

# Lab Sheet – 3

## Graph Algorithms in Real – Life Applications

### Problem 1: Social Network Friend Suggestion

#### ➤ Time Complexity:

The algorithm's time complexity is  **$O(V + E)$** , where  $V$  is the number of users (vertices) and  $E$  is the number of friendships (edges).

- To find "friends of friends," we first iterate through the target user's direct friends (let's say they have  $k$  friends).
- Then, we iterate through the friends of each of those  $k$  friends.
- In the worst case, this traversal resembles a Breadth-First Search (BFS) or Depth-First Search (DFS) that extends two levels out from the user. The complexity of a full graph traversal is  $O(V+E)$ , and this operation is a subset of that, making its upper bound  $O(V+E)$ .

#### ➤ Scalability for Large Networks:

- For a massive network like LinkedIn or Facebook, which has billions of users and edges, performing an  $O(V+E)$  traversal in real-time every time a user requests suggestions is not feasible. The query would be too slow.
- Real-world systems solve this scalability problem by **pre-computation**. Friend suggestions are calculated offline (e.g., during off-peak hours) using distributed graph processing frameworks.
- The results are then stored in a database or cache. When you visit your profile, the application simply performs a quick lookup to retrieve this pre-computed list of suggestions, which feels instantaneous to you.

••• Finding friend suggestions for: Alice

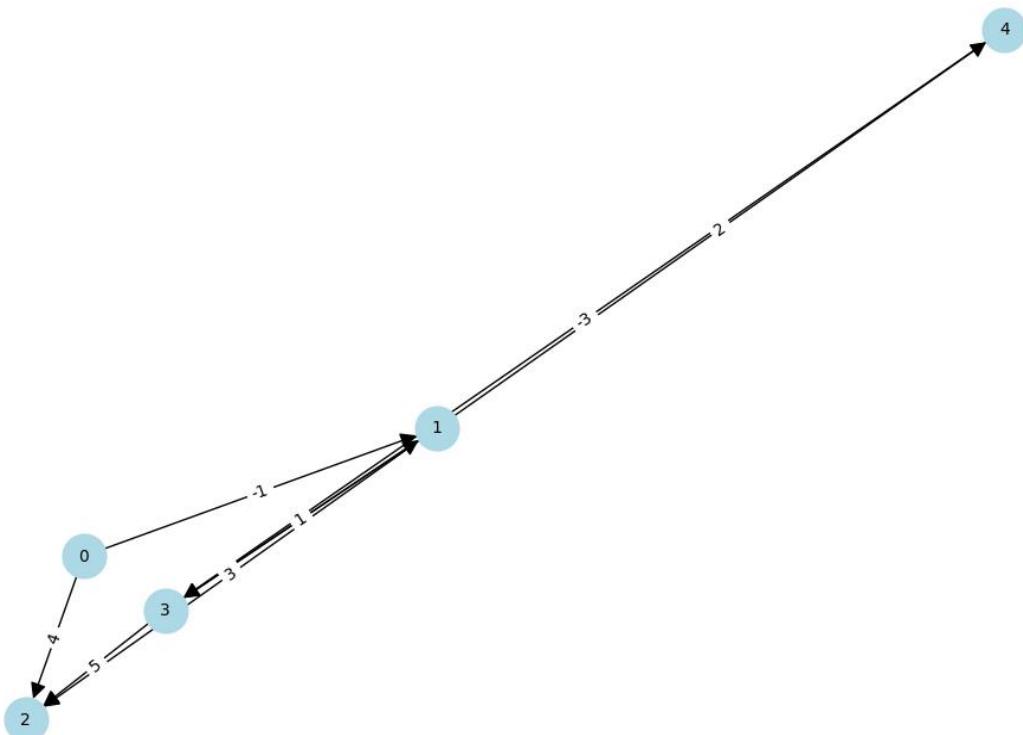
Graph Adjacency List: {'Alice': {'Carol', 'Bob'}, 'Bob': {'Eve', 'David', 'Alice'}, 'Carol': {'David', 'Alice', 'Fiona'},

Direct friends for Alice: ['Carol', 'Bob']

Suggested friends for Alice: ['Eve', 'David', 'Fiona']

Execution Time: 0.000084639 seconds

## Problem 2: Route Finding on Google Maps



➤ Preference for Negative Weights:

Bellman-Ford is preferred over Dijkstra's algorithm in graphs with negative edge weights because Dijkstra's greedy approach fails in this scenario. Dijkstra's assumes that once a path to a node is finalized, it's the shortest

possible path. However, a negative edge encountered later could create a "shortcut" that violates this assumption, leading to an incorrect result. Bellman-Ford's iterative approach relaxes all edges  $V-1$  times, which guarantees it finds the correct shortest path even if negative weights are present. It can also detect negative weight cycles<sup>2</sup>, which Dijkstra's cannot.

