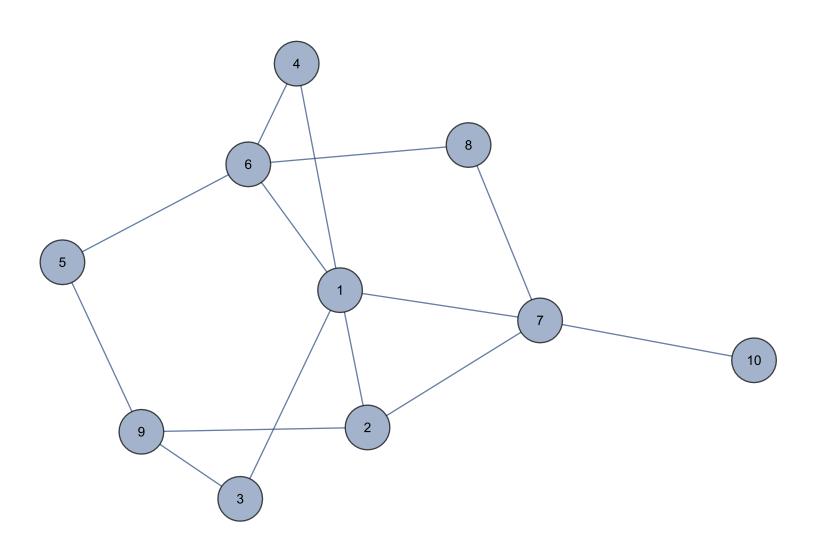
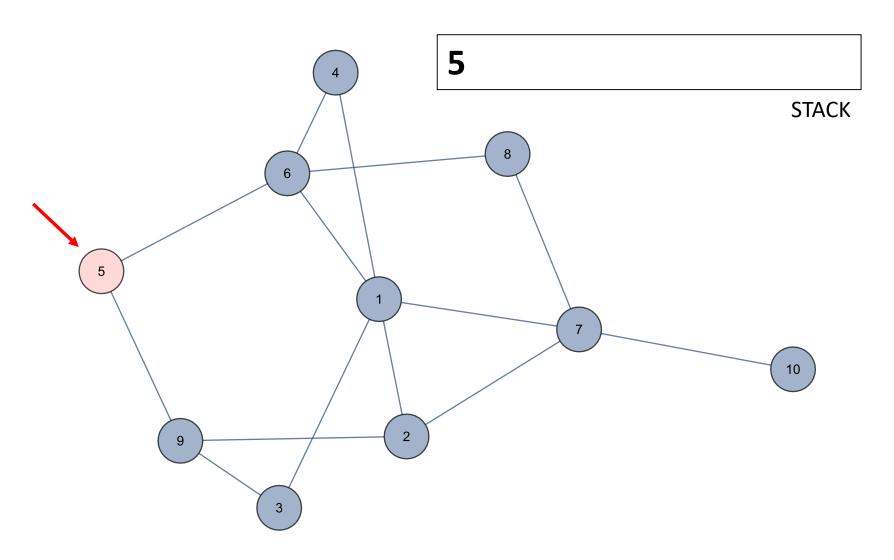


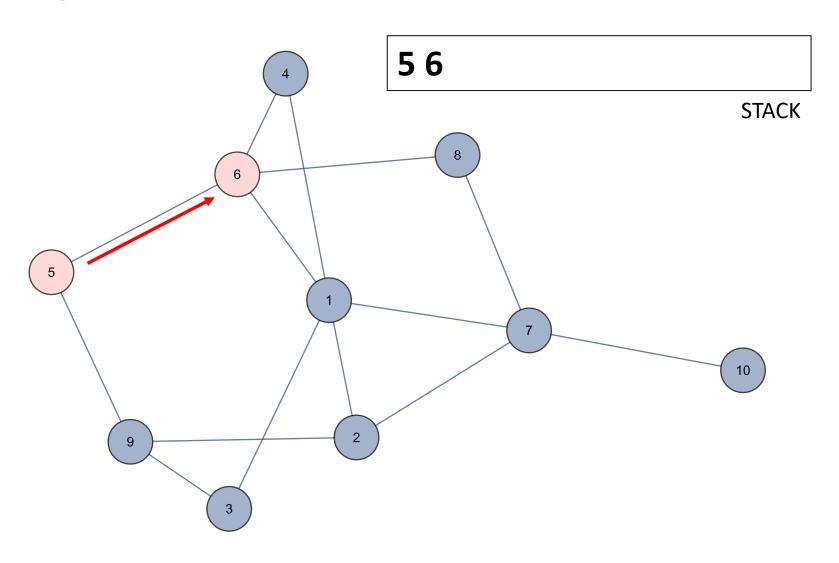
Graph Search Algorithms

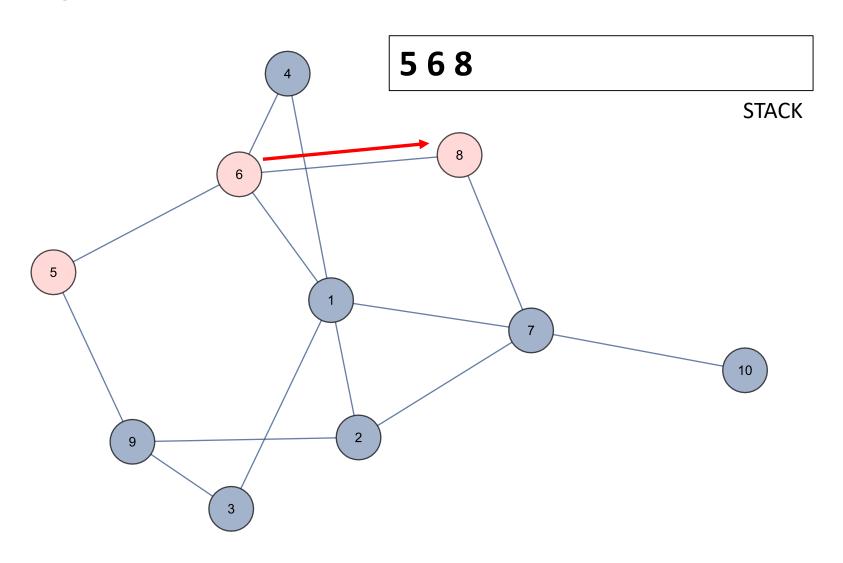
Tang Anke

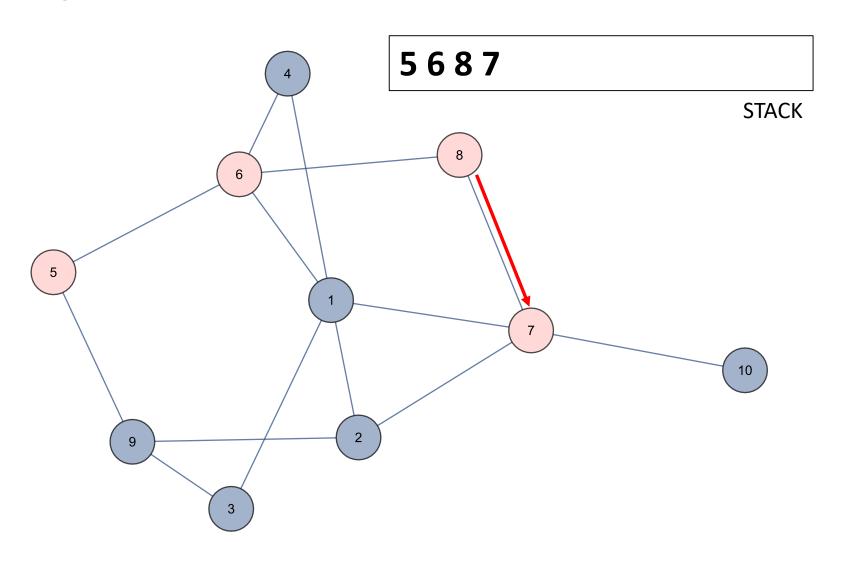
Depth-first search (**DFS**) is an <u>algorithm</u> for traversing or searching <u>tree</u> or <u>graph</u> data structures. The algorithm starts at the <u>root node</u> (selecting some arbitrary node as the root node in the case of a graph) and explores as far as possible along each branch before <u>backtracking</u>. A version of depth-first search was investigated in the 19th century by French mathematician <u>Charles Pierre Trémaux</u> as a strategy for <u>solving</u> mazes.

