

### Graph Theory

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#### Topics of Today

- 1. What is a graph?
- 2. Graph Traversal (BFS + DFS)
- 3. Shortest Distance (Dijkstra's Algorithm)
- 4. Minimum Spanning Tree (Kruskal's Algorithm)
- 5. Graph Bi-Coloring (Bipartite Checking Algorithm)



#### What is a Graph?

- A data structure formed by connecting nodes (a.k.a vertex) using edges
- Types of Graphs:



#### How is a Graph Represented?

- Two ways:
  - 1. Adjacency List
  - 2. Adjacency Matrix

#### Graph Representation – Adjacency List

- Using a dictionary that maps each node to a list of connected nodes
- Good for both directed and undirected graphs
- Defaults to unweighted, can be weighted through some "hacking"

graph	=	{				
		0:	[1],	,		
		1:	[0,	2,	3,	4],
		2:	[1,	5],		
		3:	[1,	4],	,	
		4:	[1,	3,	5],	
		5:	[]			
}						

## Graph Representation – Adjacency Matrix

- Uses a n×n list to represent connection between nodes
- Good for directed/undirected, weighted/unweighted

#### Graph Representation – Adj. List vs. Adj. Matrix

- Let *E* be the number of edges in our graph, and *N* be the number of nodes in our graph
- Space complexity of adjacency list: O(E+N)
- Space complexity of adjacency matrix: O(N<sup>2</sup>)

#### Graph Traversal - Motivation

- A ton of use cases, just to name a few:
  - 1. Searching
  - 2. Graph manipulation
  - 3. Foundations for other algorithms
  - 4. Finding shortest path between two nodes
    - Only efficient for unweighted graph

#### Graph Traversal – Depth First Search (DFS)

- Similar to DFS for trees
- A visited set is used to keep track of nodes already visited
- Pseudocode:

```
def dfs(graph, curr_node, visited):
    print(curr_node)
    visited << curr_node
    for neighbour of graph[curr_node]:
        if (neighbour is not in visited):
            dfs(graph, neighbour, visited)</pre>
```

# Graph Traversal – Breadth First Search (BFS)

- Similar to BFS for trees
- queue is used to keep track of visit order
- Pseudocode:

```
def bfs(graph):
  visited <- list
  queue <- Queue
  queue << graph[0]
  visited[0] = True
  while queue is not empty:
     curr = queue.pop()
     print(curr)
     for adj in graph[curr]:
        if (visited[adj] == False):
          queue << adj
          visited[adj] = True
```

#### Dijkstra's Algorithm

- Algorithm for finding the shortest path from one node to every other node
- Graph can be weighted/unweighted, directed/undirected, cyclic/acyclic BUT NO NEGATIVE EDGES

```
def dijkstra(adj_matrix, source):
   nodes <- Set
   dist <- dict
   prev <- dict
   for vertex in adj_matrix:
      dist[vertex] <- INFINITY</pre>
      prev[vertex] <- None</pre>
      nodes << vertex
   dist[source] <- 0
   while nodes is not empty:
      u <- vertex in nodes with min dist[u]
      remove u from nodes
      for neighbour of u:
         new_path = dist[u] + adj_matrix[source][u]
         if new path < dist[source]:</pre>
            dist[source] <- new path</pre>
            prev[source] <- u
```

**return** (dist, prev)

#### Dijkstra's Algorithm - Complexities

- Let *V* be the number of vertices, *E* the number of edges
- Time complexity: O(V+E)
- Space complexity: O(V)
- Further exploration: Bellman-Ford algorithm, Floyd-Warshall algorithm

### Minimum Spanning Trees (MST)

- A MST is a subgraph that connects all vertices together with the minimum possible total edge weight
- We will only consider MST for undirected graphs
- Intuitive example: a telephone company wants to lay cables for a community, a MST will be the most efficient way to lay these cables to reach every home

#### MST – Kruskal's Algorithm – Disjoint Set

- A set that is partitioned into a number of subsets
- Operations:
  - makeset (node): Adds a node to the disjoint set in its own subset
  - find (node): Finds the representative element (root node) of the node
  - union(x, y): Merges nodes x and y

MST – Kruskal's Algorithm – Disjoint Set

```
node <- int
  parent <- DSNode
class DisjointSet:
   ds set <- Set
  def makeset(node):
      if node not in ds set:
         ds set << node
  def find(node):
      if node.parent is not node:
         node.parent = find(node.parent)
      return node.parent
  def union(x, y):
     x_{root}, y_{root} = find(x), find(y)
      y root.parent = x root
```

#### MST – Kruskal's Algorithm

```
def mst(graph):
    edges <- sort(graph.edges)
    vertices <- DisjointSet
    for vertex in graph:
        vertices << vertex
    mst_edges <- list
    for e in edges:
        if (vertices.find(e.pointA) != vertices.find(e.pointB)):
            vertices.union(e.pointA, e.pointB)
            mst_edges << e
    return mst edges
```

#### MST – Kruskal's Algorithm – Complexities

- Let *E* be the number of edges, V be the number of vertices
- Average case complexity: O(E logV)
- Space complexity: O(E+V)
- Further exploration: Prim's MST algorithm



#### **Bipartite Graph**

- A bipartite graph is a graph whose vertices can be decomposed into two disjoint sets such that no two vertices within the same set are adjacent
- Used on undirected, unwieghted graphs

#### Bipartite Checking Algorithm

```
def bipartite(graph):
  colors <- dict
  colors[graph[0]] = True
  queue <- Queue
  queue << graph[0]
  while queue is not empty:
     curr node = queue.pop()
     curr color = colors[curr node]
     for child in curr node.neighbours:
        if colors[child] is None:
           colors[child] = not curr color
           queue << child
        elif colors[child] == curr color:
           return False
```

return True

### Bipartite Checking Algorithm

- Let V be the number of vertices and E be the number of edges
- Time complexity: O(V+E)
- Space complexity: O(V)

#### Further Exploration

- In addition to the previously mentioned:
  - Tarjan's Algorithm for finding Strongly Connected Components
  - Tarjan's Algorithm for Articulation Points
  - Johnson's Algorithm for finding all the cycles in a directed graph
  - Bellman-Ford Algorithm to detect negative cycles in the graph
  - N-Queen problem
  - Travelling Salesman problem