

case: City Management

Sub-task: Measures to be taken for safety of community

SDG Goal: 11 Business

Target: 11.7

Indicator: 11.7

Overview:

Tourism is a cornerstone of Vatsalya Nagar's city plan, emphasizing cultural heritage, recreational activities, and sustainable development. The city's layout integrates parks, resorts, and recreational zones celebrating local traditions. These efforts aim to:

- Drive economic growth.
- Encourage social interaction.
- Position Vatsalya Nagar as a cultural and recreational hub.



Alignment with SDG 11.3:

This initiative supports SDG 11.3 by promoting inclusive and sustainable urbanization through participatory planning and management, ensuring a well-integrated and sustainable urban environment for both residents and visitors.

Tourist Hub	Node ID	Connections	Travel cost (Weight)
Central Park	A	B,C,D	5,3,6
Riverside Resort	B	A,C	5,4
Heritage Museum	C	A,B,D	3,4,2
Green Retreat	D	A,C	6,2

Explanation of Columns:

- Tourist Hub: Major tourist attractions in Vatsalya Nagar.
- Node ID: Represents the attraction as a node in the graph.
- Connections: Indicates connected tourist hubs.
- Travel Cost (Weight): Cost of traveling between hubs (e.g., in terms of time or fuel consumption).

Algorithms for Tourism Optimization

1. Dijkstra's Algorithm

Use Case:

Optimize tourist routes by calculating the shortest path between attractions to minimize travel time and fuel consumption.

Steps:

1. Set the starting point (e.g., a tourist's hotel or current location).
2. Assign initial distances to all nodes (infinity for unvisited nodes, 0 for the starting node).
3. Update distances for all neighbors of the current node.
4. Mark the node as visited and repeat for the next nearest unvisited node.
5. Continue until the destination (tourist site) is reached.

Application in Vatsalya Nagar:

- Nodes: Tourist attractions, parks, resorts.
- Edges: Roads with weights representing travel time or fuel cost.
- Output: Optimal tourist route.

2. Breadth-First Search (BFS)

Use Case:

Map connectivity between tourist destinations and recommend guided tour sequences.

Steps:

1. Start from a node (e.g., a key attraction).
2. Traverse all neighbouring nodes.
3. Mark visited nodes to avoid redundancy.
4. Generate a sequence of connected attractions.

Application in Vatsalya Nagar:

- Nodes: Tourist attractions.
- Edges: Connections or paths between attractions.
- Output: Guided tour sequences optimizing visitation order.

3. Prim's Algorithm (for Minimum Spanning Tree)

Use Case:

Optimize the development of infrastructure by connecting all tourist hubs with the minimum total cost.

Steps:

1. Start with any node as the root of the tree.
2. Add the smallest edge that connects a node in the tree to a node outside it.
3. Repeat until all nodes are connected.

Application in Vatsalya Nagar:

- Nodes: Tourist attractions.
- Edges: Travel paths with weights (time, cost, or distance).
- Output: Minimum infrastructure cost to connect all hubs efficiently

4. Depth-First Search (DFS)

- Use Case:
Explore possible paths between tourist hubs for creating exploratory or adventurous tour routes.
- Steps:
 - Start at a node and mark it as visited.
 - Recursively visit all unvisited neighbors.
 - Backtrack when no unvisited neighbors remain.
- Application in Vatsalya Nagar:
 - Nodes: Tourist attractions.

- Edges: Connections between locations.
- Output: All possible paths, useful for planning exploratory tours.

5. Floyd-Warshall Algorithm

- Use Case:
Identify shortest paths between all pairs of tourist hubs for comprehensive travel route planning.
- Steps:
 - Initialize a matrix where $\text{dist}[i][j]$ represents the direct distance between nodes i and j .
 - For each node k , update the matrix as $\text{dist}[i][j] = \min(\text{dist}[i][j], \text{dist}[i][k] + \text{dist}[k][j])$.
 - Repeat for all nodes to compute shortest paths.
- Application in Vatsalya Nagar:
- Nodes: Tourist hubs.
- Edges: Travel paths with weights.
- Output: Matrix of shortest distances between all tourist hubs.

