

Graph Theory Problem Set - Solutions

Comprehensive Solutions with Explanations

Section A: Multiple Choice Questions - Solutions

Q1) Select all statements that are TRUE about adjacency matrix representation:

Correct Answers: A, B, E

A) TRUE: An adjacency matrix for V vertices requires $V \times V$ space, hence $O(V^2)$.

B) TRUE: To check if edge (i,j) exists, we simply access $matrix[i][j]$, which is $O(1)$.

C) FALSE: Adding a new vertex requires resizing the matrix from $V \times V$ to $(V+1) \times (V+1)$, which takes $O(V^2)$ time to copy all elements.

D) FALSE: For sparse graphs (few edges), adjacency list $O(V+E)$ is better than adjacency matrix $O(V^2)$ when $E \ll V^2$.

E) TRUE: In an undirected graph, if there's an edge between i and j , then $matrix[i][j] = matrix[j][i] = 1$, making it symmetric.

Q2) What is the time complexity of BFS traversal on a graph with V vertices and E edges represented using an adjacency list?

Correct Answer: c) $O(V + E)$

BFS visits each vertex once ($O(V)$) and explores each edge once in undirected graphs or exactly once in directed graphs ($O(E)$). Total: $O(V + E)$.

Q3) For the adjacency matrix of an undirected graph, what can be said about the diagonal elements?

Correct Answer: a) They are always 0 unless there are self-loops

Diagonal elements $matrix[i][i]$ represent edges from vertex i to itself. In graphs without self-loops, these are 0. If self-loops exist, they would be 1 (or the weight of the self-loop).

Q4) In DFS traversal of a graph, which data structure is implicitly used?

Correct Answer: b) Stack

DFS uses a stack either explicitly (iterative implementation) or implicitly through recursion (call stack). It explores as deep as possible before backtracking.

Q5) Which of the following statements about Dijkstra's algorithm are TRUE?

Correct Answers: B, C, D

A) FALSE: Dijkstra's algorithm fails with negative weights as it assumes once a vertex is finalized, no shorter path exists.

B) TRUE: Dijkstra finds shortest paths from source to all reachable vertices.

C) TRUE: A min-heap/priority queue efficiently extracts the vertex with minimum distance.

D) TRUE: With binary heap: $O(V)$ extract-min operations at $O(\log V)$ each, and $O(E)$ decrease-key operations at $O(\log V)$ each = $O((V + E) \log V) = O(E \log V)$ for connected graphs.

E) FALSE: Dijkstra cannot detect negative cycles; Bellman-Ford is needed for that.

Q6) What is the maximum number of edges in a complete undirected graph with n vertices?

Correct Answer: c) $n(n-1)/2$

*In a complete undirected graph, every vertex connects to every other vertex.
Using combination formula: $C(n,2) = n!/(2!(n-2)!) = n(n-1)/2$.*

Q7) Which algorithm(s) can be used to find a Minimum Spanning Tree?

Correct Answers: A, B

A) TRUE: Kruskal's algorithm sorts edges and adds them if they don't create cycles.

B) TRUE: Prim's algorithm grows the MST from a starting vertex.

C) FALSE: Dijkstra finds shortest paths, not MST.

D) FALSE: Floyd-Warshall finds all-pairs shortest paths.

E) FALSE: Bellman-Ford finds shortest paths with negative weights.

Q8) In BFS traversal, nodes are visited in what order relative to the source?

Correct Answer: c) Increasing order of their distance from source

BFS explores level by level, visiting all vertices at distance k before any vertex at distance $k+1$ from the source.

Q9) A Minimum Spanning Tree of a graph with V vertices must have how many edges?

Correct Answer: b) $V - 1$

A spanning tree connects all V vertices with the minimum number of edges without cycles. A tree with V vertices always has exactly $V-1$ edges.

Q10) Which property must a graph satisfy for Kruskal's algorithm to work?

Correct Answer: b) Graph must be connected

For an MST to exist, the graph must be connected (or we find a minimum spanning forest). The graph can be directed/undirected, have any weights (positive/negative), and edges need not be unique. However, for practical MST, we typically work with connected, undirected, weighted graphs.