➤ Time Complexity  $O(V * E)$ :

The time complexity of the Bellman-Ford algorithm is  $O(V * E)$ , where  $V$  is the number of vertices (locations) and  $E$  is the number of edges (roads).

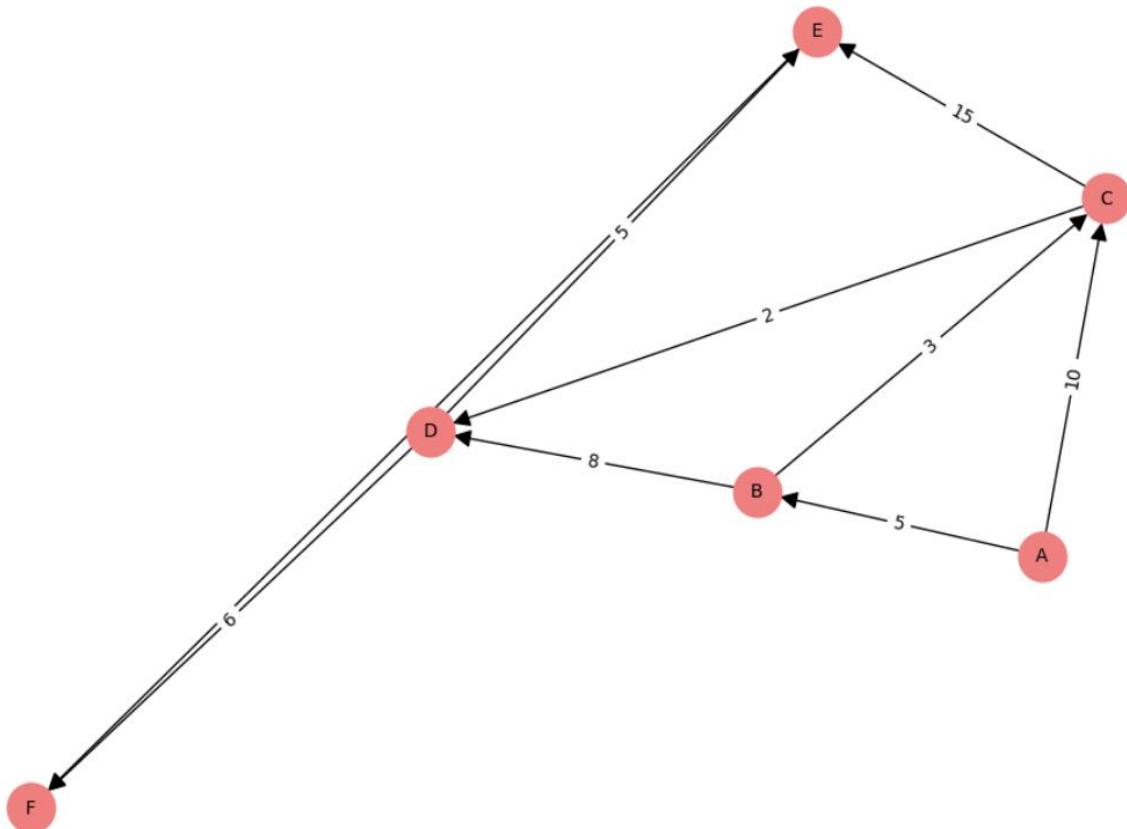
- The algorithm's core is a main loop that iterates  **$V-1$  times**.
- Inside this main loop, it iterates through **every edge ( $E$ )** in the graph to "relax" it (i.e., check if a shorter path can be found via that edge).
- This results in a complexity of  $(V-1) \times E$ .
- An additional pass over all  $E$  edges is performed to detect negative weight cycles<sup>4</sup>.
- Therefore, the total time complexity simplifies to  **$O(V * E)$** .

```
...
--- Bellman-Ford Results ---
Shortest distances from source node 0:
Node 0: 0
Node 1: -1
Node 2: 2
Node 3: -2
Node 4: 1

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### Problem 3: Emergency Response System

Emergency Response Map (Travel Times)



➤ Time Complexity:  $O(E \log V)$  using min-heap

The time complexity for Dijkstra's algorithm, when implemented efficiently using a min-heap (priority queue), is  $O(E \log V)$ .

- $V$  is the number of vertices (intersections), and  $E$  is the number of edges (roads).