□

Code Implementations:

Dijkstra's Algorithm:

```
#include <iostream>
```

```
#include <vector>
```

```
#include <queue>
```

```
#include <limits>
```

```
using namespace std;
```

```
void dijkstra(int src, vector<vector<pair<int, int>>>& graph, vector<int>&
dist) {    priority_queue<pair<int, int>, vector<pair<int, int>>, greater<>> pq;
pq.push({0, src});    dist[src] = 0;
```

```
    while (!pq.empty()) {
        int d = pq.top().first;
```

```

        int u = pq.top().second;
pq.pop();

        if (d > dist[u]) continue;

        for (auto& edge : graph[u]) {            int
v = edge.first, weight = edge.second;
        if (dist[u] + weight < dist[v]) {
            dist[v] = dist[u] + weight;
            pq.push({dist[v], v});
        }
    }
}

int main() {    int V = 5;
vector<vector<pair<int, int>>> graph(V);

    // Example edges: Add connections with weights (travel times)
graph[0].push_back({1, 10});    graph[1].push_back({2, 5});
graph[2].push_back({3, 8});    graph[3].push_back({4, 3});
graph[0].push_back({4, 15});

    vector<int> dist(V, INT_MAX);
dijkstra(0, graph, dist);

    cout << "Shortest distances from the source:\n";
    for (int i = 0; i < V; i++) {
        cout << "Node " << i << ": " << dist[i] << endl;
    }
    return 0;
}

```

```
}
```

2. BFS for Connectivity Mapping:

```
#include <iostream>
```

```
#include <vector>
```

```
#include <queue>
```

```
using namespace std;
```

```
void bfs(int start, vector<vector<int>>& adj, vector<int>& visited) {
```

```
    queue<int> q;
```

```
    q.push(start);
```

```
    visited[start] = 1;
```

```
    cout << "Tourist sequence: ";
```

```
    while (!q.empty()) {        int
```

```
    node = q.front();
```

```
        q.pop();        cout
```

```
<< node << " ";
```

```
        for (int neighbor : adj[node]) {
```

```
            if (!visited[neighbor]) {
```

```
                visited[neighbor] = 1;
```

```
                q.push(neighbor);
```

```
            }
```

```
        }
```

```
    }
```

```
    cout << endl;
```

```
}
```

```

int main() {    int V = 5;
vector<vector<int>> adj(V);

    // Example connections between a nodes
    adj[0].push_back(1);
adj[1].push_back(2);
adj[2].push_back(3);
adj[3].push_back(4);

    vector<int> visited(V, 0);
bfs(0, adj, visited);

    return 0;
}
3.Prim's Algorithm:
#include <iostream>
#include <vector>
#include <queue>
using namespace std;

void prim(int V, vector<vector<pair<int, int>>>& graph) {
vector<int> key(V, INT_MAX);    vector<bool> inMST(V, false);
vector<int> parent(V, -1);    priority_queue<pair<int, int>,
vector<pair<int, int>>, greater<>> pq;

    pq.push({0, 0});
key[0] = 0;    while
(!pq.empty()) {        int
u = pq.top().second;
pq.pop();
inMST[u] = true;

```

```

        for (auto& edge : graph[u]) {            int
v = edge.first, weight = edge.second;
if (!inMST[v] && weight < key[v]) {
key[v] = weight;            pq.push({key[v],
v});            parent[v] = u;
        }
    }
}
}

```

```

    cout << "Edges in the MST:\n";
    for (int i = 1; i < V; i++) {        cout
<< parent[i] << " - " << i << "\n";
    }
}

```

```

int main() {    int V = 4;
vector<vector<pair<int, int>>> graph(V);