Section B: Short Answer Questions - Solutions

Q11) Graph representation:

a) Adjacency Matrix:

| | A | B | C | D | E |
|---|---|---|----|---|----|
| A | 0 | 4 | 2 | 0 | 0 |
| B | 4 | 0 | 1 | 5 | 0 |
| C | 2 | 1 | 0 | 8 | 10 |
| D | 0 | 5 | 8 | 0 | 2 |
| E | 0 | 0 | 10 | 2 | 0 |

b) Adjacency List:

A: [(B, 4), (C, 2)]
B: [(A, 4), (C, 1), (D, 5)]
C: [(A, 2), (B, 1), (D, 8), (E, 10)]
D: [(B, 5), (C, 8), (E, 2)]
E: [(C, 10), (D, 2)]

c) Difference if directed:

If the graph were directed (from first to second vertex):

- The adjacency matrix would NOT be symmetric
- Each edge would appear only once in the adjacency list

Example: Edge A → B (weight 4) would mean:

- Matrix: $\text{matrix}[A][B] = 4$, but $\text{matrix}[B][A] = 0$
- List: A would have (B,4) but B would NOT have (A,4)

Q12) Adjacency matrix analysis:

a) Graph drawing:

```
A ---- B
|      |
|      |
C ---- D
```

Edges: A-B, A-C, B-D, C-D (forming a square/cycle)

b) Directed or Undirected:

The graph is **UNDIRECTED**. We can tell because the matrix is symmetric: •
 $\text{matrix}[A][B] = \text{matrix}[B][A] = 1$ • $\text{matrix}[A][C] = \text{matrix}[C][A] = 1$ • $\text{matrix}[B][D] = \text{matrix}[D][B] = 1$ • $\text{matrix}[C][D] = \text{matrix}[D][C] = 1$

c) Degree of each vertex:

Degree = sum of row (or column, since symmetric) • $\text{deg}(A) = 1 + 1 = 2$ • $\text{deg}(B) = 1 + 1 = 2$ • $\text{deg}(C) = 1 + 1 = 2$ • $\text{deg}(D) = 1 + 1 = 2$

Q13) BFS and DFS traversals:

a) BFS starting from A:

Order: A B C D E F

Level 0: A

Level 1: B, C (neighbors of A)

Level 2: D, E (from B), F (from C)

F is already discovered via C, so not repeated

b) DFS starting from A:

Order: A B D E F C

A B (first neighbor of A)

B D (first unvisited neighbor of B)

D has no unvisited neighbors, backtrack

B E (next unvisited neighbor of B)

E F (only unvisited neighbor of E)

Backtrack to A C (last unvisited neighbor of A)

Q14) Dijkstra's algorithm:

a) Order of finalized vertices:

S A B C D T

| Step | Finalized | Distances [S,A,B,C,D,T] |
|------|-----------|--|
| 0 | - | [0, ∞ , ∞ , ∞ , ∞ , ∞] |
| 1 | S | [0, 2, 4, ∞ , ∞ , ∞] |
| 2 | A | [0, 2, 3, 9, ∞ , ∞] |
| 3 | B | [0, 2, 3, 6, 8, ∞] |
| 4 | C | [0, 2, 3, 6, 8, 8] |
| 5 | D | [0, 2, 3, 6, 8, 8] |
| 6 | T | [0, 2, 3, 6, 8, 8] |

b) Shortest distances from S:

S: 0, A: 2, B: 3, C: 6, D: 8, T: 8

c) Shortest path S to T:

Path: S A B C T

Total distance: $2 + 1 + 3 + 2 = 8$

Alternative path with same length: S A B D T

Total distance: $2 + 1 + 5 + 1 = 9$ (longer, so not chosen)

Q15) Kruskal's algorithm for MST:

a) Edges considered in order (sorted by weight):