- Every vertex is added to the min-heap once. Extracting the minimum-distance vertex from the heap takes  $O(\log V)$  time.
- We perform this extraction  $V$  times.
- For every edge, we might perform a "decrease-key" operation (updating the distance), which also takes  $O(\log V)$  time.

- This gives a total complexity of  $O(V \log V + E \log V)$ , which simplifies to  **$O(E \log V)$**  in a connected graph (where  $E$  is typically greater than or equal to  $V$ ).
- Why Dijkstra is Unsuitable for Negative Weights.

Dijkstra's algorithm is unsuitable for graphs with negative edge weights because its core "greedy" strategy fails:-

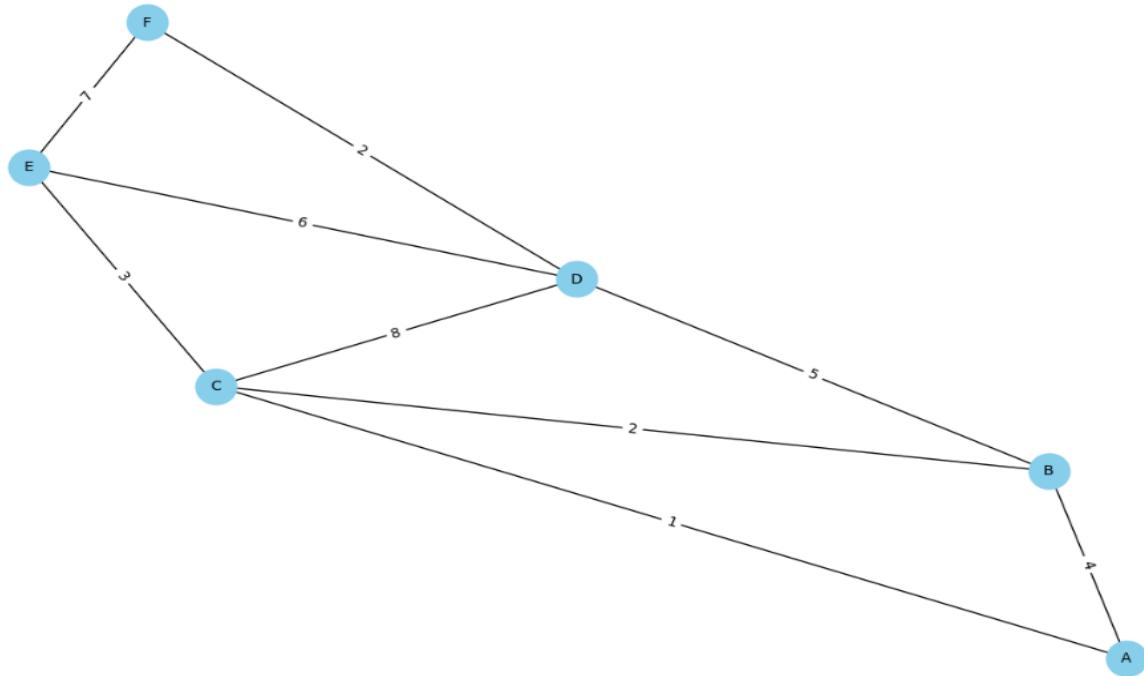
- Dijkstra's operates on the assumption that once it finalizes the shortest path to a vertex, no shorter path can ever be found.
- A **negative edge** breaks this rule. A path that looks longer might later connect to a negative edge, creating a "shortcut" that results in a shorter overall distance.
- Because Dijkstra's finalizes paths greedily without looking ahead for potential negative shortcuts, it will fail to find the correct shortest path in such cases. Bellman-Ford (Problem 2) is used instead as it correctly handles negative weights.