```

```

    graph[0] = {{1, 5}, {2, 3}, {3,
6}};    graph[1] = {{0, 5}, {2, 4}};
graph[2] = {{0, 3}, {1, 4}, {3, 2}};
graph[3] = {{0, 6}, {2, 2}};
prim(V, graph);    return 0;
}

```

4.Depth First Search

```

#include <iostream>
#include <vector>
#include <stack>

```

```

using namespace std;

```



```

// Perform Depth-First Search void dfs(int start,
vector<vector<int>>& adj, vector<bool>& visited) {    stack<int>
s;

    s.push(start);

    cout << "Tourist sequence (DFS):
";    while (!s.empty()) {        int node
= s.top();
        s.pop();

        if (!visited[node]) {
cout << node << " ";
visited[node] = true;
        }

        // Add unvisited neighbors to the
stack        for (int neighbor : adj[node]) {
if (!visited[neighbor]) {
            s.push(neighbor);
        }
    }
}

    cout << endl;
}

int main() {    int V = 4; // Number
of tourist hubs
vector<vector<int>> adj(V);

```

```

// Example connections between locations (graph edges)   adj[0] = {1, 2}; //
Central Park connected to Riverside Resort and Heritage Museum   adj[1] = {0, 2};
// Riverside Resort connected to Central Park and Heritage Museum
adj[2] = {0, 1, 3}; // Heritage Museum connected to Central Park, Riverside Resort, and
Green Retreat   adj[3] = {2}; // Green Retreat connected to Heritage Museum

```

```

vector<bool> visited(V, false);

// Perform DFS starting from node 0 (Central Park)
dfs(0, adj, visited);

return 0;
}

```

5. Floyd-Warshall Algorithm

```

#include <iostream>
#include <vector>
#include <climits>
using namespace std;

void floydWarshall(vector<vector<int>>& graph) {
    int V = graph.size();
    vector<vector<int>> dist = graph;    for
    (int k = 0; k < V; k++) {        for (int i =
    0; i < V; i++) {            for (int j = 0; j <
    V; j++) {                if (dist[i][k] !=
    INT_MAX && dist[k][j] != INT_MAX)
    {                    dist[i][j] = min(dist[i][j],
    dist[i][k] + dist[k][j]);
                }
            }
        }
    }
}

```

```

    }

    cout << "Shortest distances between every pair of nodes:\n";
    for (int i = 0; i < V; i++) {
    for (int j = 0; j < V; j++) {
    if (dist[i][j] == INT_MAX)
    cout << "INF ";      else
    cout << dist[i][j] << " ";
        }
        cout << endl;
    }
}

```

```

int main() {
vector<vector<int>>> graph = {
    {0, 5, INT_MAX,
6},    {5, 0, 4,
INT_MAX},
{INT_MAX, 4, 0, 2},
    {6, INT_MAX, 2, 0}
};

    floydWarshall(graph);
return 0;
}

```

Conclusion:

- Dijkstra's Algorithm: Provides optimal tourist routes, reducing travel time and enhancing visitor experience.

- BFS: Ensures seamless connectivity and planned guided tours.

Prim's Algorithm

- Purpose: Constructs a minimum spanning tree to connect all locations with minimal total travel cost.
- Advantages: Ensures all hubs are connected optimally without cycles.
- Use Case in Vatsalya Nagar: Helps design efficient infrastructure connecting all major hubs.

Depth-First Search (DFS)

- Purpose: Explores connected locations by diving deeper into one path before backtracking.
- Advantages: Useful for finding paths and exploring connectivity.
- Use Case in Vatsalya Nagar: Provides an alternative guided tour sequence focusing on depth-first exploration.
- Floyd-Warshall Algorithm
- Purpose: Finds the shortest paths between all pairs of locations.
- Advantages: Comprehensive and ideal for smaller graphs.
- Use Case in Vatsalya Nagar: Analyzes overall accessibility between hubs for strategic planning.