1. B-C: 2
2. A-B: 3
3. D-E: 3
4. B-D: 4
5. E-F: 4
6. A-C: 5
7. C-D: 6
8. D-F: 7
9. C-E: 8

b) Edges included in MST:

| Edge | Weight | Action |
|------|--------|--------------------------------------|
| B-C | 2 | Add (connects B, C) |
| A-B | 3 | Add (connects A) |
| D-E | 3 | Add (connects D, E) |
| B-D | 4 | Add (connects component ABC with DE) |
| E-F | 4 | Add (connects F) |
| A-C | 5 | Skip (creates cycle in ABC) |
| C-D | 6 | Skip (already connected) |

MST Edges: B-C, A-B, D-E, B-D, E-F

c) Total weight of MST:

Total = 2 + 3 + 3 + 4 + 4 = 16

Section C: Long Answer Questions - Solutions

Q16) True or False with justification:

1. **TRUE** - Sum of degrees equals $2E$ (each edge contributes 2 to the total degree sum).
2. **FALSE** - BFS finds shortest path only in *unweighted* graphs. For weighted graphs, use Dijkstra's or Bellman-Ford.
3. **TRUE** - If all edge weights are distinct, no two spanning trees can have the same total weight. The unique minimum weight determines a unique MST.
4. **TRUE** - DFS can detect cycles: if we encounter an already visited vertex (that's not the parent), a cycle exists.
5. **FALSE** - Dijkstra assumes all edges are non-negative. Negative weights can lead to incorrect results because the algorithm finalizes vertices assuming no shorter path exists.
6. **FALSE** - The graph must also be *connected* to be a tree. Example: Two separate edges (4 vertices, 2 edges forming two components) is not a tree.
7. **TRUE** - Prim's with binary heap: $O(V)$ extract-min at $O(\log V) + O(E)$ decrease-key at $O(\log V) = O(E \log V)$.
8. **FALSE** - BFS typically uses more memory because it stores all vertices at current level in the queue, which can be $O(V)$ in worst case. DFS uses $O(h)$ where h is height/depth.
9. **TRUE** - Every connected graph has at least one spanning tree (can be constructed using DFS or BFS).
10. **FALSE** - In adjacency list, checking if edge (u,v) exists requires traversing u 's adjacency list, which takes $O(\text{degree}(u))$ time, or $O(V)$ in worst case.

Q17) Connected Components Algorithm:

Algorithm:

```
function countConnectedComponents(n, edges):
    # Build adjacency list
    graph = [[] for _ in range(n)]
    for (u, v) in edges:
        graph[u].append(v)
        graph[v].append(u)

    visited = [False] * n
    count = 0

    # Run DFS/BFS from each unvisited vertex
    for vertex in range(n):
        if not visited[vertex]:
            DFS(vertex, graph, visited)
            count += 1

    return count

function DFS(vertex, graph, visited):
    visited[vertex] = True
    for neighbor in graph[vertex]:
        if not visited[neighbor]:
            DFS(neighbor, graph, visited)
```

Explanation:

1. Build adjacency list representation: $O(E)$
2. Initialize visited array: $O(V)$
3. For each unvisited vertex, perform DFS/BFS to mark all vertices in that component
4. Each DFS call marks one connected component
5. Count increments for each new component discovered

Time Complexity: $O(V + E)$

- We visit each vertex once: $O(V)$
- We examine each edge once: $O(E)$
- Total: $O(V + E)$

Space Complexity: $O(V + E)$

- Adjacency list: $O(V + E)$
- Visited array: $O(V)$
- Recursion stack (worst case): $O(V)$

Q18) Path Existence Algorithm:

Algorithm: Use BFS or DFS

```
function pathExists(graph, source, destination):  
    if source == destination:  
        return True  
  
    visited = set()  
    queue = [source] # For BFS  
    visited.add(source)  
  
    while queue:  
        current = queue.pop(0)  
  
        if current == destination:  
            return True  
  
        for neighbor in graph[current]:  
            if neighbor not in visited:  
                visited.add(neighbor)  
                queue.append(neighbor)  
  
    return False
```

Why BFS/DFS?

