  
  
queue<int> Q;  
Q.push(A);  
dist[A] = 0;  
  
while (!Q.empty()) {  
    int u = Q.front(); Q.pop();  
  
    for (int i = 0; i < adj[u].size(); i++) {  
        int v = adj[u][i];  
        if (dist[v] == -1) {  
            Q.push(v);  
            dist[v] = 1 + dist[u];  
        }  
    }  
}  
  
printf("%d\n", dist[B]);
```