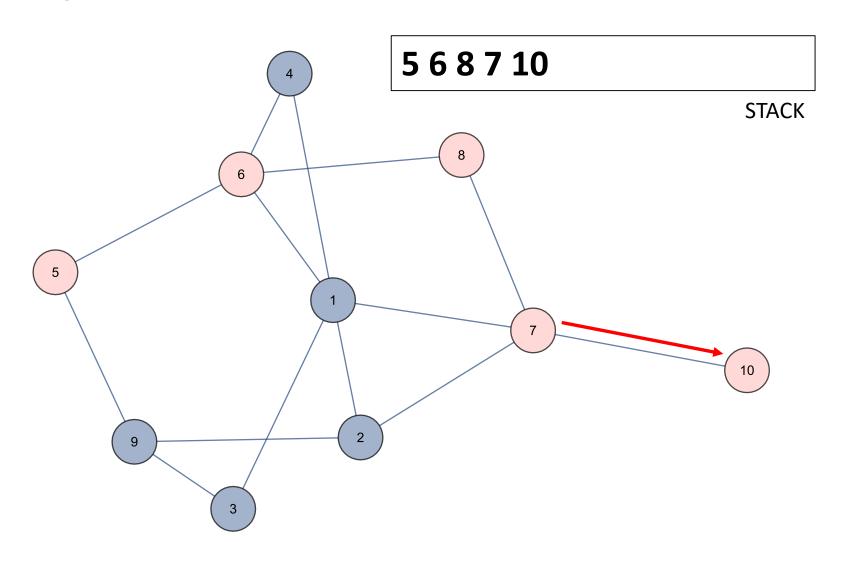


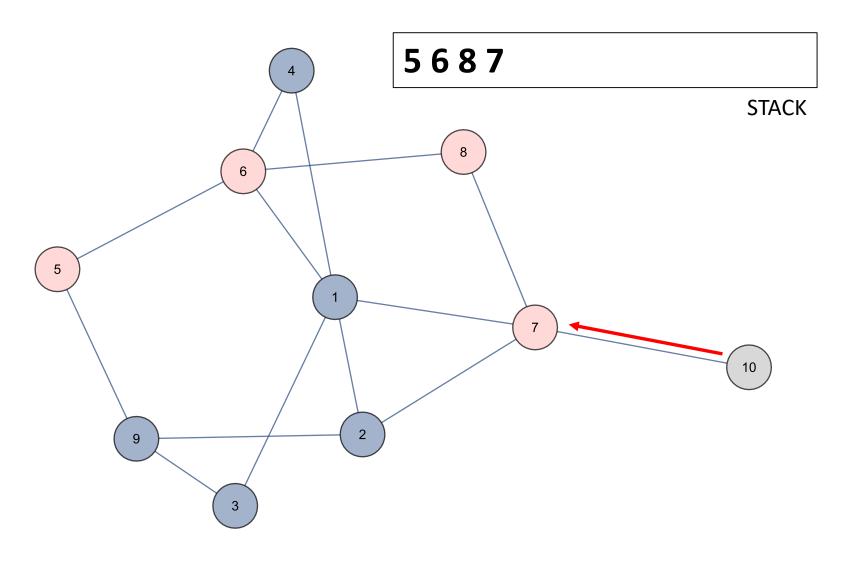


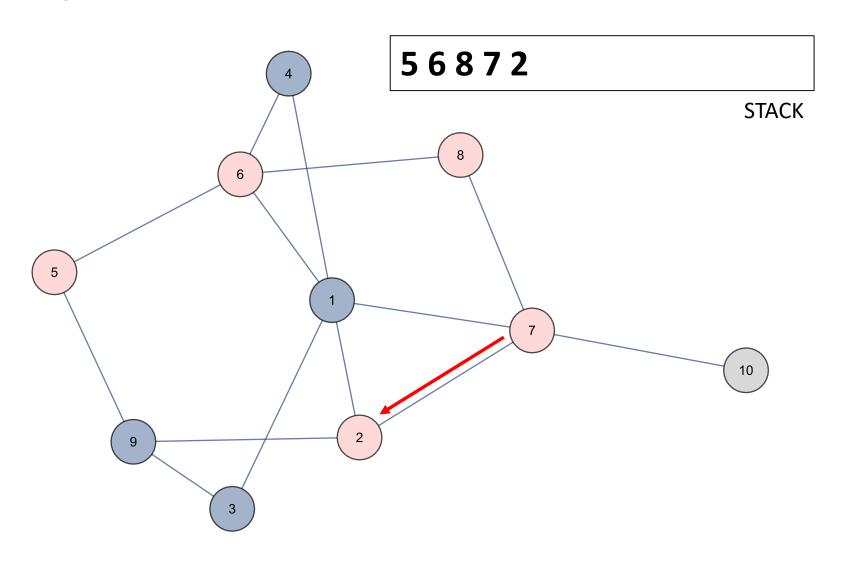


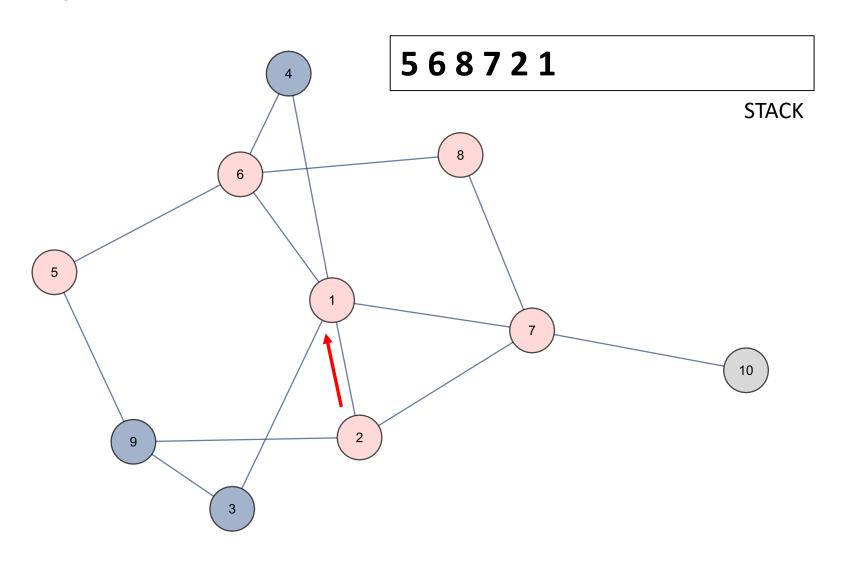


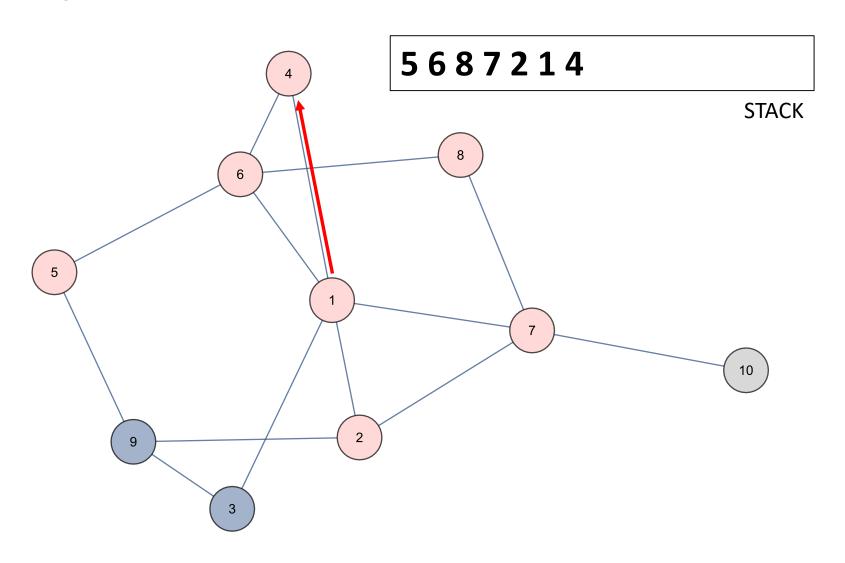


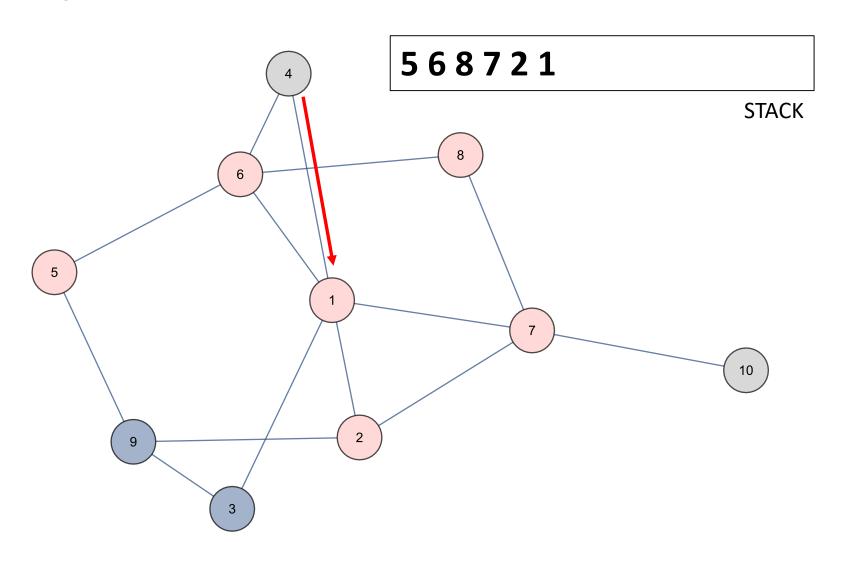


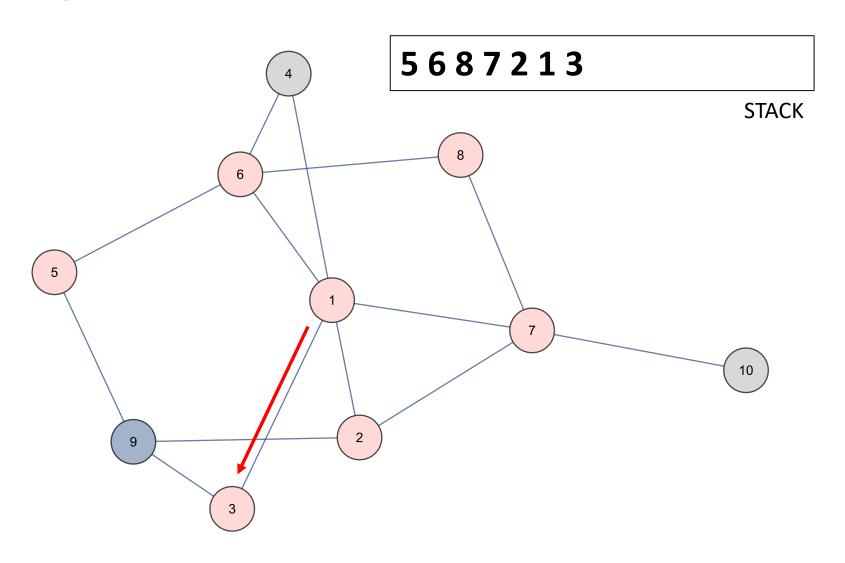


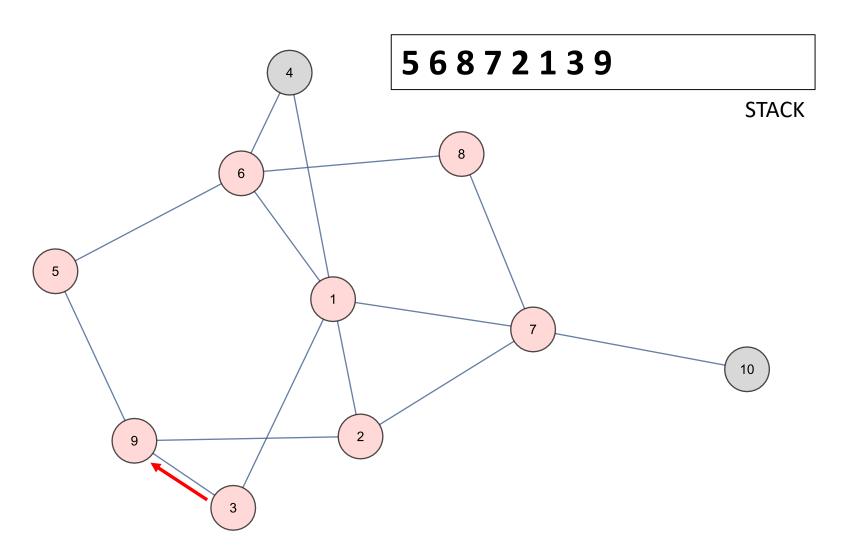


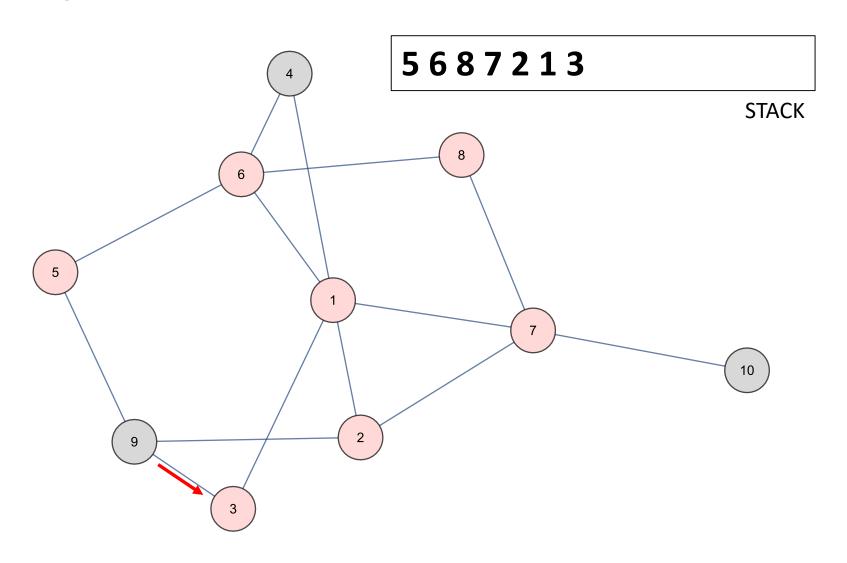


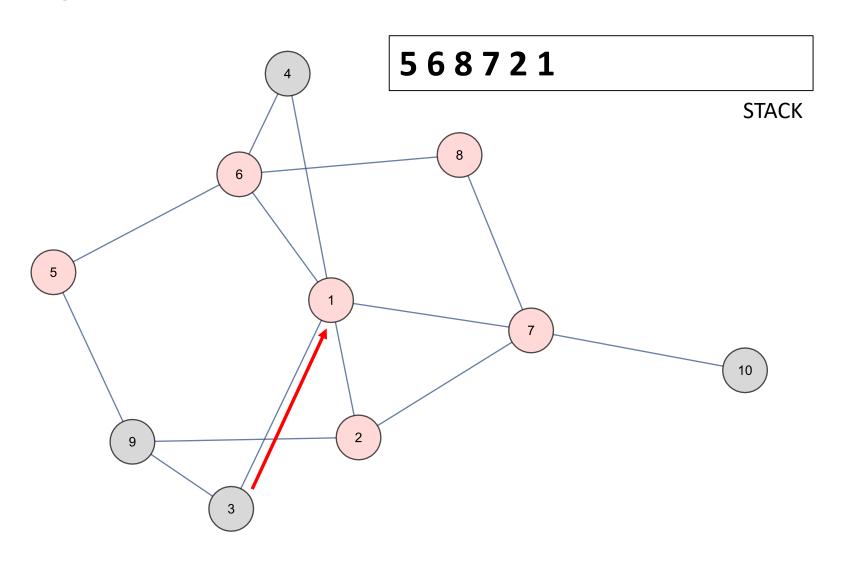


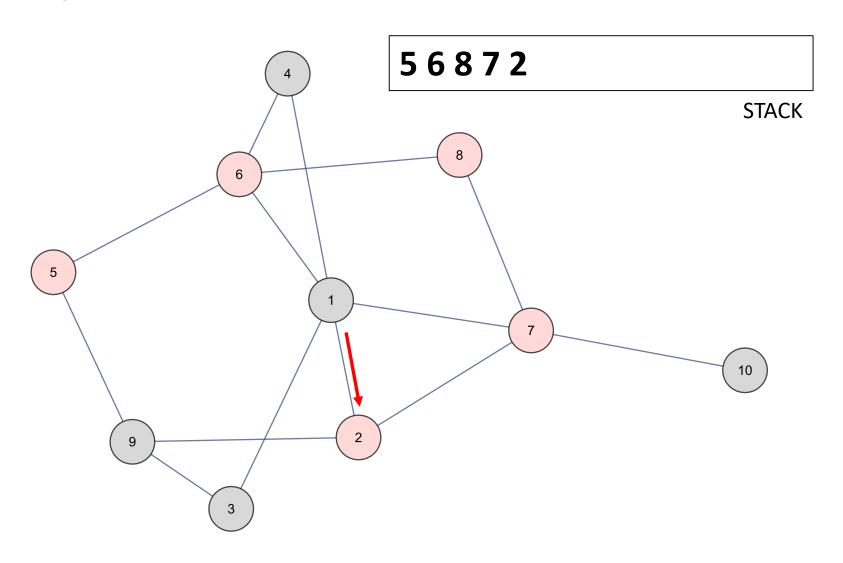


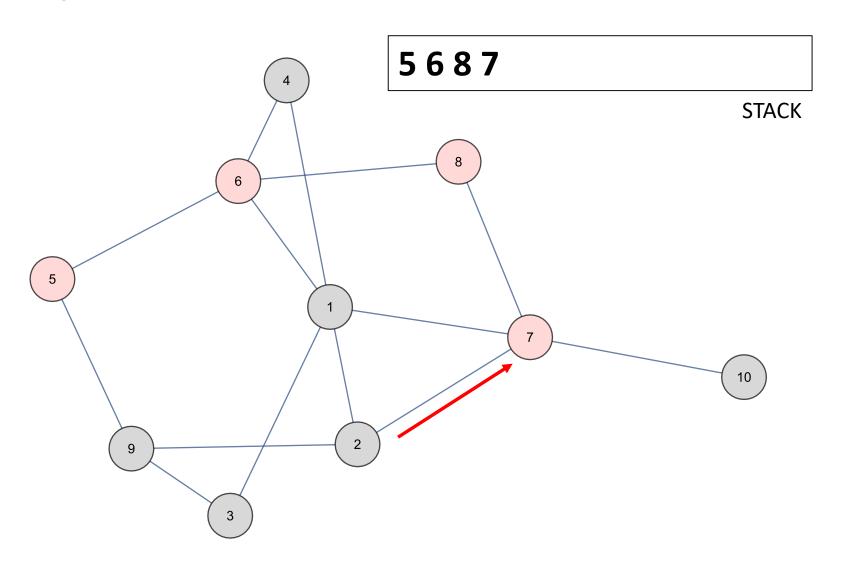


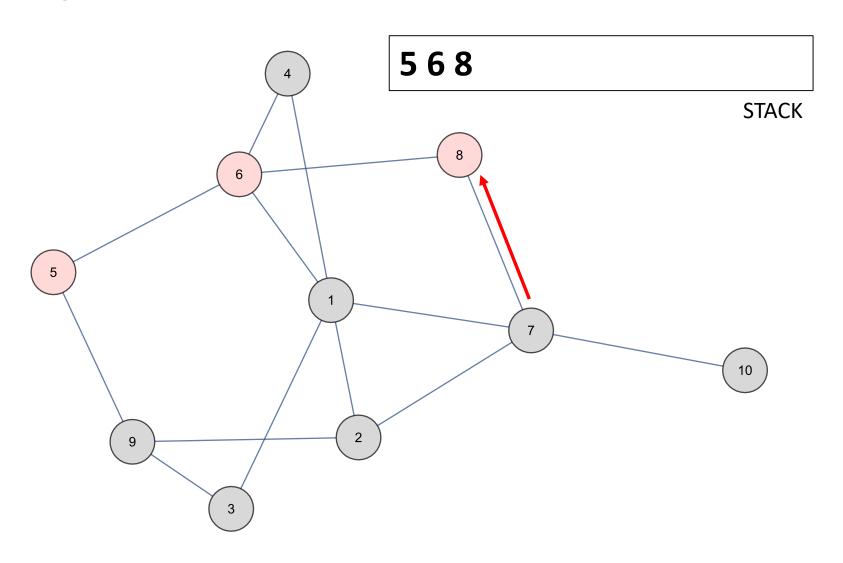


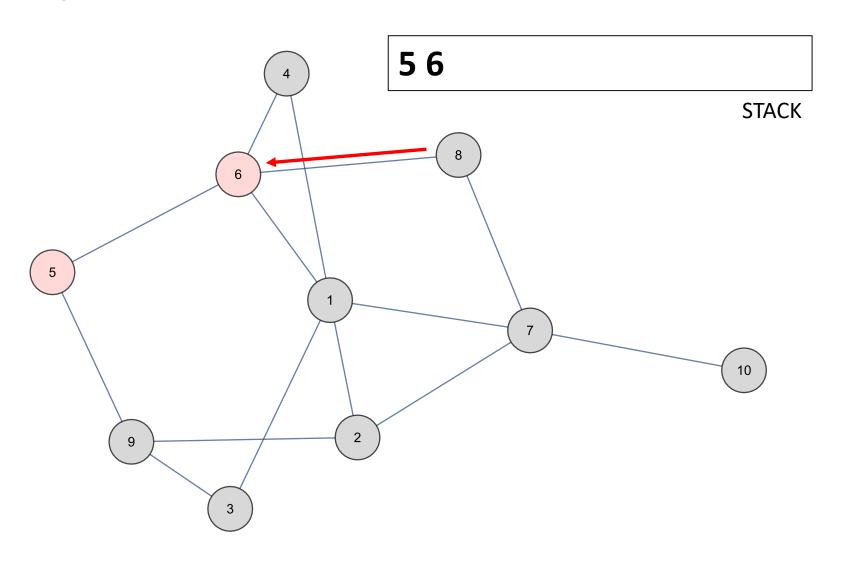


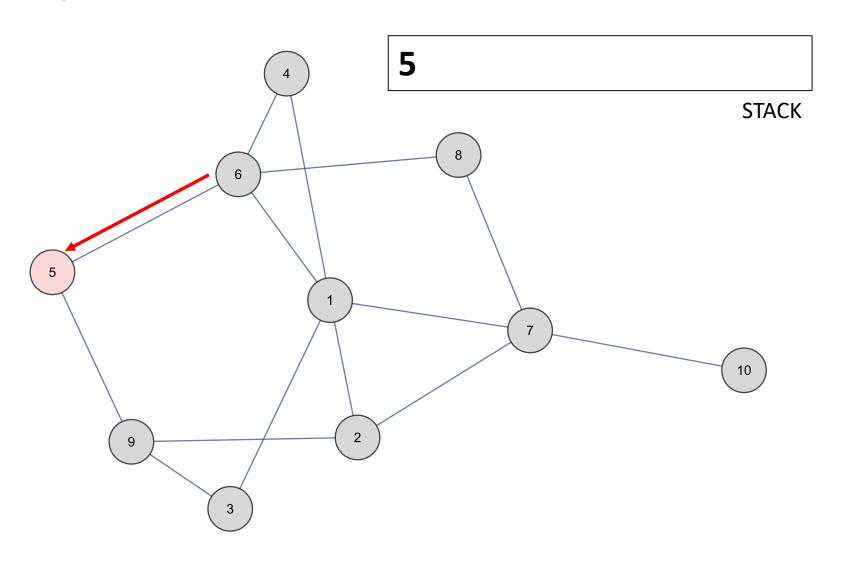


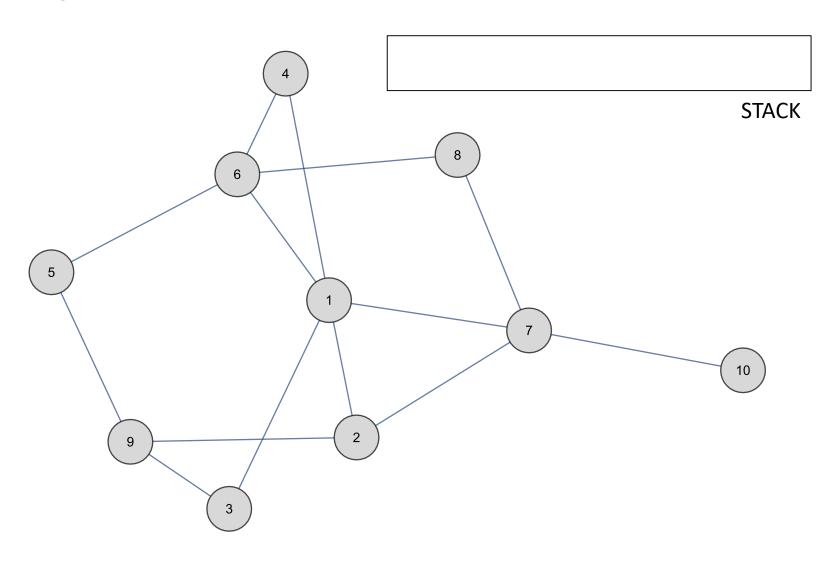






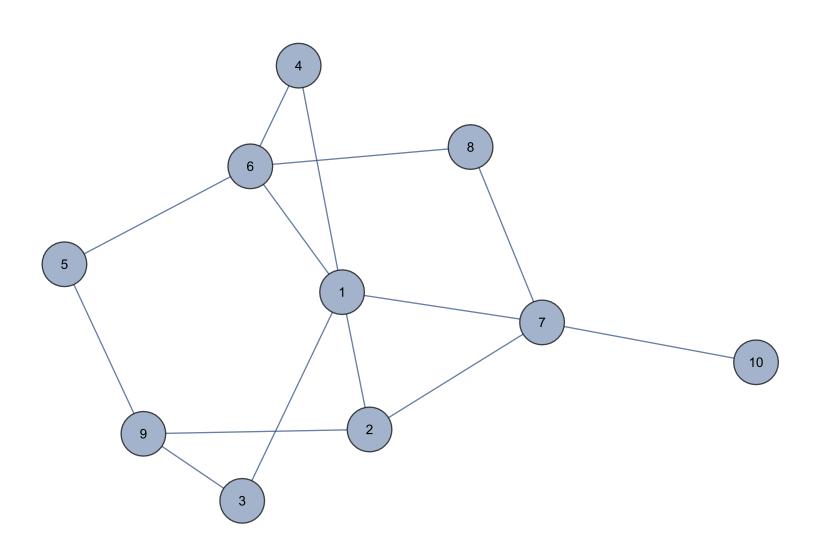


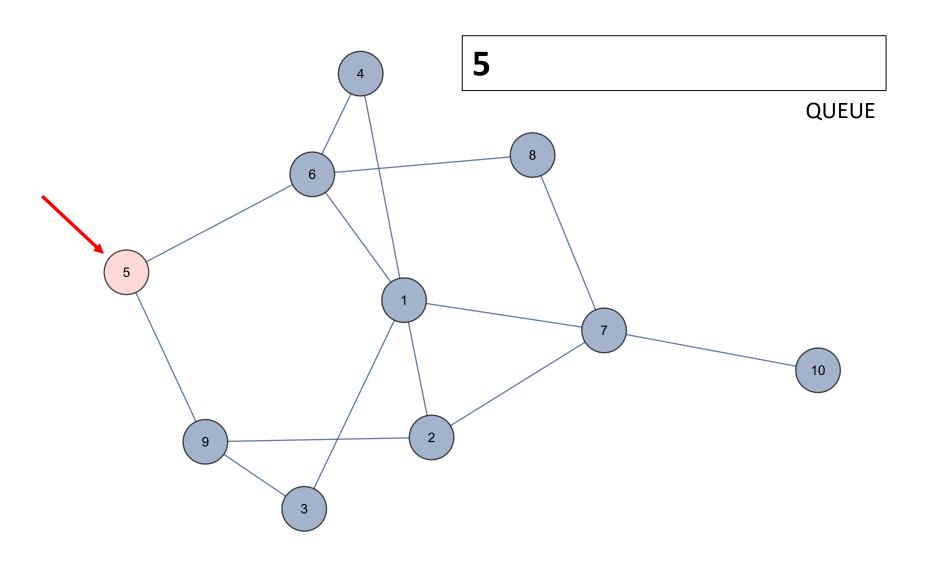