Either BFS or DFS works equally well:

BFS Advantages:

- Finds shortest path (in terms of edges) if needed
- Better for finding nearby destinations

DFS Advantages:

- Uses less memory ($O(h)$ vs $O(w)$ where h =height, w =width)
- Easier to implement recursively

Time Complexity: $O(V + E)$

- We may need to explore all vertices and edges in worst case

Space Complexity:

- BFS: $O(V)$ for queue
- DFS: $O(h)$ for recursion stack, where h is maximum depth

Q19) Comparison Questions:

a) Adjacency Matrix vs Adjacency List:

| Aspect | Adjacency Matrix | Adjacency List |
|----------------------|------------------|---------------------------------|
| Space Complexity | $O(V^2)$ | $O(V + E)$ |
| Add Edge | $O(1)$ | $O(1)$ |
| Remove Edge | $O(1)$ | $O(V)$ worst case |
| Check if edge exists | $O(1)$ | $O(\text{degree}) = O(V)$ worst |
| Find all neighbors | $O(V)$ | $O(\text{degree})$ |

When to use Adjacency Matrix:

- Dense graphs ($E \approx V^2$)
- Need $O(1)$ edge existence checks
- Fixed size graph
- Memory is not a constraint

When to use Adjacency List:

- Sparse graphs ($E \ll V^2$)
- Need to iterate over neighbors efficiently
- Dynamic graphs (adding/removing vertices)
- Memory is limited

b) BFS vs DFS:

| Aspect | BFS | DFS |
|------------------|-----------------------|-----------------------------|
| Data Structure | Queue | Stack (or recursion) |
| Traversal Order | Level by level | Depth-first, then backtrack |
| Space Complexity | $O(w)$ w =max width | $O(h)$ h =max depth |
| Shortest Path | Yes (unweighted) | No |
| Memory Usage | Higher (wide graphs) | Lower (deep graphs) |

BFS Applications:

- *Finding shortest path in unweighted graphs*
- *Level-order traversal*
- *Finding all nodes within k distance*
- *Testing bipartiteness*

DFS Applications:

- *Topological sorting*
- *Cycle detection*
- *Finding connected components*
- *Path finding (any path, not shortest)*
- *Solving mazes/puzzles*

When to prefer BFS:

- *Need shortest path*
- *Target is likely near source*
- *Graph is wide and shallow*

When to prefer DFS:

- *Don't need shortest path*
- *Graph is deep and narrow*
- *Want to use less memory*
- *Need to explore all paths*

Given Graph Edges: A-B:7, A-D:5, B-C:8, B-D:9, B-E:7, C-E:5, D-E:15, D-F:6, E-F:8, E-G:9, F-G:11

Step 1: Start from A

MST = {}, Priority Queue = [(B,7), (D,5)]

Choose D (minimum weight 5)

Add edge: A-D (weight 5)

Step 2: From D, add neighbors

MST = {A-D}, PQ = [(B,7), (B,9), (E,15), (F,6)]

Choose F (minimum weight 6)

Add edge: D-F (weight 6)

Step 3: From F, add neighbors

MST = {A-D, D-F}, PQ = [(B,7), (B,9), (E,15), (E,8), (G,11)]

Choose B (minimum weight 7)

Add edge: A-B (weight 7)

Step 4: From B, add neighbors

MST = {A-D, D-F, A-B}, PQ = [(B,9), (E,15), (E,8), (G,11), (C,8), (E,7)]

Choose E via B (minimum weight 7)

Add edge: B-E (weight 7)

Step 5: From E, add neighbors

MST = {A-D, D-F, A-B, B-E}, PQ = [(B,9), (E,15), (E,8), (G,11), (C,8), (C,5), (F,8), (G,9)]

Choose C (minimum weight 5)

Add edge: C-E (weight 5)

Step 6: From C, no new neighbors

MST = {A-D, D-F, A-B, B-E, C-E}, PQ = [(B,9), (E,15), (E,8), (G,11), (F,8), (G,9)]

Choose G via E (minimum weight 9)

Add edge: E-G (weight 9)

Final MST Edges:

1. A-D (5)
2. D-F (6)
3. A-B (7)
4. B-E (7)

5. C-E (5)

6. E-G (9)

Total Weight: $5 + 6 + 7 + 7 + 5 + 9 = 39$

The MST connects all 7 vertices (A, B, C, D, E, F, G) with 6 edges.

End of Solutions