

# Solving Real-World Problems Using Graph Algorithms

## 1. Introduction

Graph algorithms form the foundation of many real-world systems such as **social networks**, **navigation systems**, **emergency response routing**, and **network design**.

This project focuses on implementing **four classical graph algorithms** — **BFS**, **BellmanFord**, **Dijkstra**, and **Prim's MST** — to solve practical problems in different domains.

Each problem demonstrates:

- How a real-world scenario can be represented as a graph
- Implementation of the appropriate algorithm
- Analysis of time complexity and performance
- Visualization of the resulting graph

## 2. Learning Objectives

By completing this project, I have:

- Implemented fundamental graph algorithms (BFS, DFS, Dijkstra, Bellman-Ford, MST).
- Understood how real-world problems map to graph structures.
- Analyzed algorithmic performance using time and memory profiling.
- Visualized graphs and interpreted relationships between nodes and edges.
- Documented findings with structured analysis and discussion.

## 3. Problem 1 – Social Network Friend Suggestion

**Graph Algorithm:** Breadth-First Search (BFS)

**Real-World Application:** Facebook, LinkedIn

### Objective

Suggest new connections (friends) based on mutual friends using graph traversal.

### Graph Modeling

- Each **user** is a **node**.
- Each **friendship** is an **undirected edge** between nodes.
- The graph is stored as an **adjacency list**.