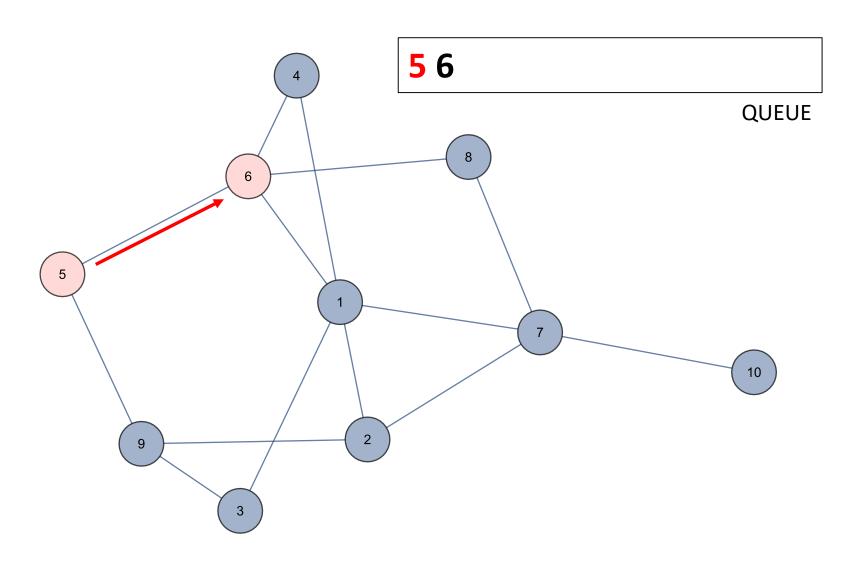


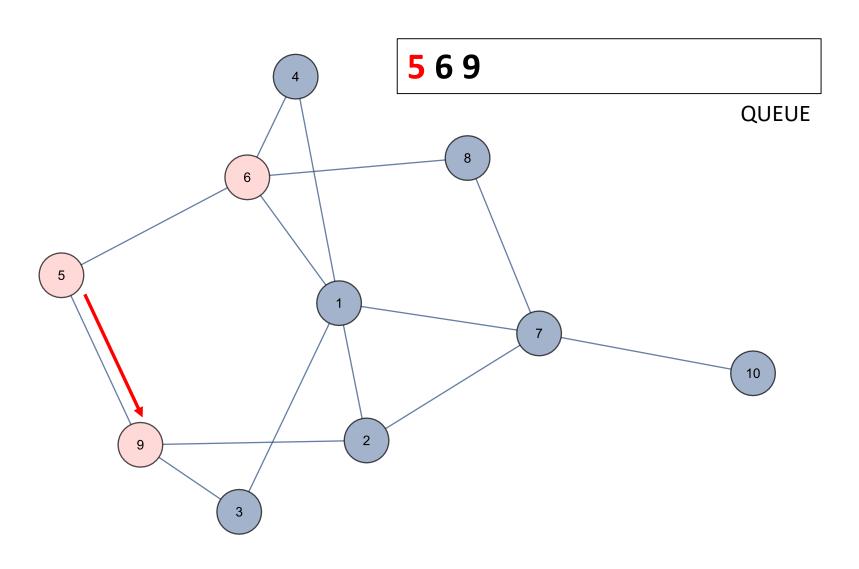
```
# Global or class scope variables
n = number of nodes in the graph
g = adjacency list representing graph
visited = [false, ..., false] # size n
function dfs(at):
  if visited[at]: return
  visited[at] = true
  neighbours = graph[at]
  for next in neighbours:
     dfs(next)
# Start DFS at node zero
start_node = 0
dfs(start_node)
```

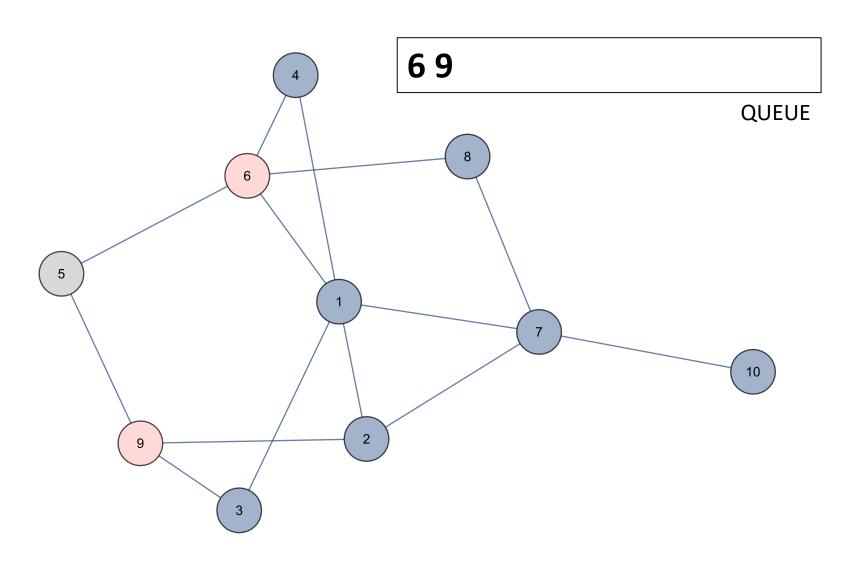
Breadth-first search (BFS) is an <u>algorithm</u> for traversing or searching <u>tree</u> or <u>graph</u> data structures. It starts at the <u>tree root</u> (or some arbitrary node of a graph, sometimes referred to as a 'search key'), and explores all of the neighbor nodes at the present depth prior to moving on to the nodes at the next depth level.

