GRAPHS

```
You can represent a graph like G=(V,E,c) →

E = Set of edges

V = Set of vertices = nodes

c = weight

|E| = Size of a graph: the number of edges in the graph

|V| = Order of a graph: the number of nodes in the graph

deg(v) = Degree of a vertex: the number of edges adjacent to v.

For a directed graph we have e=(u,v) → u = start vertex, v = end vertex

deg⁻(v) = In-degree: the number of edges coming into v in a directed graph.

deg⁺(v) = Out-degree: the number of edges going out of v in a directed graph.

Adjacency matrix: (Adj) is a nxn (n = |V|) matrix which represents the edges between vertices.

Adj(i,j) = {

1 if the vi and vj are adjacent

0 otherwise
```

Breadth-First Search (BFS)

Time complexity: O(V+E) if we use adjacency lists

You can use BFS to/for:

- 1. Start from an initial node (u) to see what vertices are reachable from u.
- 2. Calculate the shortest distance (d(u,v)) from node (u) to any other node (v) in an unweighted graph.
- **3.** Garbage collection
- 4. Social networking
- 5. Network broadcast
- **6.** Web crawling
- 7. Solve puzzles and games like rubik's cube.
- **8.** Test if a graph bipartite (its vertices can be divided into two disjoint sets (no edge between the vertices in one set)

BFS(s):

```
\label{eq:continuous_problem} \begin{split} & \text{Initialize FIFO queue Q as being empty.} \\ & Q.\text{push(s)} \\ & s.\text{parent} = -1 \\ & \textbf{while Q != null:} \\ & u = Q.\text{pop()} & \textit{// Remove node u from Q.} \\ & \textbf{for } v \in u.\text{adj} & \textit{// for every neighbor of u:} \\ & \textbf{if } v.\text{parent} == \text{null} \\ & v.\text{parent} = u \\ & Q.\text{push(v)} \end{split}
```

Depth First Search (DFS)

Time complexity: O(V+E) if we use adjacency lists

You take a path and keep going until you reach a dead-end, then you go back and take another path till you explore all the possible paths/graph. The algorithm is exactly the same as BFS, the only difference is that it uses a **stack** instead of a FIFO queue.

You can use DFS to/for:

```
1. Start from an initial node (u) to see what vertices are reachable from u.
```

- **2.** Edge classification
- **3.** Cycle detection
- 4. Topological sorting: job scheduling
- 5. Solve mazes.

```
\begin{array}{ll} \text{timer} = 0; \\ \textbf{DFS(V)} & \text{$//$ Explore the whole graph} \\ \textbf{for } s \in V & \\ \textbf{if } s.parent == null \\ s.parent = -1 \\ DFS\_visit(s) & \end{array}
```

```
DFS_visit(s):  // Start from a node (u) to see what vertices are reachable from that timer++
s.start = timer
for v ∈ s.adj
if v.start == null  // or if v.parent == null
v.parent = s
DFS_visit(v)
else if v.start != null && v.end == null
// Cycle detected!
end
timer++
s.end = timer
```

Edge classification:

```
<u>Tree edge</u>: the edge we take to visit a new vertex

<u>Forward edge</u>: connects a node to a descendant

<u>Backward edge</u>: connects a node to an ancestor

<u>Cross edge</u>: all the other edges

((x,y) \in E \Rightarrow x.start < y.start < y.end < x.end)
((x,y) \in E \Rightarrow y.start < x.start < x.end < y.end)
((x,y) \in E \Rightarrow y.start < y.finish < x.start < x.finish)
```

Cycle detection:

✓ G has e cycle iff G has a back edge.

Topological sorting: gives you a topological order/sort of vertices of a **directed acyclic graph** (DAG). For every edge in a DAG $((v_i, v_j) \in E)$ we have v_i is before v_j in a topological order of $v_1, v_2, ..., v_n$.

✓ Run DFS to calculate the start and end time of the vertices. (Make sure that you have a DAG before going to the next step).

- ✓ As each vertex finishes, insert it onto a linked list.
- ✓ Printing the linked list, gives you the topological order of the vertices.

Dijkstra Algorithm

Time complexity: O((E+V)logV) if we use a min-heap

This algorithm calculates the shortest distance (d(u,v)) from node (u) to any other node (v).

```
Dijkstra(V, s)
       for v \in V
                v.dst = inf;
       s.dst = 0;
       s.parent = -1
       min heap.build(V)
                                                                  // Build a min-heap based on dst
       while min heap \neq null:
               u = min_heap.delete(root)
                                                                  // Choose a node with the smallest distance.
               for v \in u.adj
                                                                  // for every neighbor of u
                        if v.parent == null
                                                                  // Not explored
                                v.parent = u
                        if u.dst + w(u,v) < v.dst
                                v.parent = u.
                                v.dst = u.dst + w(u,v)
                                min_heap.heapify(v)
                                                                 // Update the min_heap
```

Minimum Spanning Tree (MST)

Tree: a tree with **n** nodes has **n-1** edges. \rightarrow |E|=|V|-1 and does not have any cycles.

Spanning tree: a tree that is connected to all the nodes in a given connected undirected graph.

Minimum Spanning Tree (MST): a spanning tree with the smallest weight between all the other spanning trees.

Prim's Algorithm

Time complexity: O(ElogV) if we use min-heaps

This algorithm finds MST on a given connected undirected graph.

- 1. Initialize an empty tree
- 2. Choose a random node and add it to your tree.

- **3.** Find an edge with the smallest weight among the edges that connect a vertex of the tree to a vertex **not** in the tree.
- **4.** Add the selected edge and its vertex to the tree.
- **5.** Go back to step 3 until you cover all the vertices in your tree (|E|=|V|-1)

```
Prim(V, s)
        for v \in V
                v.cost = inf;
        s.cost = 0:
        s.parent = -1
        min heap.build(V)
                                                                 // Build a min-heap based on cost
        while min_heap \neq null:
                u = min\_heap.delete(root)
                                                                 // Choose a node with the smallest cost.
                                                                 // for every neighbor of u
                for v \in u.adj
                        if v.parent == null
                                                                 // Not explored
                                v.parent = u
                        if w(u,v) < v.cost
                                v.parent = u.
                                v.cost = w(u,v)
                                                                 // Update the min_heap
                                min_heap.heapify(v)
```

Kruskal's Algorithm

Time complexity: O(*ElogV*)

Kruskal's Algorithm also finds MST on a given connected undirected graph.

- 1. Sort the edges based on their weights
- 2. Picking an edge with the smallest weight if it does not make a cycle with the chosen ones
- **3.** Go back to Step 2 until the number of the chosen edges become |V|-1 This algorithm finds MST on a given connected undirected graph.