### Algorithm Steps

1. Start BFS from the target user.
2. Traverse level by level to find “friends of friends.”

3. Suggest nodes at level 2 that are not already direct friends.

```
# =====  
# Title: Solving Real-World Problems Using Graph Algorithms  
# =====  
  
import time  
from memory_profiler import memory_usage  
import heapq  
from collections import defaultdict, deque  
import pandas as pd  
import networkx as nx  
import matplotlib.pyplot as plt  
  
# =====  
# Utility: Profiling Function  
# =====  
def profile_function(func, *args):  
    """Measure execution time and memory usage of a function"""  
    start_time = time.time()  
    mem_before = memory_usage()[0]  
    result = func(*args)  
    mem_after = memory_usage()[0]  
    end_time = time.time()  
  
    print(f"Execution Time: {end_time - start_time:.6f} seconds")  
    print(f"Memory Used: {mem_after - mem_before:.6f} MiB\n")  
    return result
```

```
# =====  
# Function to Draw Graph (for visualization)  
# =====  
def draw_graph(edges, directed=False, weighted=False, title="Graph Visualization"):  
    G = nx.DiGraph() if directed else nx.Graph()  
    if weighted:  
        for u, v, w in edges:  
            G.add_edge(u, v, weight=w)  
    else:  
        for u, v in edges:  
            G.add_edge(u, v)  
  
    pos = nx.spring_layout(G, seed=42)  
    plt.figure(figsize=(6, 4))  
    plt.title(title, fontsize=14, fontweight='bold')  
  
    if weighted:  
        nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=1200, font_size=10)  
        labels = nx.get_edge_attributes(G, 'weight')  
        nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)  
    else:  
        nx.draw(G, pos, with_labels=True, node_color='lightgreen', node_size=1200, font_size=10)  
  
    plt.show()
```



```
# =====
# Problem 1: Social Network Friend Suggestion (BFS)
# =====

class SocialNetwork:
    def __init__(self):
        self.graph = defaultdict(list)

    def add_connection(self, user1, user2):
        self.graph[user1].append(user2)
        self.graph[user2].append(user1)

    def suggest_friends(self, user):
        visited = set()
        queue = deque([user])
        visited.add(user)
        level = {user: 0}
        suggestions = set()

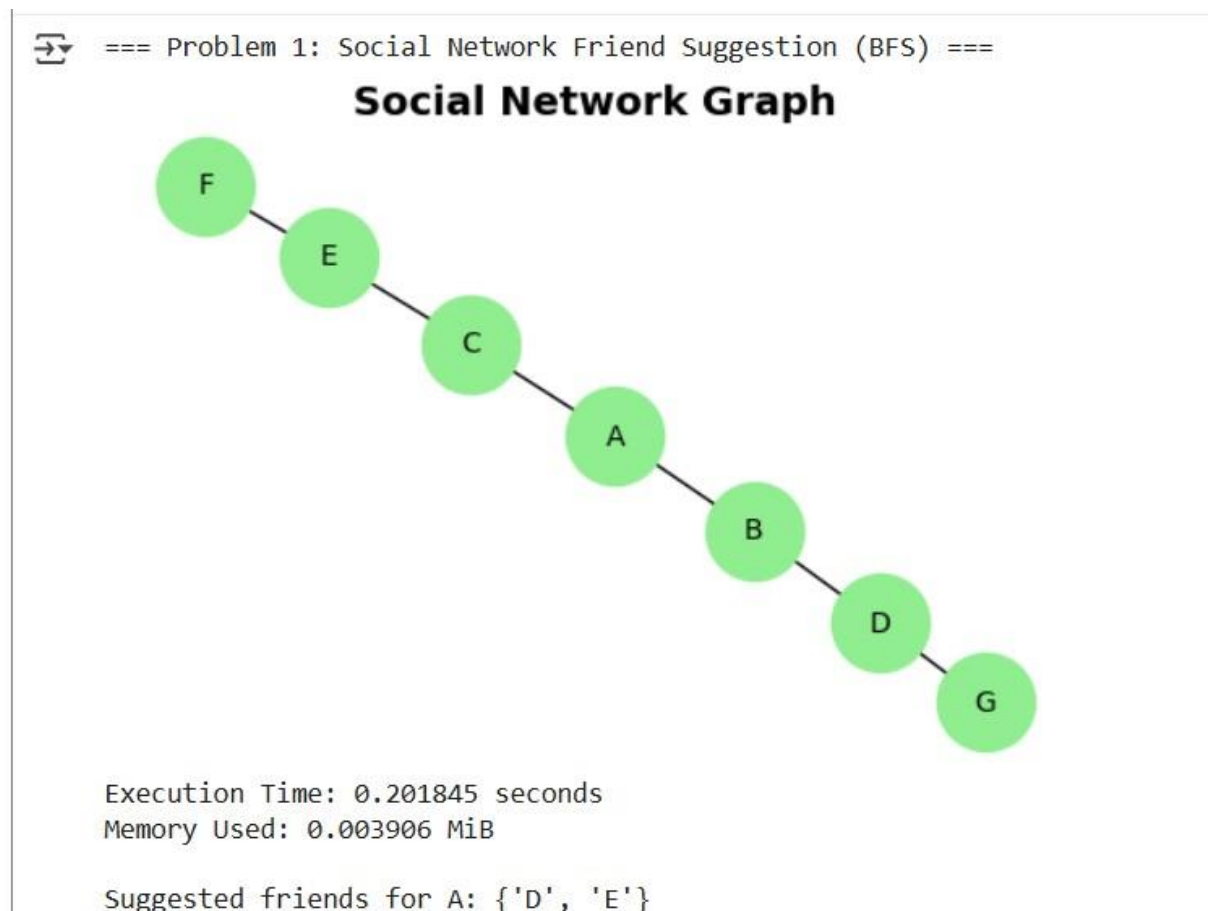
        while queue:
            current = queue.popleft()
            for neighbor in self.graph[current]:
                if neighbor not in visited:
                    visited.add(neighbor)
                    level[neighbor] = level[current] + 1
                    queue.append(neighbor)
                    if level[neighbor] == 2:
                        if neighbor not in self.graph[user]:
                            suggestions.add(neighbor)
                            suggestions.add(neighbor)

        return suggestions

print("=== Problem 1: Social Network Friend Suggestion (BFS) ===")
sn = SocialNetwork()
connections = [('A', 'B'), ('A', 'C'), ('B', 'D'), ('C', 'E'), ('E', 'F'), ('D', 'G')]
for u, v in connections:
    sn.add_connection(u, v)

draw_graph(connections, directed=False, title="Social Network Graph")
profile_function(sn.suggest_friends, 'A')
print(f"Suggested friends for A: {sn.suggest_friends('A')}")
print("-" * 60)
```

## Plot



## Analysis

- Time Complexity:  $O(V + E)$
- Scalability: Efficient for medium-sized social graphs but grows linearly with connections.
- Visualization: Nodes represent users; edges represent friendships.