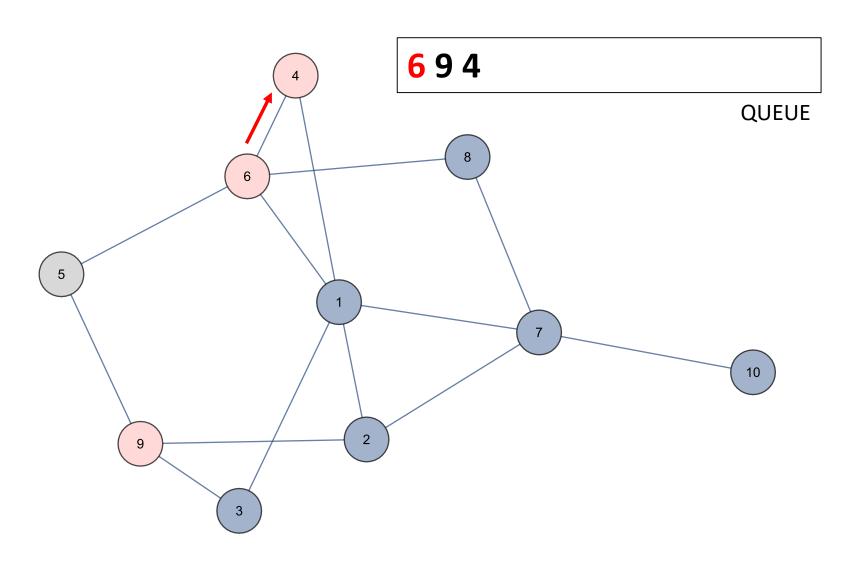


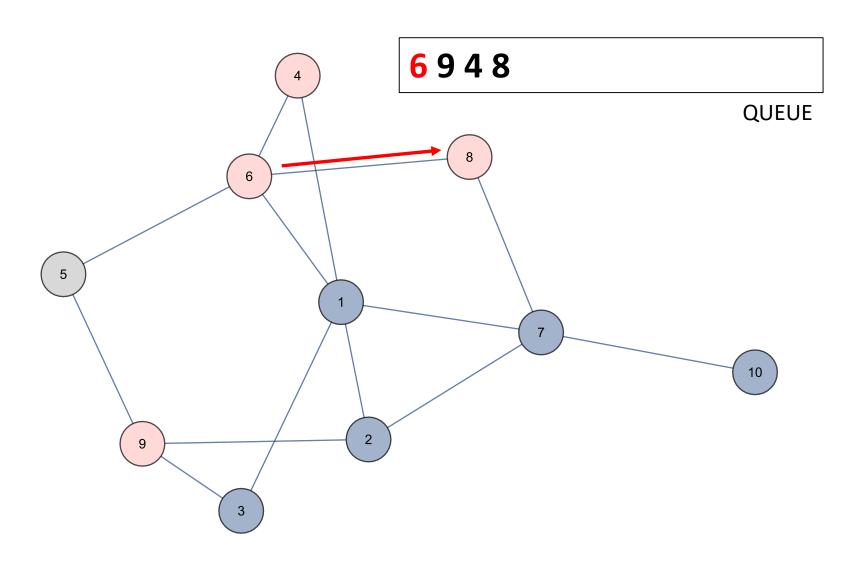


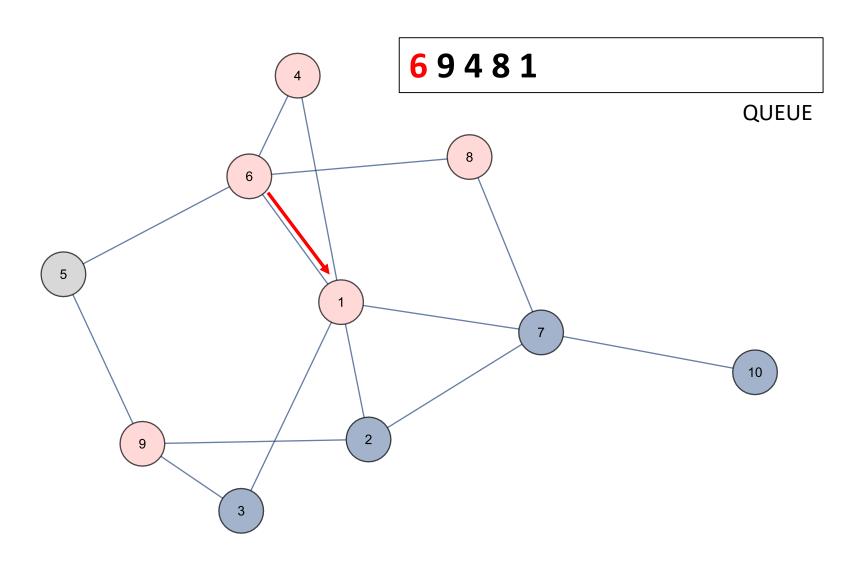


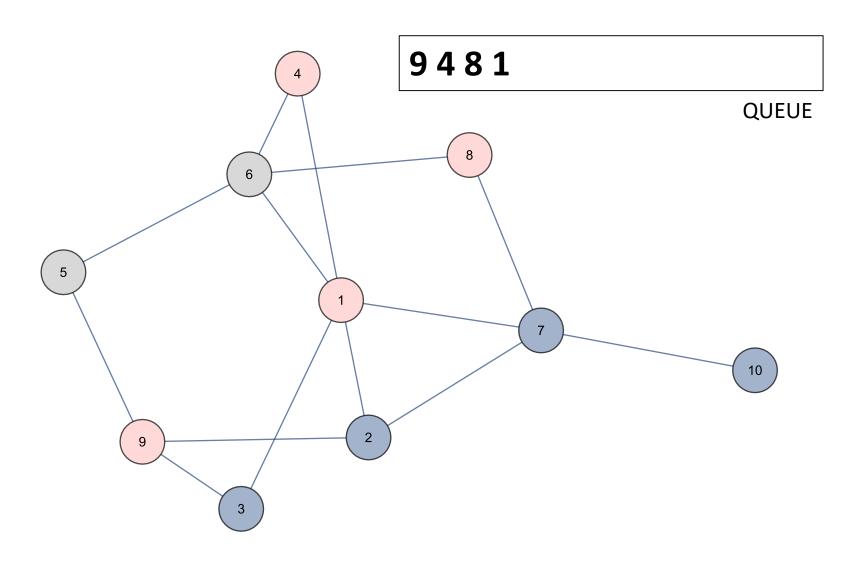


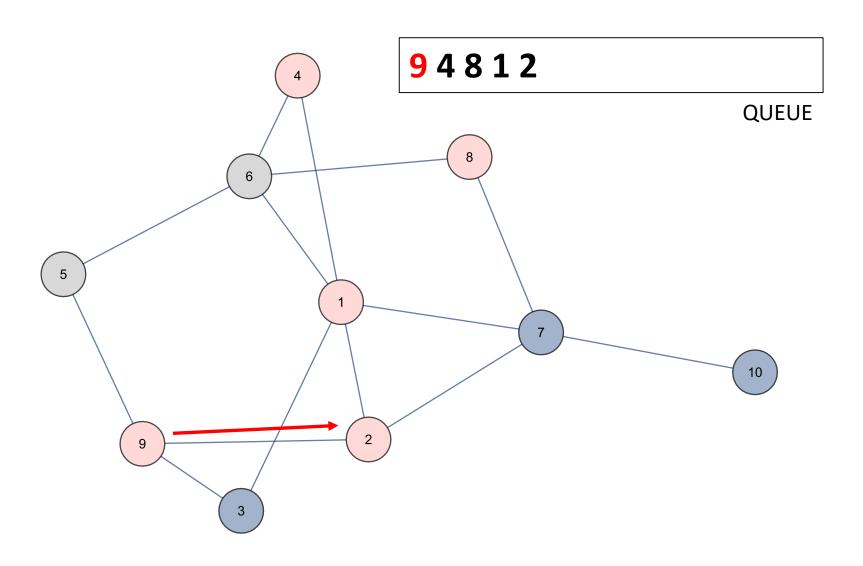


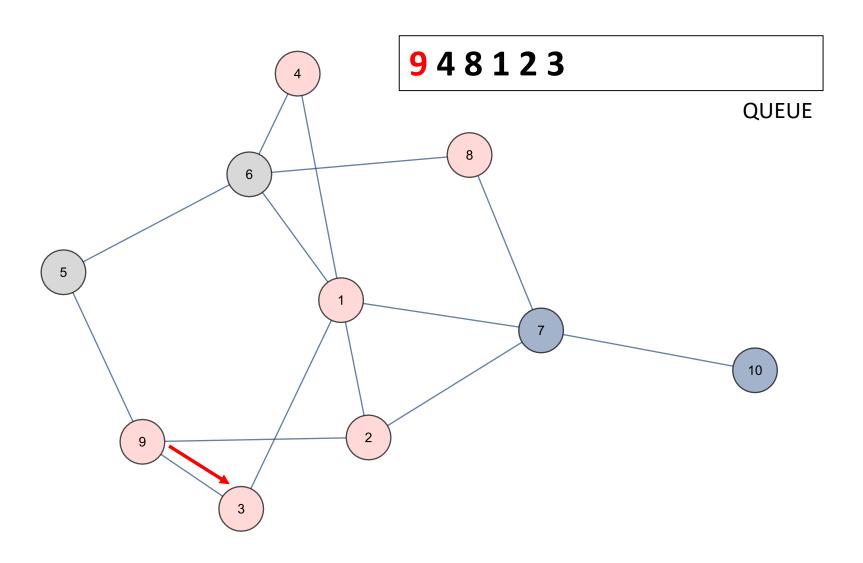


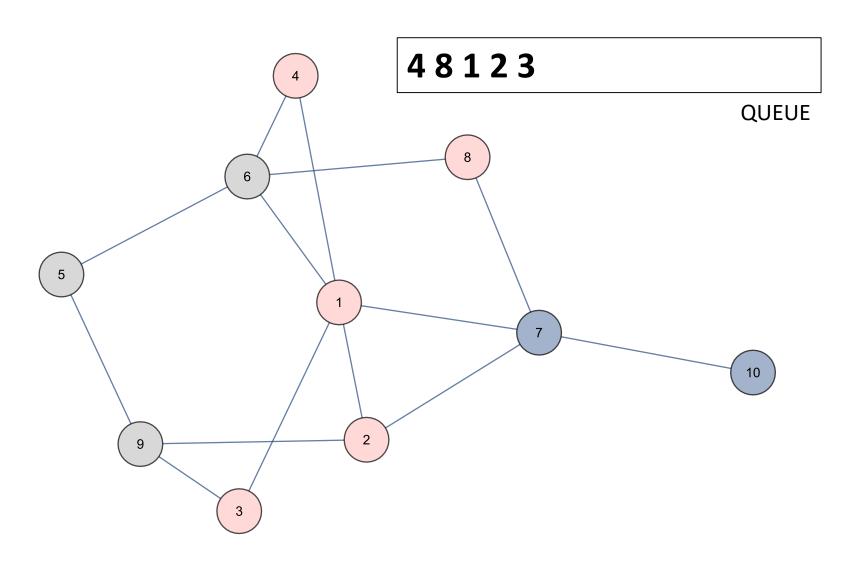


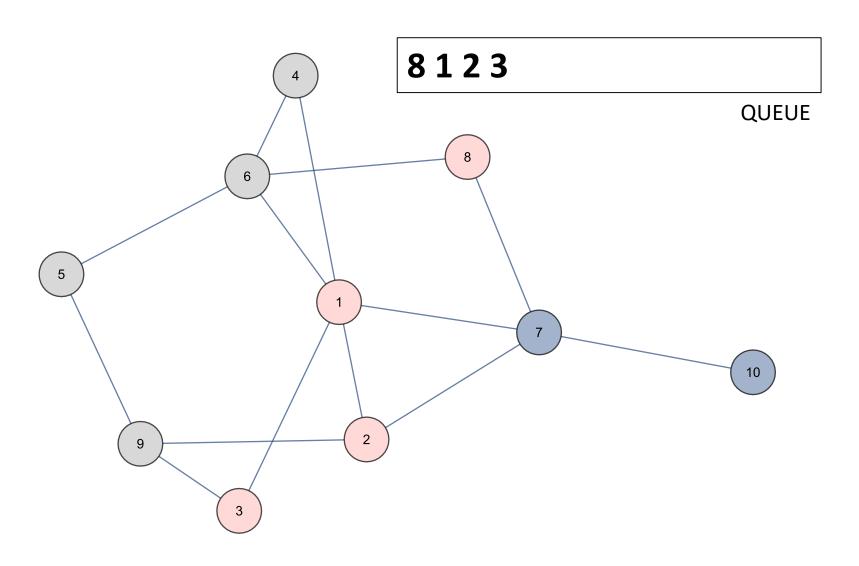


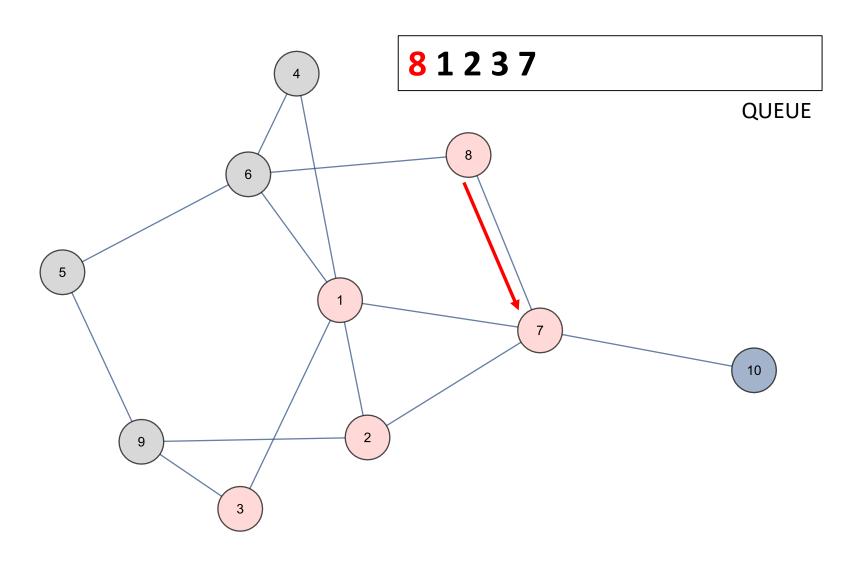


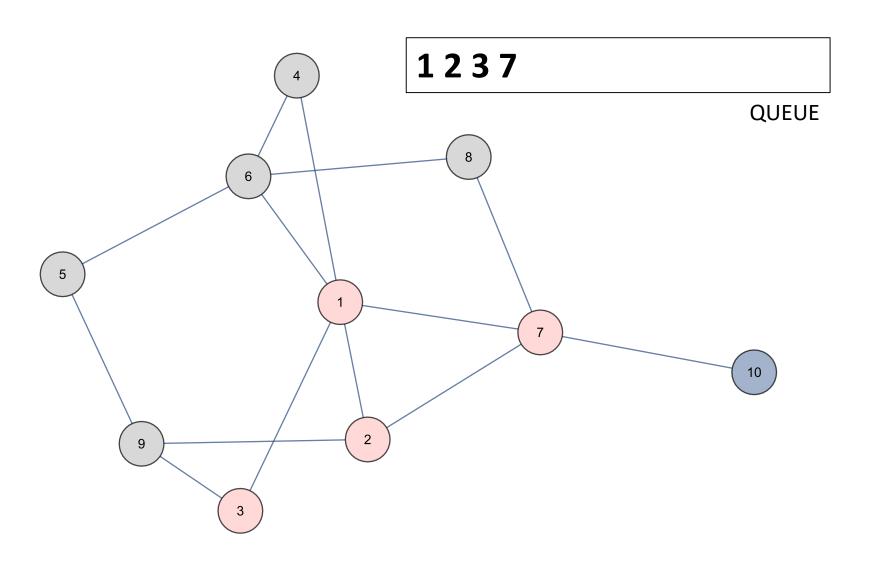


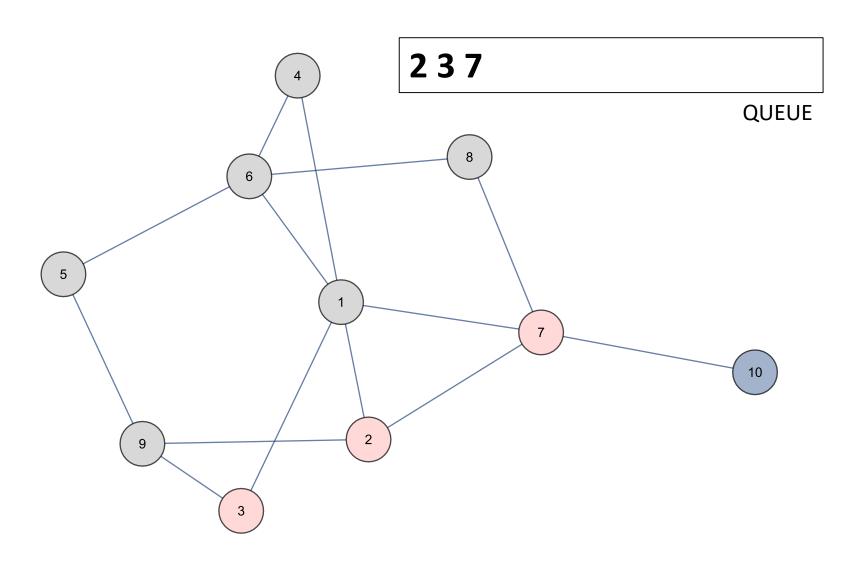


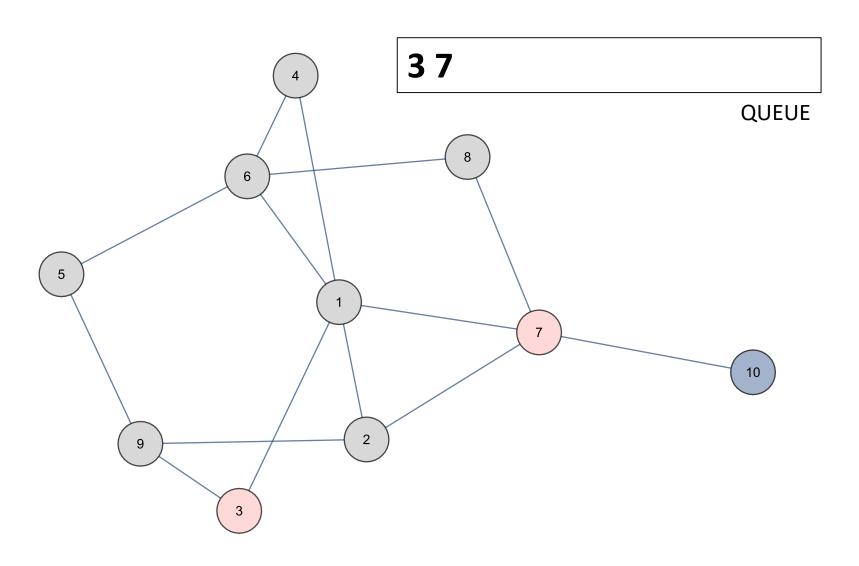


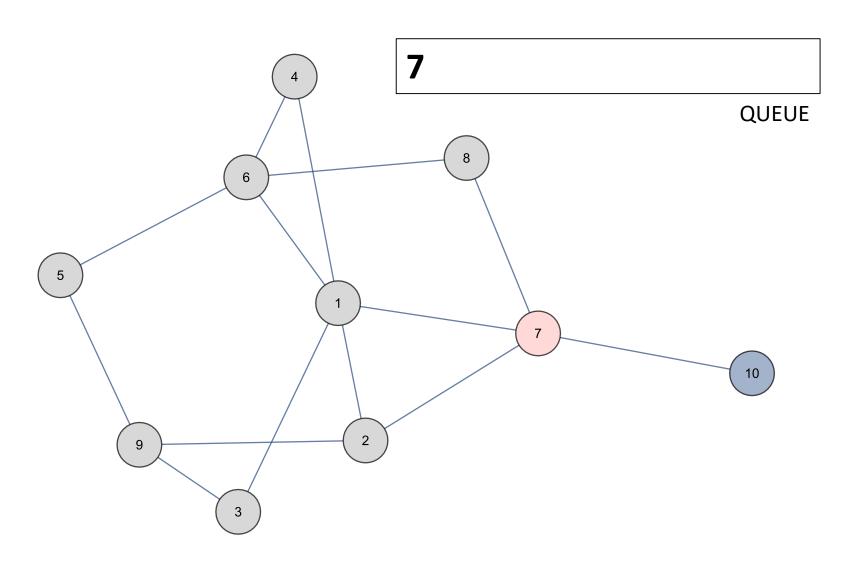


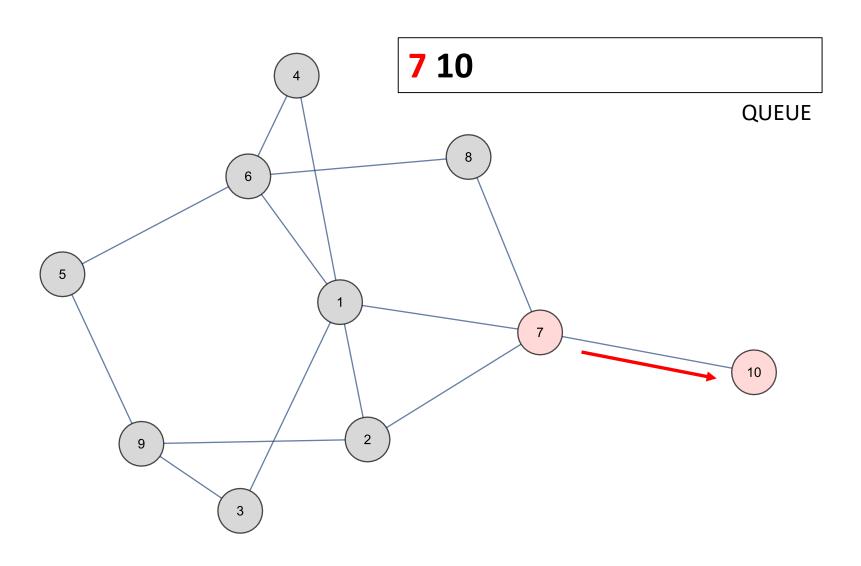


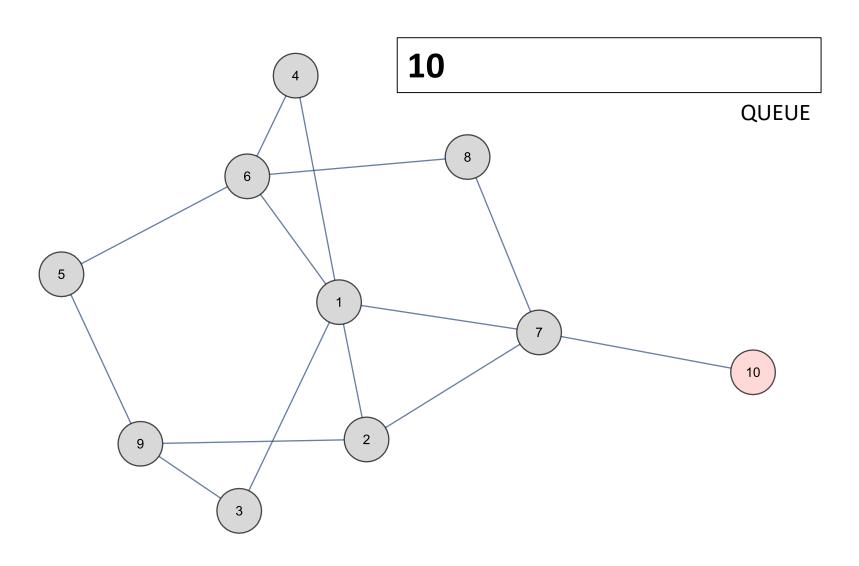


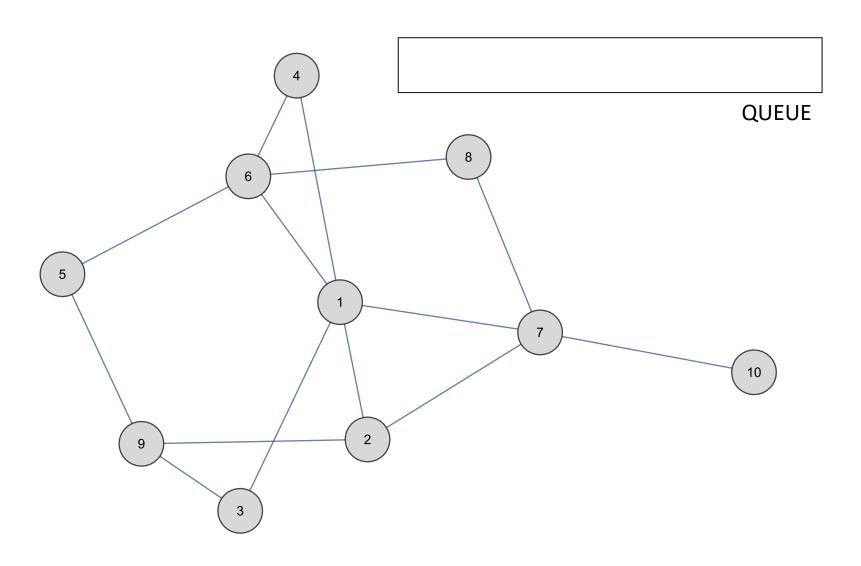












```
# Global/class scope variables
n = number of nodes in the graph
g = adjacency list representing unweighted graph
# s = start node, e = end node, and 0 ≤ e,s < n
function bfs(s, e):

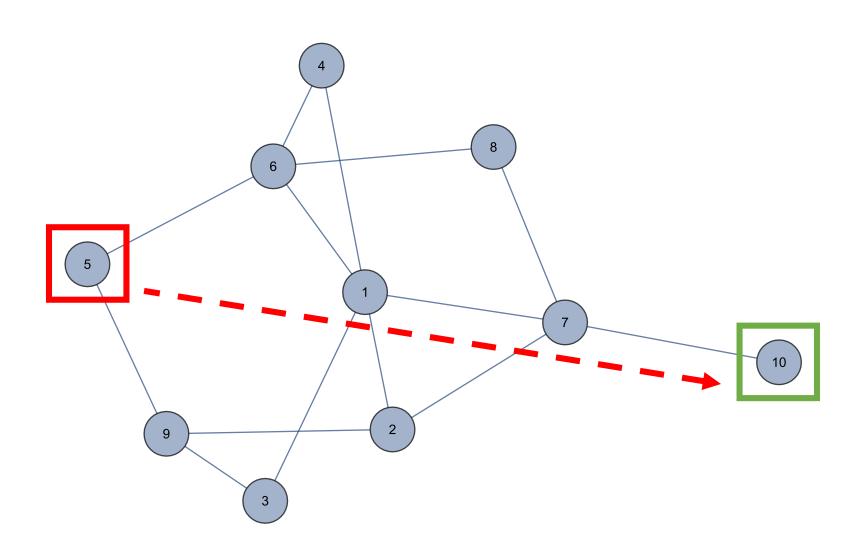
# Do a BFS starting at node s
prev = solve(s)

# Return reconstructed path from s → e
return reconstructPath(s, e, prev)</pre>
```