```
...
--- Dijkstra's Algorithm Results ---
Fastest travel times from A:
  To A: 0
  To B: 5
  To C: 8
  To D: 10
  To E: 15
  To F: 16

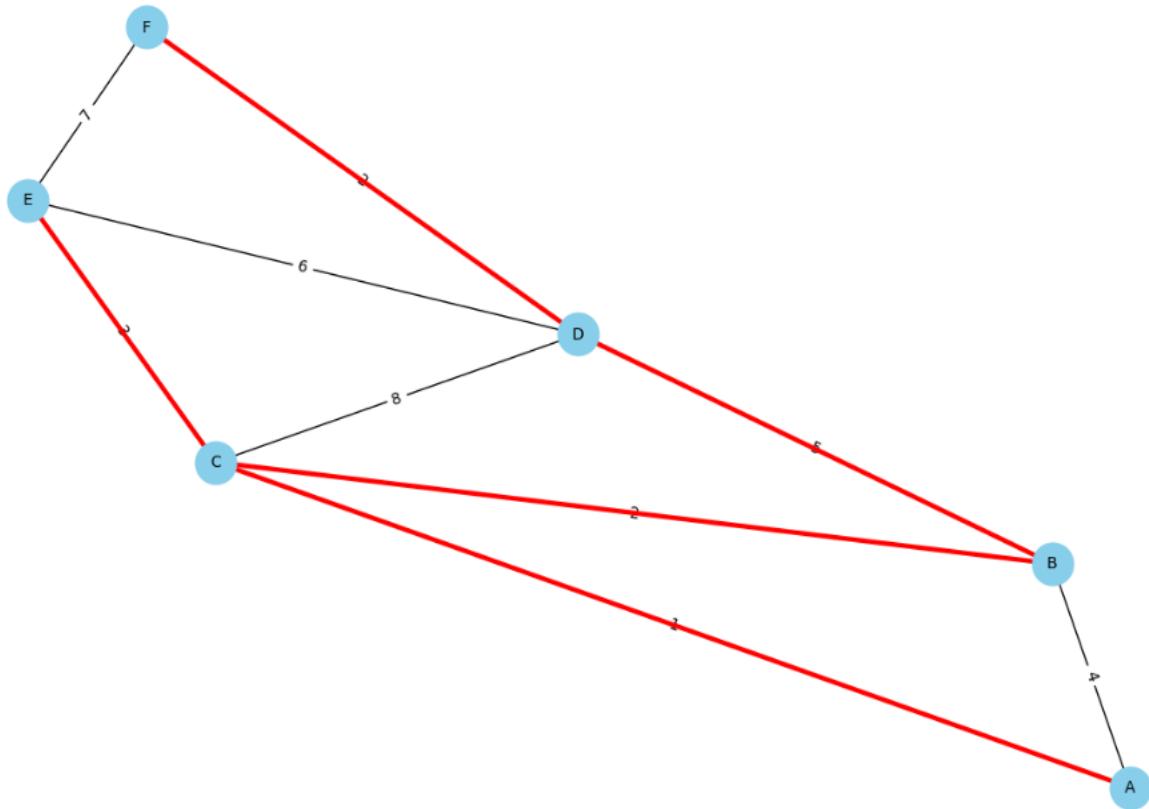
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## Problem 4: Network Cable Installation

Potential Cable Installation Paths (Costs)



Minimum Spanning Tree (Total Cost: 13)



- Complexity Comparison: Prim's vs. Kruskal's <sup>1</sup>
- Prim's Algorithm (using a min-heap): The time complexity is  $O(E \log V)$ . This is because every edge ( $E$ ) is processed, and heap operations (like adding an edge or extracting the minimum edge) take  $O(\log V)$  time.
- Kruskal's Algorithm (using Union-Find): The time complexity is  $O(E \log E)$ . This is dominated by the initial step of sorting all  $E$  edges by weight. The subsequent Union-Find operations are very fast (nearly constant time on average).
- Comparison: Prim's is generally faster for dense graphs (where  $E$  is close to  $V^2$ ), while Kruskal's is often faster for sparse graphs (where  $E$  is much smaller than  $V^2$ ).

### ➤ Applicability in Infrastructure Cost Optimization

Minimum Spanning Tree (MST) algorithms are directly applicable to infrastructure cost optimization. This problem is a classic example: finding the minimum total length of cable needed to connect all offices. This same logic applies to many real-world scenarios, such as:

- Telecom & IT: Connecting all nodes in a computer network with the least amount of fiber optic cable<sup>4</sup>.
- Power Grids: Designing the cheapest way to connect all cities or substations to a power source.
- Transportation: Building the minimum length of road or railway required to connect a set of towns.

```
--- Prim's Algorithm MST Results ---
Total minimum cost to connect all offices: 13
```

```
Edges selected in MST:
A -- C (Cost: 1)
C -- B (Cost: 2)
C -- E (Cost: 3)
B -- D (Cost: 5)
D -- F (Cost: 2)
```

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<b>Problem</b>	<b>Graph Algorithm</b>	<b>Time Complexity</b>	<b>Application Domain</b>
Social Network Suggestion	BFS / DFS	$O(V + E)$	Social Media
Google Maps Routing	Bellman-Ford	$O(VE)$	Navigation
Emergency Path Planning	Dijkstra's	$O(E \log V)$	Disaster Response
Cable Installation	MST (Prim/Kruskal)	$O(E \log V)$	Infrastructure