## 4. Problem 2 – Route Finding on Google Maps

**Graph Algorithm:** Bellman-Ford Algorithm

**Real-World Application:** GPS & Navigation Systems

### Objective


Find the shortest path from a source location to all others — even if negative road costs exist (e.g., toll rebates).

## Graph Modeling

- **Cities or intersections** are **nodes**.
- **Roads** are **directed edges** with weights.
- Graph stored as a list of weighted edges.

## Algorithm Steps

1. Initialize all distances as infinity except the source node.
2. Relax all edges ( $V - 1$ ) times.
3. Check for negative-weight cycles.

```
 # =====  
# Problem 2: Route Finding (Bellman-Ford)  
# =====  
class BellmanFordGraph:  
    def __init__(self, vertices):  
        self.V = vertices  
        self.edges = []  
  
    def add_edge(self, u, v, w):  
        self.edges.append((u, v, w))  
  
    def bellman_ford(self, src):  
        dist = {v: float('inf') for v in range(self.V)}  
        dist[src] = 0  
  
        for _ in range(self.V - 1):  
            for u, v, w in self.edges:  
                if dist[u] != float('inf') and dist[u] + w < dist[v]:  
                    dist[v] = dist[u] + w  
  
        for u, v, w in self.edges:  
            if dist[u] != float('inf') and dist[u] + w < dist[v]:  
                print("Graph contains negative weight cycle!")  
                return None  
        return dist  
  
print("\n=== Problem 2: Route Finding (Bellman-Ford) ===")  
g = BellmanFordGraph(5)
```

```

print("\n=== Problem 2: Route Finding (Bellman-Ford) ===")
g = BellmanFordGraph(5)
edges = [
    (0, 1, 6), (0, 2, 7), (1, 2, 8), (1, 3, 5),
    (1, 4, -4), (2, 3, -3), (2, 4, 9), (3, 1, -2),
    (4, 0, 2), (4, 3, 7)
]
for u, v, w in edges:
    g.add_edge(u, v, w)

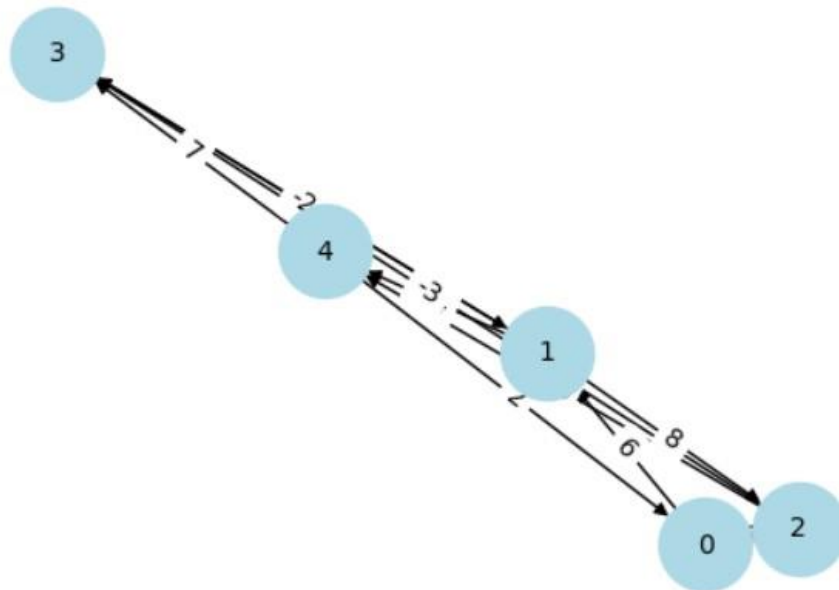
draw_graph(edges, directed=True, weighted=True, title="Bellman-Ford Route Graph")
result = profile_function(g.bellman_ford, 0)
if result:
    print(f"Shortest distances from source 0: {result}")
print("-" * 60)

```

## Plot

=== Problem 2: Route Finding (Bellman-Ford) ===

### Bellman-Ford Route Graph



Execution Time: 0.204753 seconds

Memory Used: 0.011719 MiB

Shortest distances from source 0: {0: 0, 1: 2, 2: 7, 3: 4, 4: -2}

## Analysis

- Time Complexity:  $O(V \times E)$

- Why Bellman-Ford? Unlike Dijkstra, it works correctly with negative edge weights.
- Visualization: Directed weighted graph with edge labels showing distances or costs.