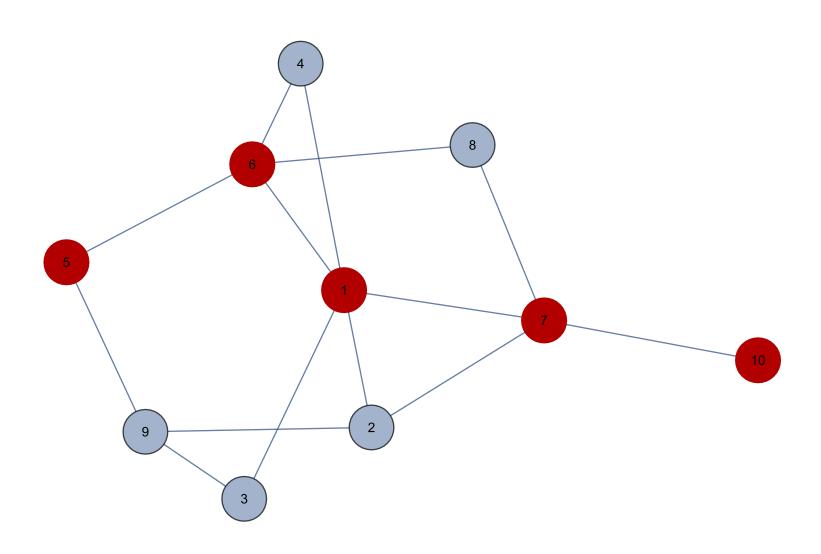
```
function solve(s):
 q = queue data structure with enqueue and dequeue
 q.enqueue(s)
 visited = [false, ..., false] # size n
 visited[s] = true
  prev = [null, ..., null] # size n
 while !q.isEmpty():
    node = q.dequeue()
    neighbours = g.get(node)
    for(next : neighbours):
      if !visited[next]:
        q.enqueue(next)
        visited[next] = true
        prev[next] = node
  return prev
```

```
function reconstructPath(s, e, prev):
 # Reconstruct path going backwards from e
  path = []
  for(at = e; at != null; at = prev[at]):
    path.add(at)
  path.reverse()
 # If s and e are connected return the path
  if path[0] == s:
    return path
  return []
```

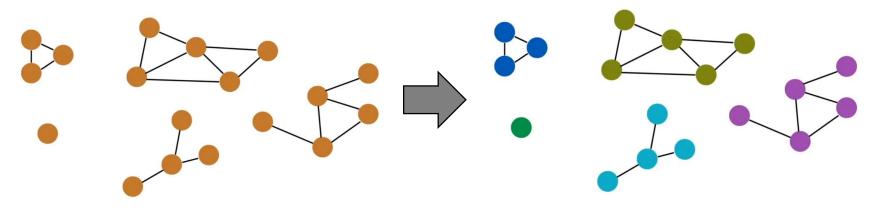
Shortest Path Problem



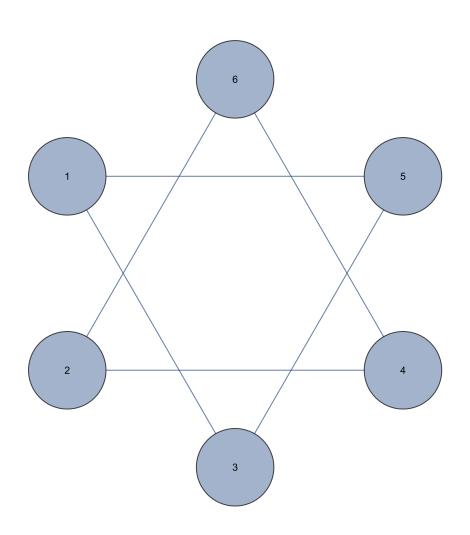
Shortest Path Problem

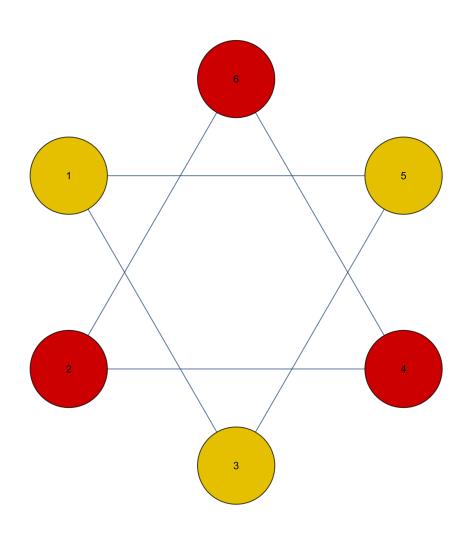


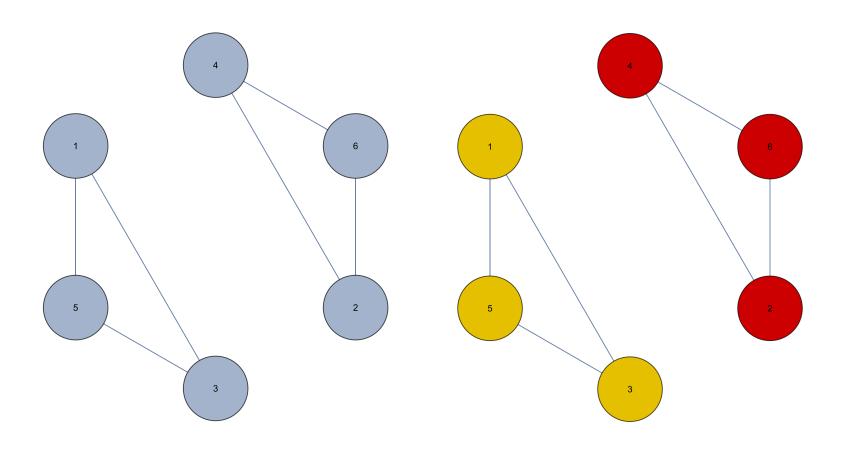
In graph theory, a **component**, sometimes called a **connected component**, of an <u>undirected graph</u> is a <u>subgraph</u> in which any two <u>vertices</u> are <u>connected</u> to each other by <u>paths</u>, and which is connected to no additional vertices in the <u>supergraph</u>. For example, the graph shown in the illustration has three components. A vertex with no incident edges is itself a component. A graph that is itself connected has exactly one component, consisting of the whole graph.



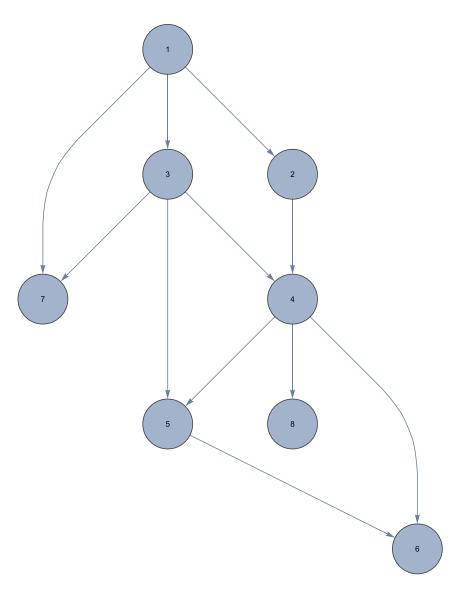
A graph with 5 components.

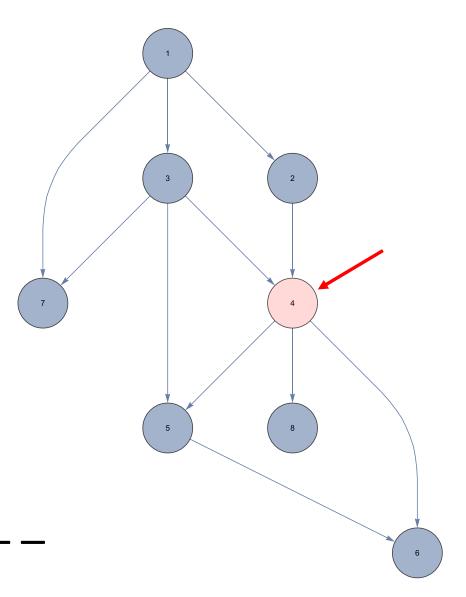


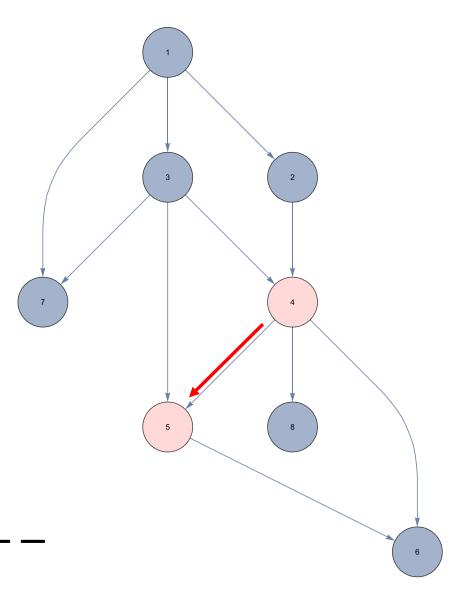


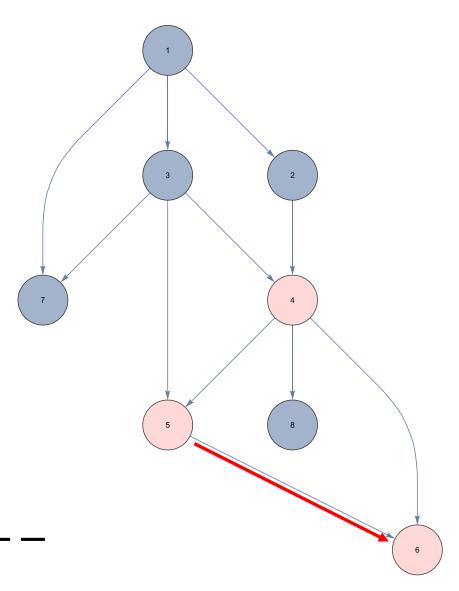


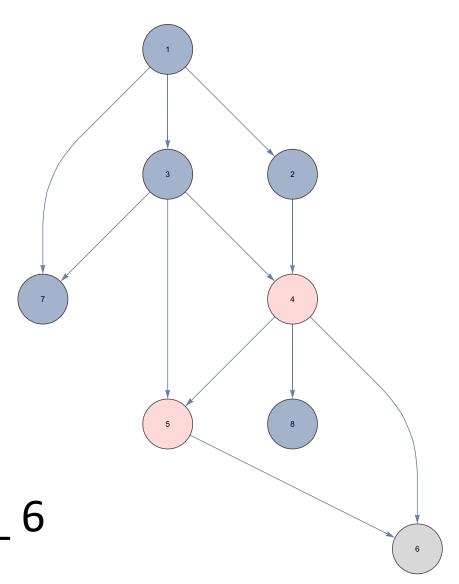
```
# Global or class scope variables
n = number of nodes in the graph
g = adjacency list representing graph
count = 0
components = empty integer array # size n
visited = [false, ..., false] # size n
function findComponents():
  for (i = 0; i < n; i++):
    if !visited[i]:
      count++
      dfs(i)
  return (count, components)
function dfs(at):
  visited[at] = true
  components[at] = count
  for (next : g[at]):
    if !visited[next]:
      dfs(next)
```

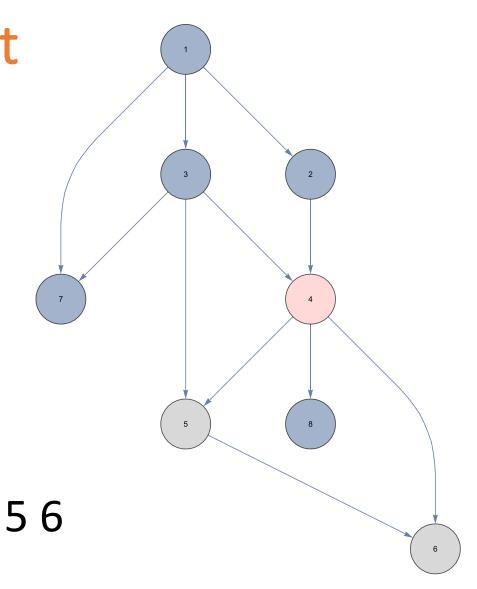


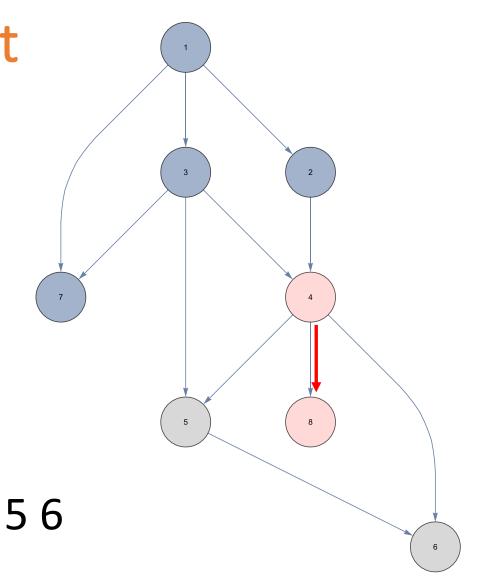


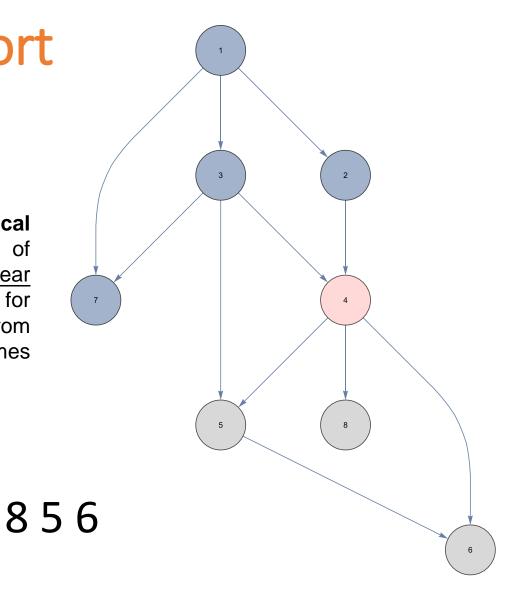


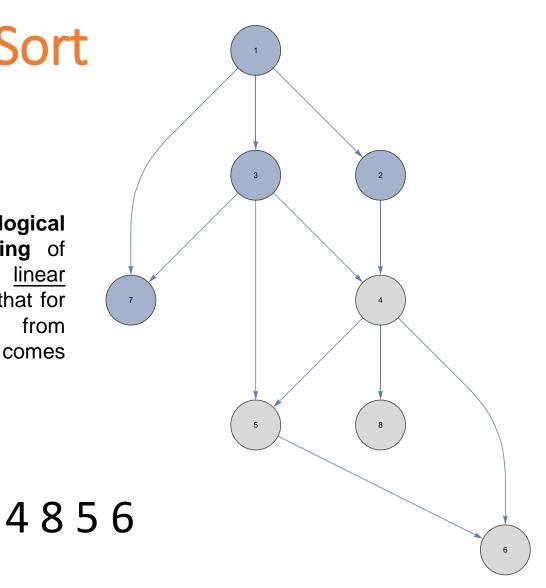


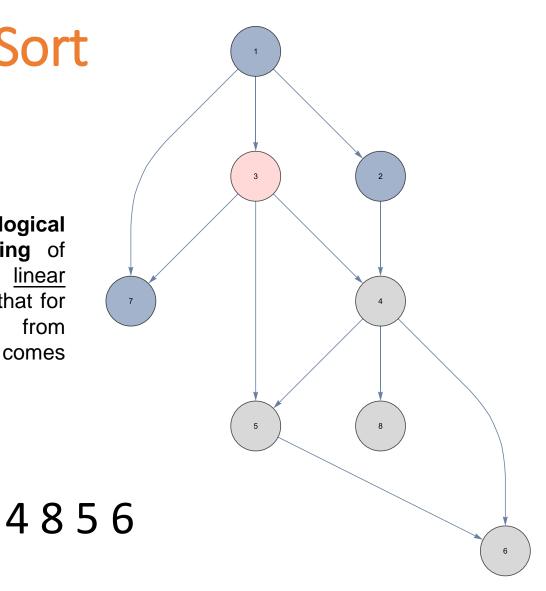




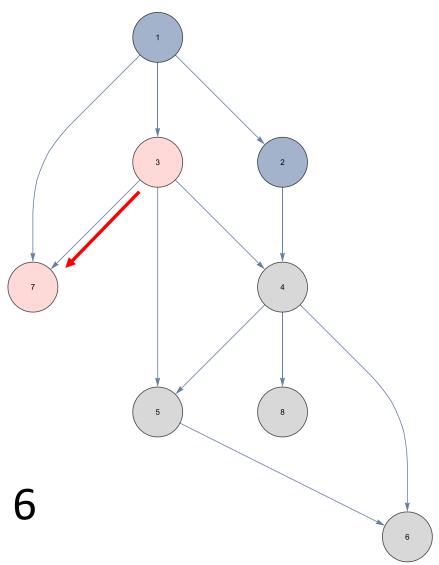




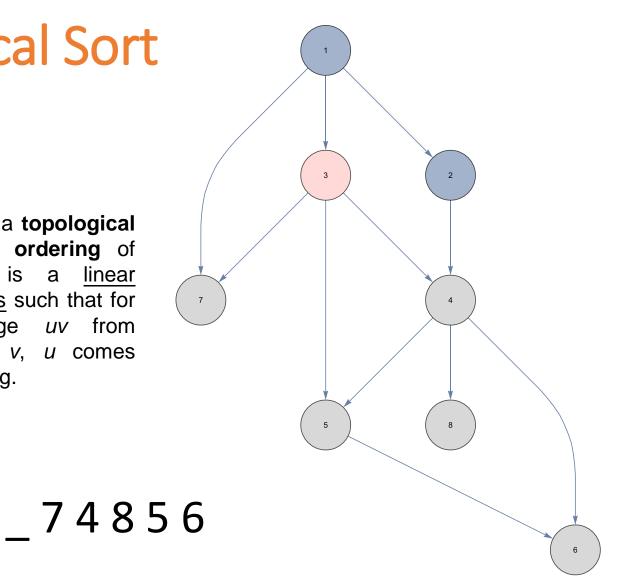




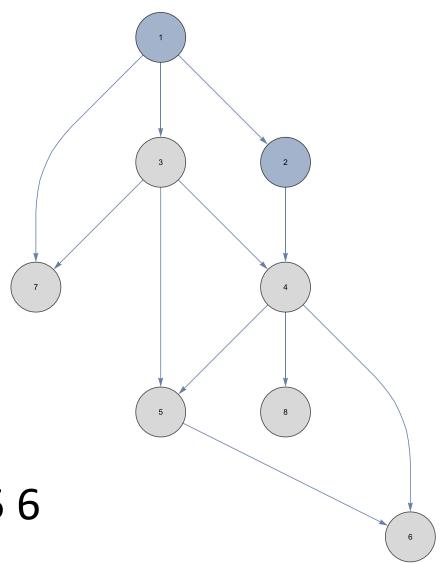
In <u>computer science</u>, a **topological sort** or **topological ordering** of a <u>directed graph</u> is a <u>linear ordering</u> of its <u>vertices</u> such that for every directed edge *uv* from vertex *u* to vertex *v*, *u* comes before *v* in the ordering.



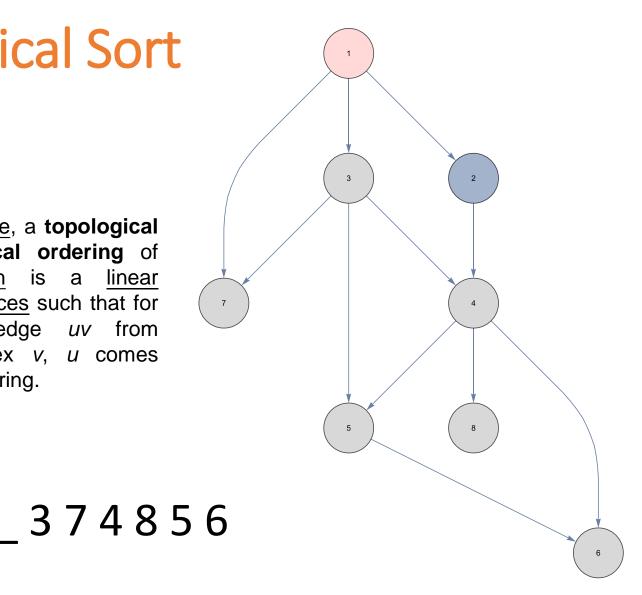
_ _ _ _ 4 8 5 6



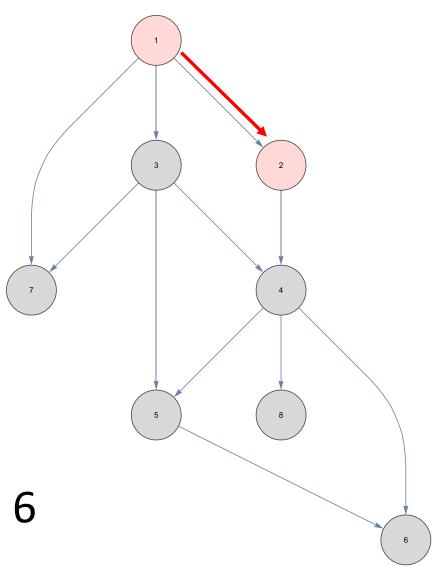
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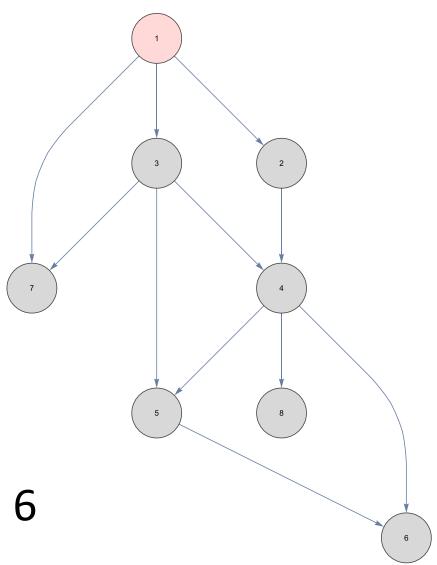


In <u>computer science</u>, a **topological sort** or **topological ordering** of a <u>directed graph</u> is a <u>linear ordering</u> of its <u>vertices</u> such that for every directed edge *uv* from vertex *u* to vertex *v*, *u* comes before *v* in the ordering.



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In <u>computer science</u>, a **topological sort** or **topological ordering** of a <u>directed graph</u> is a <u>linear ordering</u> of its <u>vertices</u> such that for every directed edge *uv* from vertex *u* to vertex *v*, *u* comes before *v* in the ordering.



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Topological Sort

In <u>computer science</u>, a **topological sort** or **topological ordering** of a <u>directed graph</u> is a <u>linear ordering</u> of its <u>vertices</u> such that for every directed edge *uv* from vertex *u* to vertex *v*, *u* comes before *v* in the ordering.

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Topological Sort

```
# Assumption: graph is stored as adjacency list
function topsort(graph):
  N = graph.numberOfNodes()
  V = [false,...,false] # Length N
  ordering = [0,...,0] # Length N
  i = N - 1 \# Index for ordering array
  for(at = 0; at < N; at++):
    if V[at] == false:
      visitedNodes = []
      dfs(at, V, visitedNodes, graph)
      for nodeId in visitedNodes:
        ordering[i] = nodeId
        i = i - 1
  return ordering
```

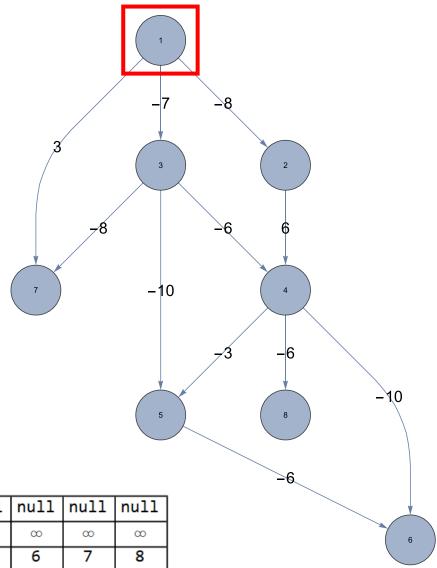
Topological Sort

```
# Execute Depth First Search (DFS)
function dfs(at, V, visitedNodes, graph):

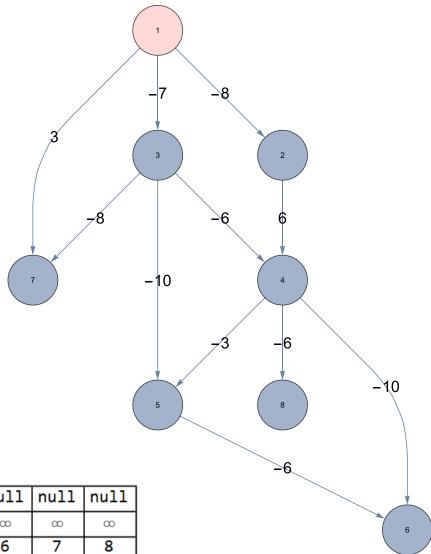
V[at] = true

edges = graph.getEdgesOutFromNode(at)
for edge in edges:
   if V[edge.to] == false:
        dfs(edge.to, V, visitedNodes, graph)

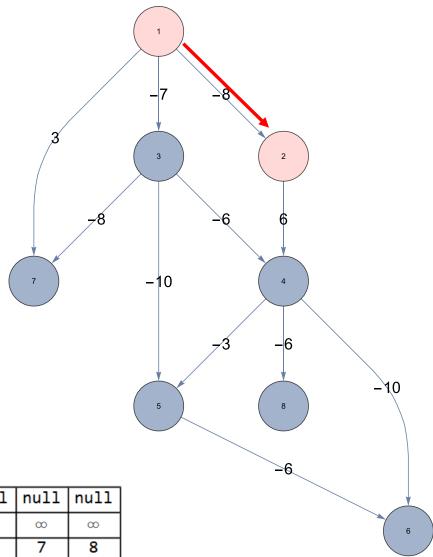
visitedNodes.add(at)
```



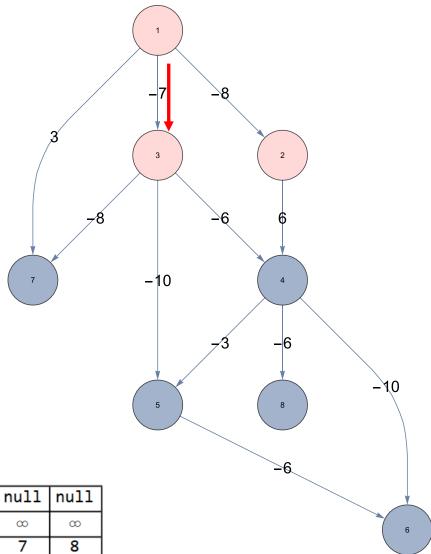
from	null	null	null	null	null	null	null	null
distance	ω	∞	∞	œ	œ	8	ω	∞
vertex	1	2	3	4	5	6	7	8



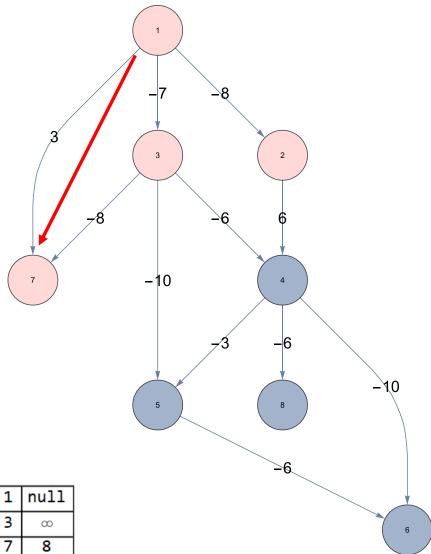
from	1	null	null	null	null	null	null	null
distance	0	∞	œ	œ	œ	∞	œ	œ
vertex	1	2	3	4	5	6	7	8



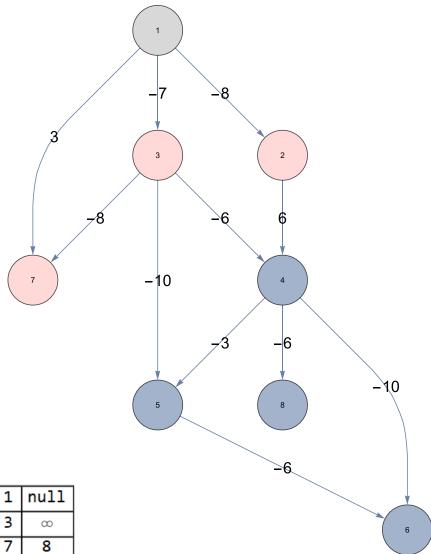
from	1	1	null	null	null	null	null	null
distance	0	-8	8	8	8	8	8	8
vertex	1	2	3	4	5	6	7	8



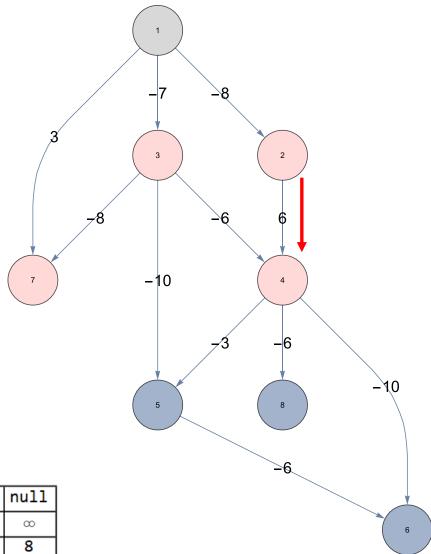
from	1	1	1	null	null	null	null	null
distance	0	-8	-7	∞	œ	œ	œ	∞
vertex	1	2	3	4	5	6	7	8



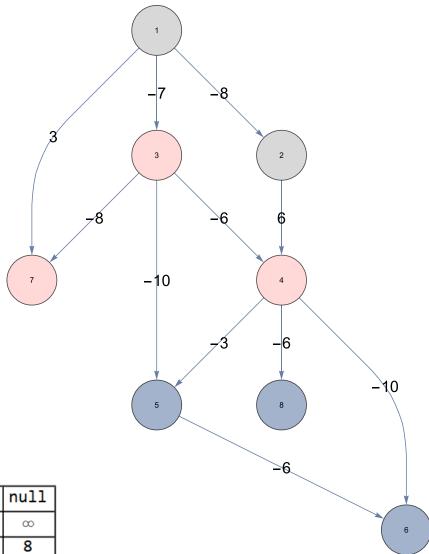
from	1	1	1	null	null	null	1	null
distance	0	-8	-7	8	œ	8	3	∞
vertex	1	2	3	4	5	6	7	8



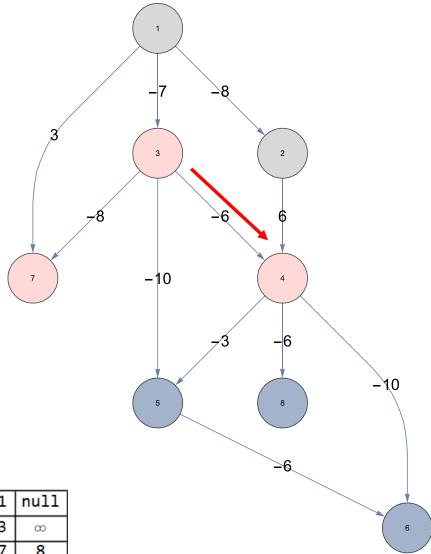
from	1	1	1	null	null	null	1	null
distance	0	-8	-7	8	œ	8	3	∞
vertex	1	2	3	4	5	6	7	8



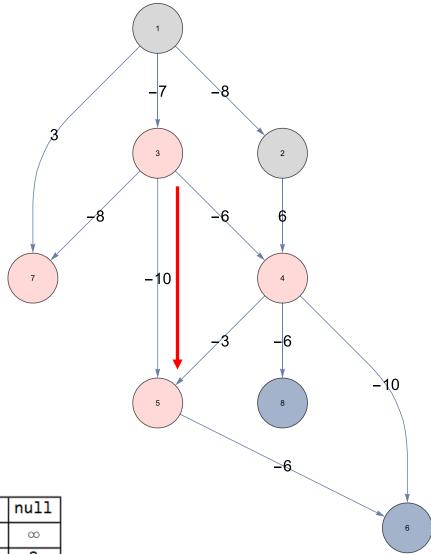
from	1	1	1	2	null	null	1	null
distance	0	-8	-7	-2	œ	∞	3	∞
vertex	1	2	3	4	5	6	7	8



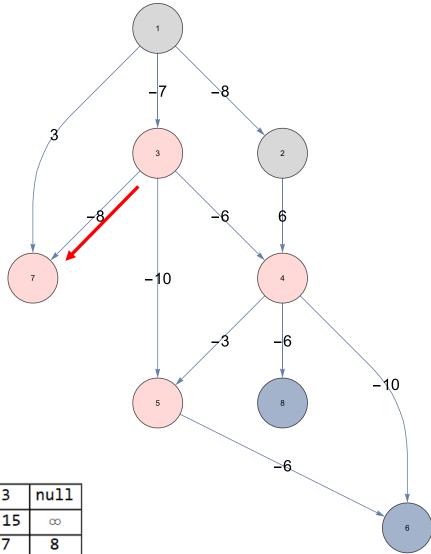
from	1	1	1	2	null	null	1	null
distance	0	-8	-7	-2	œ	8	3	∞
vertex	1	2	3	4	5	6	7	8



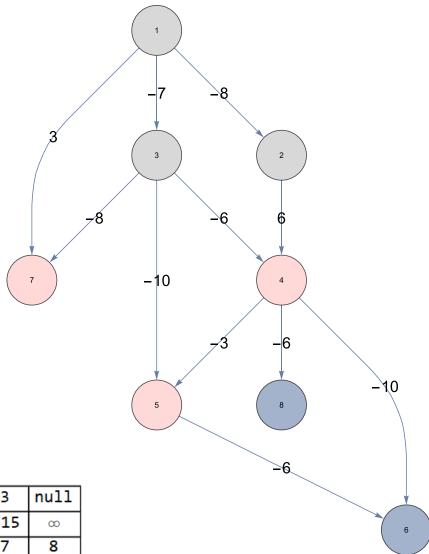
from	1	1	1	3	null	null	1	null
distance	0	-8	-7	-13	œ	œ	3	œ
vertex	1	2	3	4	5	6	7	8