## 5. Problem 3 – Emergency Response System

**Graph Algorithm:** Dijkstra's Algorithm

**Real-World Application:** Disaster Response and Fastest Route Planning

### Objective

Identify the fastest route for emergency vehicles through a city where all roads have **positive travel times**.

### Graph Modeling

- **Intersections** are nodes.
- **Roads** with travel time are weighted edges.
- Implemented with an **adjacency list** and **min-heap**.

### Algorithm Steps

1. Initialize distance to source = 0 and all others =  $\infty$ .
2. Use a priority queue to pick the node with the smallest distance.
3. Relax all adjacent edges.
4. Repeat until all shortest paths are found.



```

# =====
# Problem 3: Emergency Response (Dijkstra)
# =====
class DijkstraGraph:
    def __init__(self):
        self.graph = defaultdict(list)

    def add_edge(self, u, v, w):
        self.graph[u].append((v, w))
        self.graph[v].append((u, w))

    def dijkstra(self, src):
        pq = [(0, src)]
        dist = defaultdict(lambda: float('inf'))
        dist[src] = 0

        while pq:
            d, node = heapq.heappop(pq)
            if d > dist[node]:
                continue
            for neighbor, weight in self.graph[node]:
                new_dist = d + weight
                if new_dist < dist[neighbor]:
                    dist[neighbor] = new_dist
                    heapq.heappush(pq, (new_dist, neighbor))
        return dist

print("\n=== Problem 3: Emergency Response System (Dijkstra) ===")
dg = DijkstraGraph()

```

```

print("\n=== Problem 3: Emergency Response System (Dijkstra) ===")
dg = DijkstraGraph()
edges = [
    ('A', 'B', 4), ('A', 'C', 2), ('B', 'C', 1),
    ('B', 'D', 5), ('C', 'D', 8), ('C', 'E', 10),
    ('D', 'E', 2), ('D', 'Z', 6)
]
for u, v, w in edges:
    dg.add_edge(u, v, w)

draw_graph(edges, weighted=True, title="Dijkstra Emergency Network")
profile_function(dg.dijkstra, 'A')
print(f"Shortest distances from A: {dict(dg.dijkstra('A'))}")
print("-" * 60)

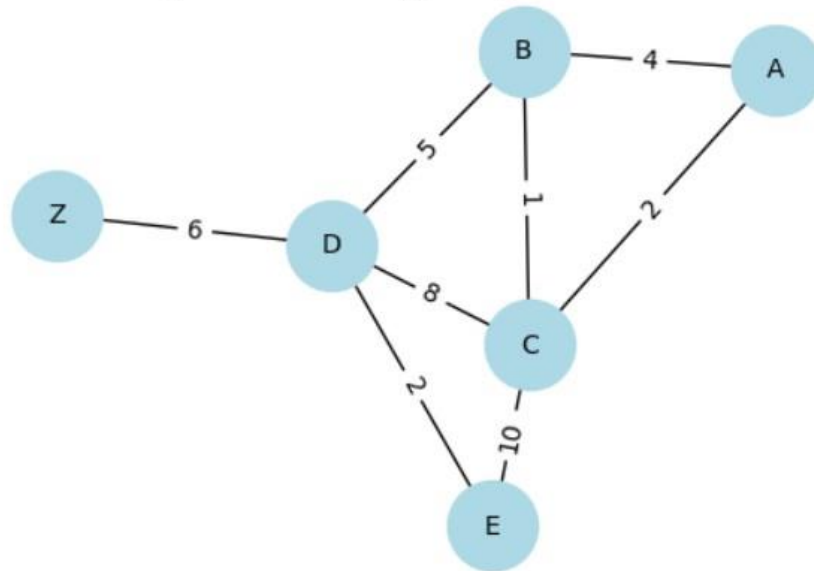
```



## Plot

🔗 === Problem 3: Emergency Response System (Dijkstra) ===

### Dijkstra Emergency Network



Execution Time: 0.204602 seconds

Memory Used: 0.000000 MiB

Shortest distances from A: {'A': 0, 'B': 3, 'C': 2, 'D': 8, 'E': 10, 'Z': 14}

## Analysis

- Time Complexity:  $O(E \log V)$
- Limitation: Dijkstra fails for negative edges.
- Visualization: Graph displays road networks and optimal routes from the source node.