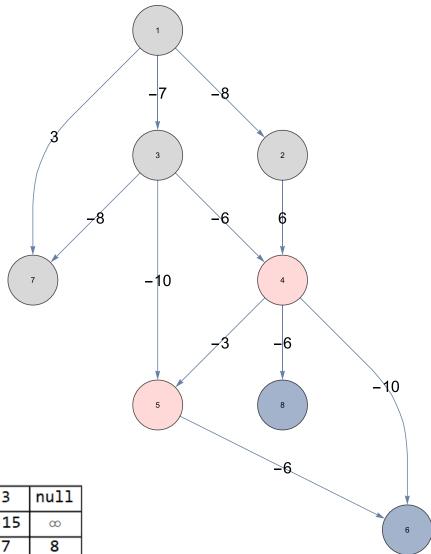
from	1	1	1	3	3	null	1	null
distance	0	-8	-7	-13	-17	∞	3	ω
vertex	1	2	3	4	5	6	7	8



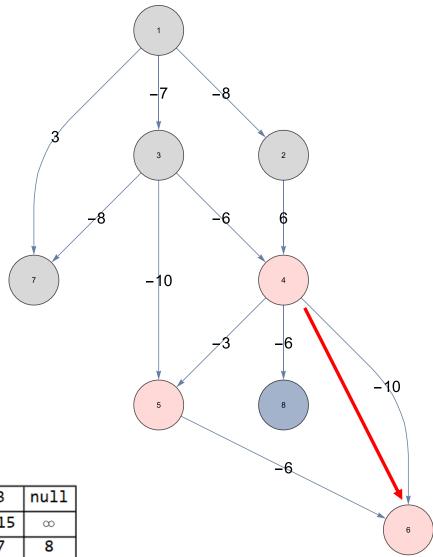
from	1	1	1	3	3	null	3	null
distance	0	-8	-7	-13	-17	œ	-15	œ
vertex	1	2	3	4	5	6	7	8



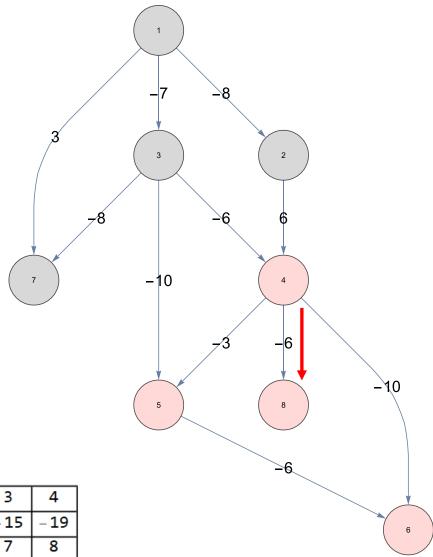
from	1	1	1	3	3	null	3	null
distance	0	-8	-7	-13	-17	œ	-15	∞
vertex	1	2	3	4	5	6	7	8



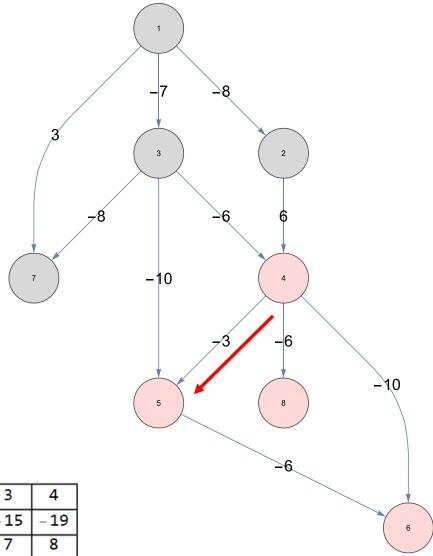
from	1	1	1	3	3	null	3	null
distance	0	-8	-7	-13	-17	œ	-15	œ
vertex	1	2	3	4	5	6	7	8



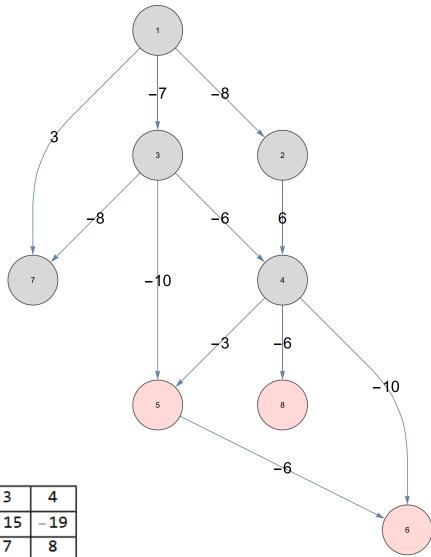
from	1	1	1	3	3	4	3	null
distance	0	-8	-7	-13	-17	-23	-15	∞
vertex	1	2	3	4	5	6	7	8



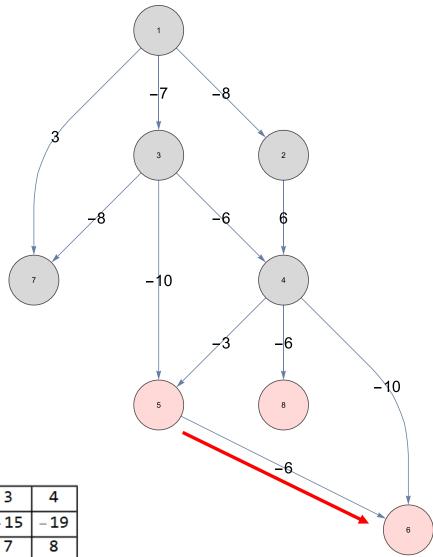
from	1	1	1	3	3	4	3	4
distance	0	-8	-7	-13	-17	-23	-15	-19
vertex	1	2	3	4	5	6	7	8



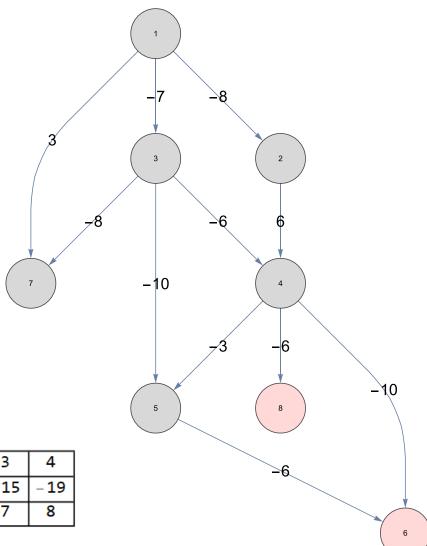
from	1	1	1	3	3	4	3	4
distance	0	-8	-7	-13	-17	-23	-15	- 19
vertex	1	2	3	4	5	6	7	8



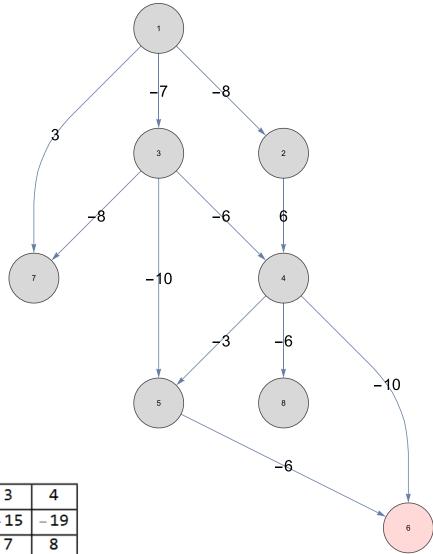
from	1	1	1	3	3	4	3	4
distance	0	-8	-7	-13	-17	-23	-15	- 19
vertex	1	2	3	4	5	6	7	8



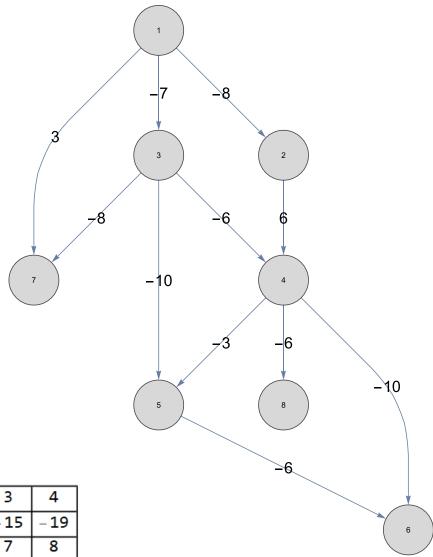
from	1	1	1	3	3	4	3	4
distance	0	-8	-7	-13	-17	-23	-15	- 19
vertex	1	2	3	4	5	6	7	8



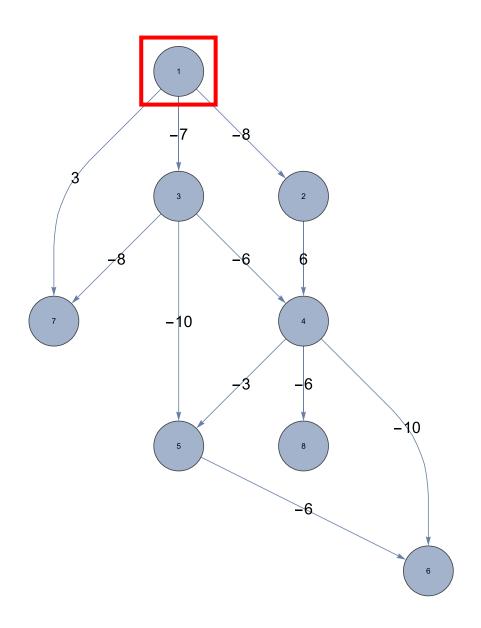
from	1	1	1	3	3	4	3	4
distance	0	-8	-7	-13	-17	-23	-15	-19
vertex	1	2	3	4	5	6	7	8



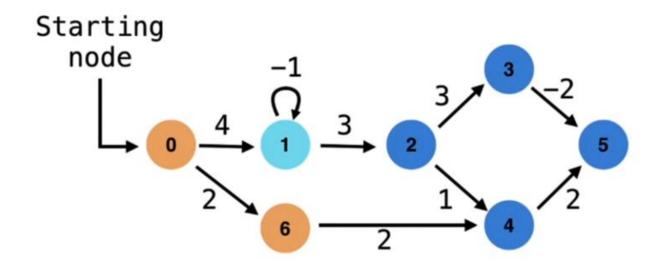
from	1	1	1	3	3	4	3	4
distance	0	-8	-7	-13	-17	-23	-15	- 19
vertex	1	2	3	4	5	6	7	8



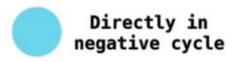
from	1	1	1	3	3	4	3	4
distance	0	-8	-7	-13	-17	-23	-15	-19
vertex	1	2	3	4	5	6	7	8

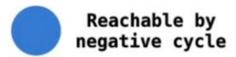


Negative cycles can manifest themselves in many ways...

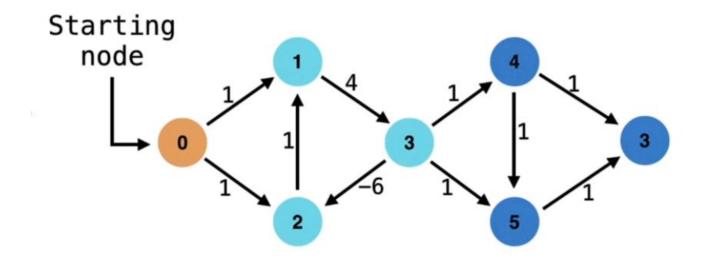


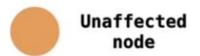


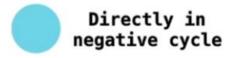


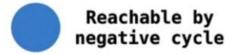


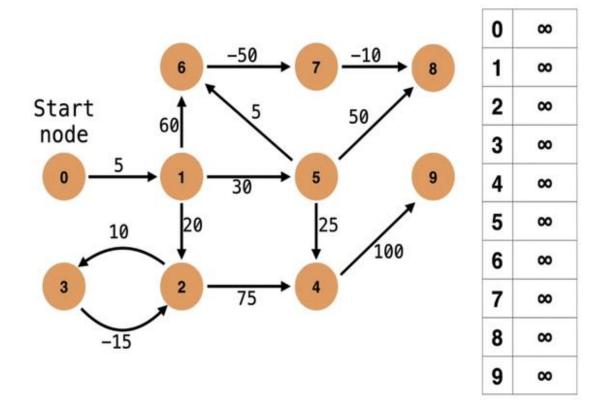
Negative cycles can manifest themselves in many ways...

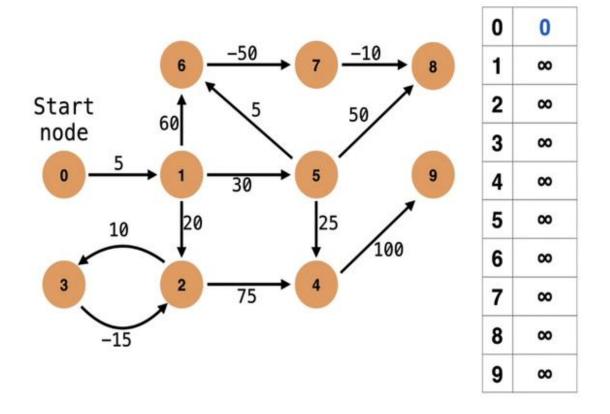


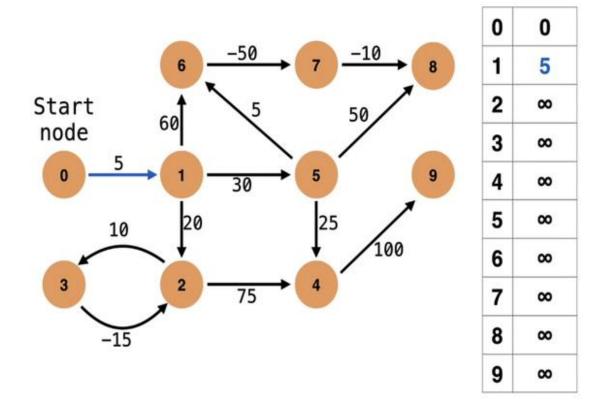


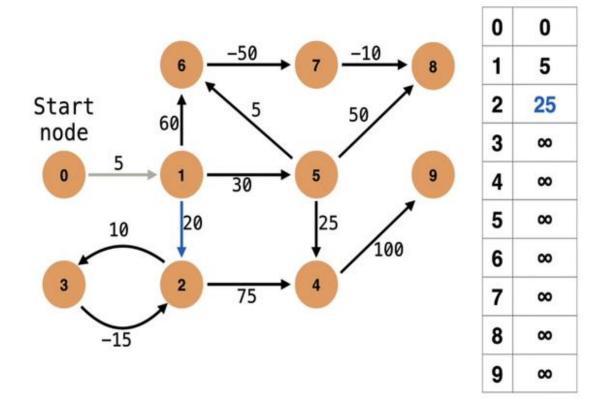


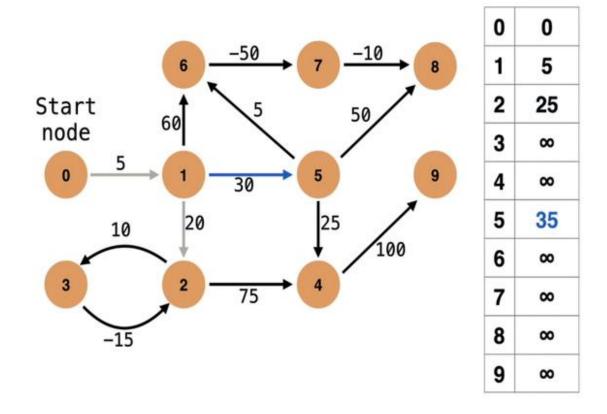


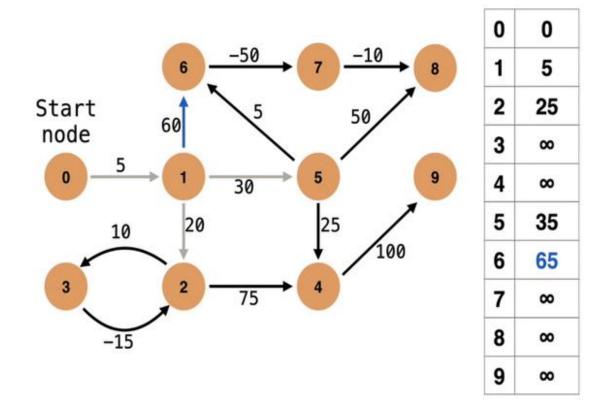


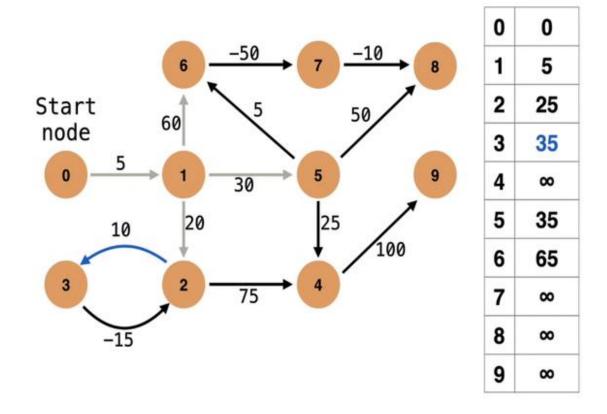


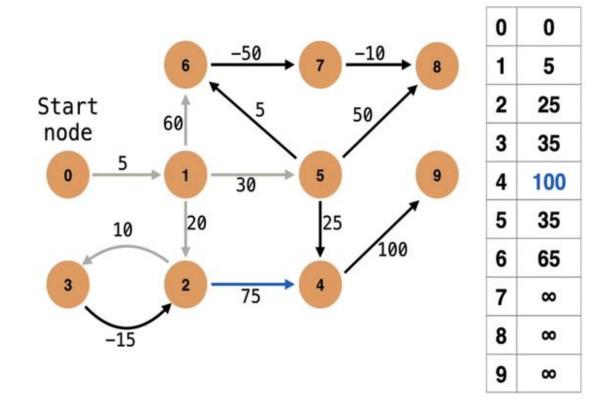


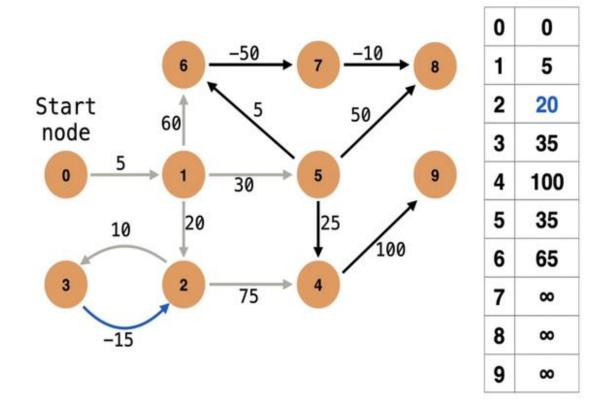


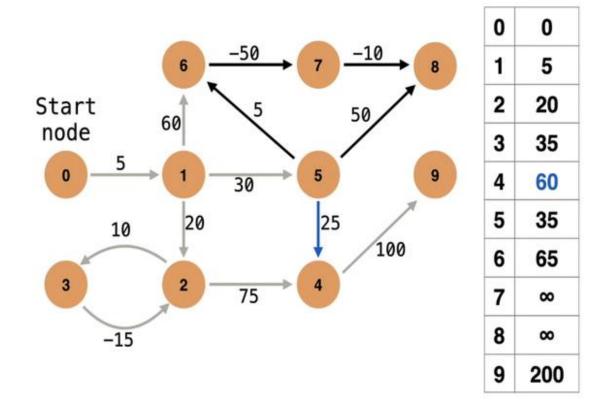


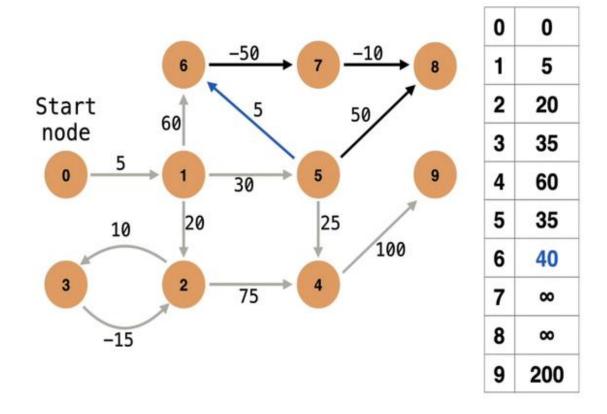


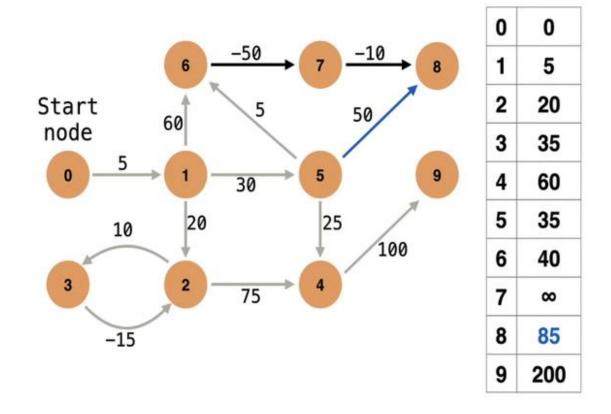


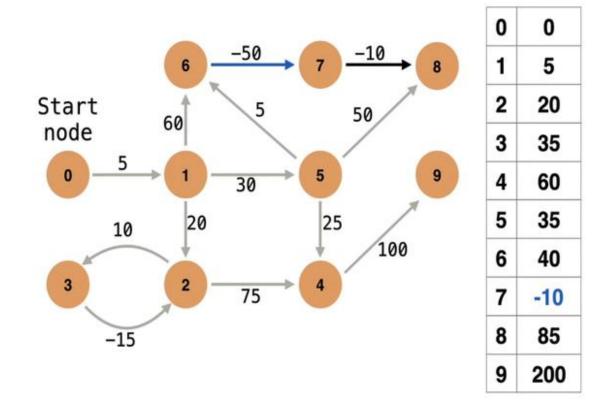


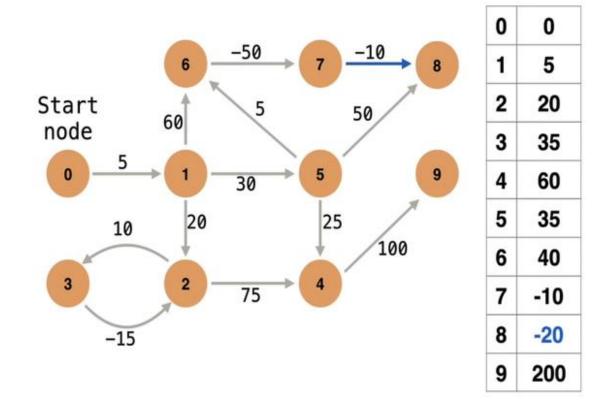


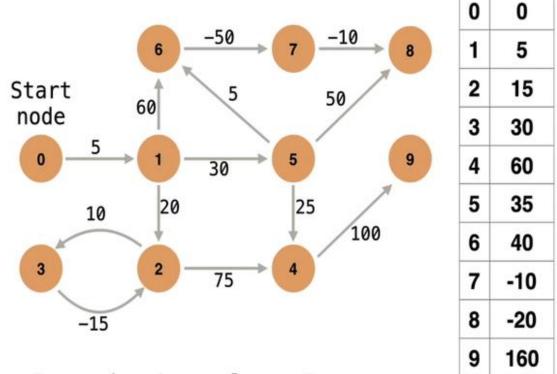






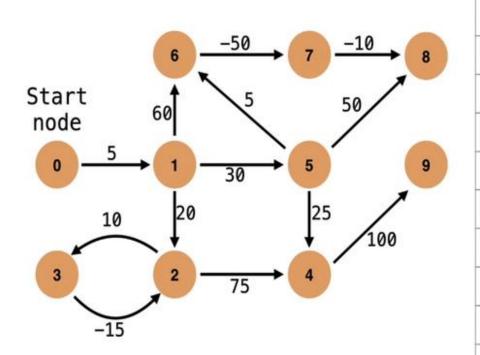






Iteration 2 complete, 7 more to go...

Let's fast-forward to the end...



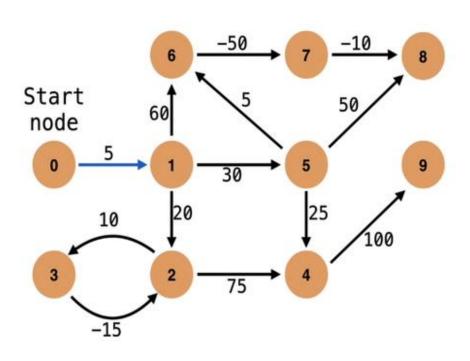
-20

-5

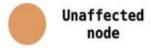
-10

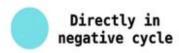
-20

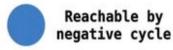
We're finishe	d with	the S	SSSP	part.	Now	let's
detect tho	se nega	tive	cycle	es. If	we	can
relax an edge	then	there	's a	negati	ive o	cycle.

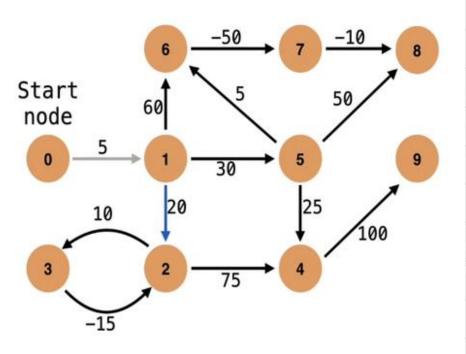


0	0
1	5
2	-20
3	-5
4	60
5	35
6	40
7	-10
8	-20
9	160

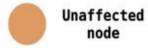


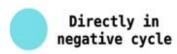


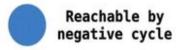


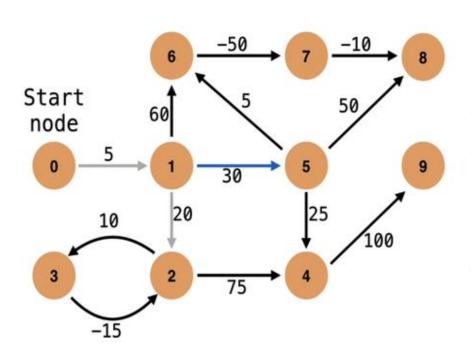


0	0
1	5
2	-20
3	-5
4	60
5	35
6	40
7	-10
8	-20
9	160

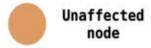


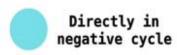


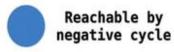


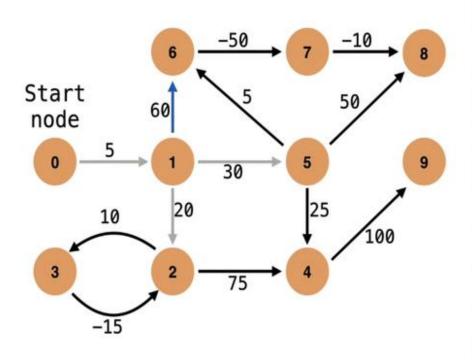


0	0
1	5
2	-20
3	-5
4	60
5	35
6	40
7	-10
8	-20
9	160

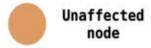


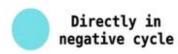


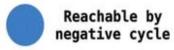


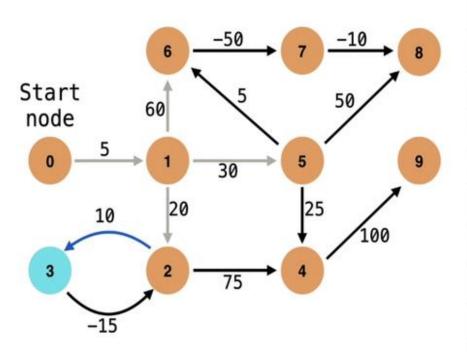


0	0
1	5
2	-20
3	-5
4	60
5	35
6	40
7	-10
8	-20
9	160

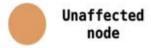


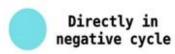


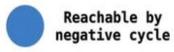


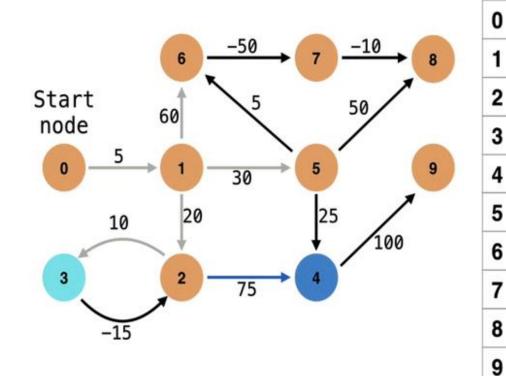


0	0
1	5
2	-20
3	-00
4	60
5	35
6	40
7	-10
8	-20
9	160

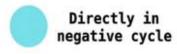


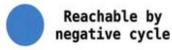






Unaffected node





0

5

-20

-00

-00

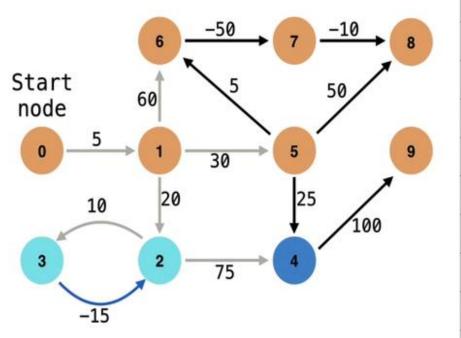
35

40

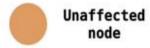
-10

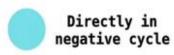
-20

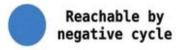
160



0	0
1	5
2	-∞
3	-00
4	-00
5	35
6	40
7	-10
8	-20
9	160

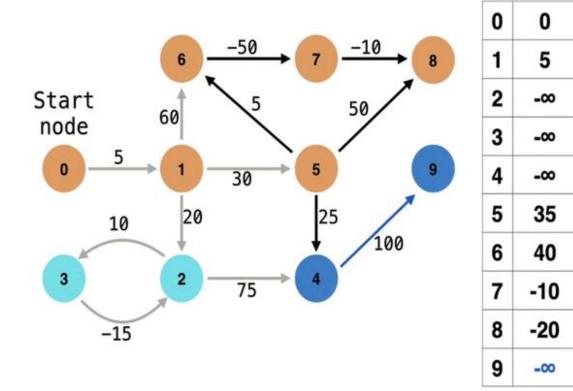






Unaffected

node



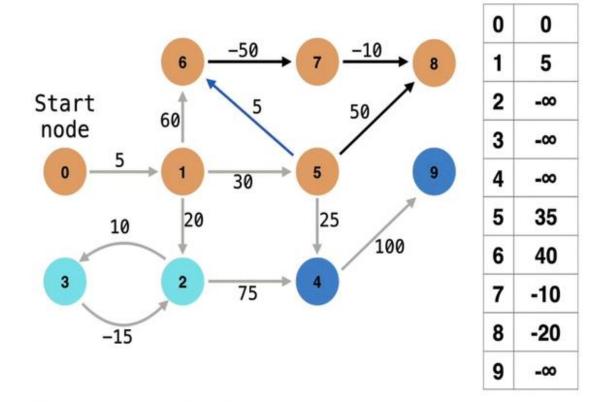
Directly in

negative cycle

Reachable by

Unaffected

node



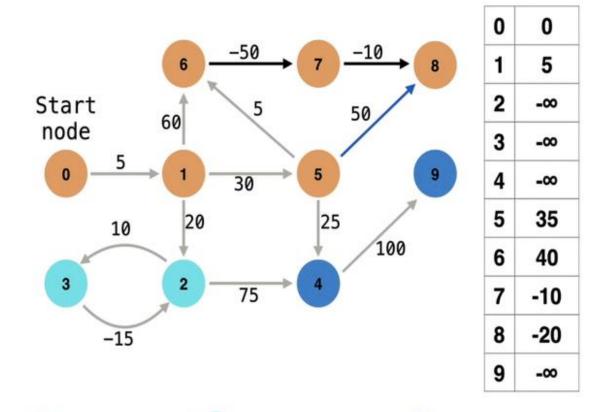
Directly in

negative cycle

Reachable by

Unaffected

node



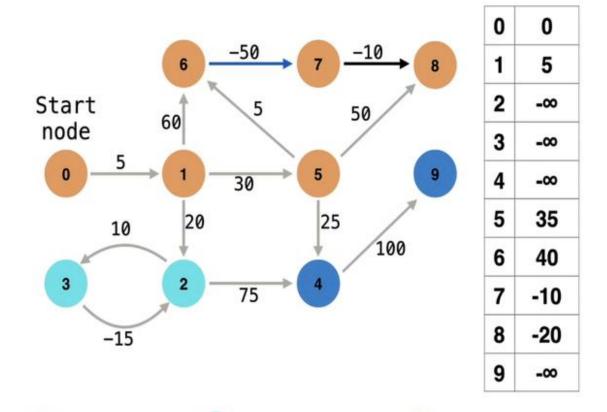
Directly in

negative cycle

Reachable by

Unaffected

node



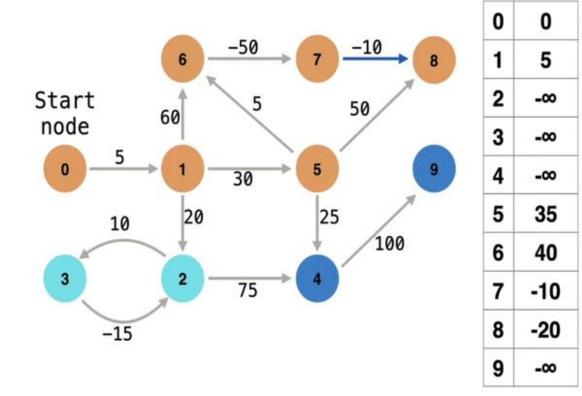
Directly in

negative cycle

Reachable by

Unaffected

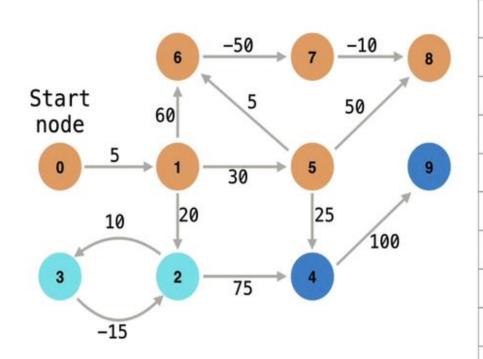
node



Directly in

negative cycle

Reachable by



0

5

-00

-00

-00

35

40

-10

-20

-00

0

2

3

4

5

6

8

9

Repeat this for another 8 iterations in order to ensure the cycles fully propagate. In this example, we happened to detect all cycles on the first iteration, but this was a coincidence.

Dijkstra algorithm Bellman-Ford Algorithm Floyd-Warshall algorithm

• • •

The basic DFS and BFS algorithm is the building block of some more complicated algorithms.

References

https://github.com/williamfiset/Algorithms/tr ee/master/src/main/java/com/williamfiset/alg orithms/graphtheory