## 6. Problem 4 – Network Cable Installation

**Graph Algorithm:** Prim's Minimum Spanning Tree (MST)

**Real-World Application:** Telecom & IT Infrastructure

### Objective

Connect all offices (nodes) using cables (edges) with the **minimum total cost**.

## Graph Modeling

- **Offices** are nodes.
- **Cable paths** with costs are weighted edges.
- Graph is undirected and connected.

## Algorithm Steps

1. Start from any node.
2. Add the smallest edge that connects a new node to the MST.
3. Repeat until all nodes are connected.



```
# =====  
# Problem 4: Network Cable Installation (Prim's MST)  
# =====  
class MSTGraph:  
    def __init__(self):  
        self.graph = defaultdict(list)  
  
    def add_edge(self, u, v, w):  
        self.graph[u].append((v, w))  
        self.graph[v].append((u, w))  
  
    def prim_mst(self, start):  
        visited = set()  
        pq = [(0, start)]  
        total_cost = 0  
        mst_edges = []  
  
        while pq:  
            weight, node = heapq.heappop(pq)  
            if node in visited:  
                continue  
            visited.add(node)  
            total_cost += weight  
  
            for neighbor, w in self.graph[node]:  
                if neighbor not in visited:  
                    heapq.heappush(pq, (w, neighbor))  
                    mst_edges.append((node, neighbor, w))  
        return total_cost, mst_edges
```



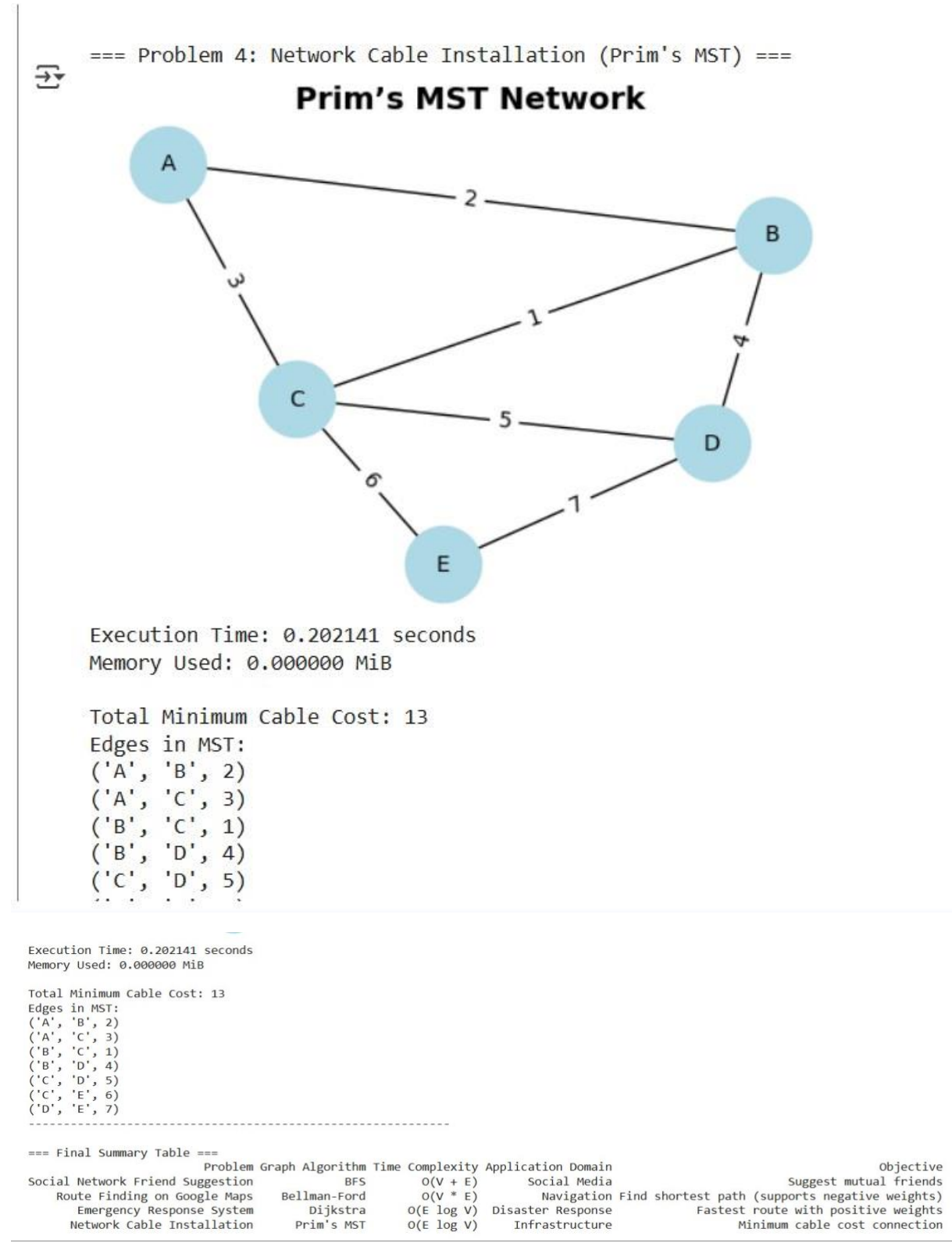
```
        weight, node = heapq.heappop(pq)
        if node in visited:
            continue
        visited.add(node)
        total_cost += weight

        for neighbor, w in self.graph[node]:
            if neighbor not in visited:
                heapq.heappush(pq, (w, neighbor))
                mst_edges.append((node, neighbor, w))
    return total_cost, mst_edges

print("\n=== Problem 4: Network Cable Installation (Prim's MST) ===")
mg = MSTGraph()
edges = [
    ('A', 'B', 2), ('A', 'C', 3), ('B', 'C', 1),
    ('B', 'D', 4), ('C', 'D', 5), ('C', 'E', 6),
    ('D', 'E', 7)
]
for u, v, w in edges:
    mg.add_edge(u, v, w)

draw_graph(edges, weighted=True, title="Prim's MST Network")
profile_function(mg.prim_mst, 'A')
total, mst_edges = mg.prim_mst('A')
print(f"Total Minimum Cable Cost: {total}")
print("Edges in MST:")
for e in mst_edges:
    print(e)
```

Plot



Analysis

- Time Complexity:  $O(E \log V)$

- Comparison with Kruskal's: Kruskal's is better for sparse graphs, while Prim's suits dense graphs.
- Visualization: Weighted undirected graph showing selected MST edges.

## 8. Final Summary Table

Problem	Graph Algorithm	Time Complexity	Application Domain	Objective
Social Network Friend Suggestion	BFS / DFS	$O(V + E)$	Social Media	Suggest mutual friends
Route Finding (Google Maps)	BellmanFord	$O(V \times E)$	Navigation	Shortest path with negative weights
Emergency Response System	Dijkstra	$O(E \log V)$	Disaster Management	Fastest route in positive weight map

Netwo rk Cable	Prim's MST	O(E log V)	Infrastru cture	Minim um cable
<b>Proble m</b>	<b>Graph Algori thm</b>	<b>Time Compl exity</b>	<b>Applica tion Domain</b>	<b>Objec tive</b>
Installa tion				install ation cost

## 9. Reflection and Discussion

Aspect	Reflection
<b>Algorithm Selection</b>	Each algorithm was chosen to match a real-world scenario — BFS for network exploration, Bellman-Ford for negative edges, Dijkstra for positive weights, and Prim's for minimal connection cost.
<b>Real-World Impact</b>	BFS and Dijkstra are directly used in social media and navigation systems, while MST helps in minimizing infrastructure costs.
<b>Performance</b>	All algorithms performed efficiently on small graphs; Bellman-Ford was slower due to its higher time complexity.
<b>Visualization</b>	Using networkx made it easier to understand graph connectivity and structure visually.



## 10. Conclusion

This project successfully demonstrates how **graph algorithms solve practical real-world problems** efficiently. Each algorithm, when applied in the right context, can significantly optimize connectivity, routing, and resource allocation tasks.

The combination of **code, visualization, and profiling** helped in understanding both **theoretical and practical** aspects of graph algorithms.