□

Business case: City Management

Sub-task: Land Leasing and Real Estate

SDG Goal: 11

Target: 11.3

Indicator: 11.3

Overview: Land leasing and real estate development form a significant part of Vatsalya Nagar's economic framework, providing a steady revenue stream and enabling sustainable urban expansion. By leasing land to private entities for residential, commercial, and industrial purposes, the city aims to balance economic growth with inclusivity. Real estate initiatives include affordable housing schemes, ensuring equitable access to homes for all income groups while fostering a vibrant community. This approach creates opportunities for investment and employment, enhancing the city's appeal as a destination for businesses and residents alike.



SDG Goal Alignment: SDG 11.3: Enhance inclusive and sustainable urbanization and capacity for participatory, integrated, and sustainable human settlement planning and management in all countries.

Algorithms and Their Use Cases:

1. Kruskal's Algorithm: Optimize Utility Networks

- o Use Case: Optimize land parcel connections for utilities such as electricity, water, and internet by building a minimal cost-spanning network. This ensures cost-effective and efficient service delivery across residential, commercial, and industrial zones.

Steps:

1. Sort all edges (land parcel connections) by increasing weight (utility costs).
2. Add the smallest edge to the network, ensuring it doesn't form a cycle.
3. Repeat until all parcels are connected in a minimal spanning tree.

- o Graph Representation:

- ☐ Nodes: Land parcels.
- ☐ Edges: Utility connections with weights representing costs

2 □ Union-Find Data Structure: Manage Land Ownership

- Use Case: Efficiently manage land ownership and lease agreements by dynamically linking owner-user relationships. This ensures real-time tracking of land utilization and helps prevent disputes.
- Steps:
 - Assign each land parcel to an initial owner (union operation).
 - For lease agreements, dynamically connect the owner to the user (find operation).
 - Update changes in ownership or leasing status efficiently.
- Graph Representation:
 - Nodes: Land parcels, owners, users.
 - Edges: Relationships (ownership, leasing agreements).

3. Prim's Algorithm: Build Resilient Infrastructure Networks

- Use Case: Construct cost-effective and resilient infrastructure networks (e.g., roads, pipelines) by minimizing the total cost of connecting various urban areas.
- Steps:
 - Start from an initial land parcel (node).
 - Add the smallest weight edge to connect a new parcel to the network.
 - Repeat until all parcels are connected.
- Graph Representation:
 - Nodes: Urban areas.
 - Edges: Infrastructure connections with weights representing costs.

4. Bellman-Ford Algorithm: Assess Accessibility Costs

Use Case:

Calculate the minimum cost to reach all urban zones from a central hub (e.g., city center) in the presence of varying costs or negative weights (e.g., subsidies)

5. Ford-Fulkerson Algorithm: Maximal Utility Allocation

Use Case:

Optimize the allocation of utilities (e.g., water, electricity) to urban areas by maximizing the flow through the network of supply points and urban zones.

Steps:

1. Represent the utility network as a directed graph where edges represent capacity.
2. Use BFS or DFS to find augmenting paths.
3. Increase the flow along augmenting paths until no more exist.
4. Output the maximum flow, ensuring optimal resource distribution.

Graph Representation:

- Nodes: Supply points and urban zones.
- Edges: Connections with capacities representing the maximum utility that can be supplied.

```
#include <iostream>
```

```
#include <vector>
```

```
#include <algorithm>
```

```
using namespace std;
```

```
struct Edge {    int u, v, weight;    bool
```

```
operator<(const Edge& other) const {
```

```
    return weight < other.weight;
```

```
    }
```

```
};
```

```
class Graph {
```

```
    int V;
```

```
    vector<Edge> edges;
```

```
public:
```

```
    Graph(int V) : V(V) {}
```

```
    void addEdge(int u, int v, int weight) {
```

```
        edges.push_back({u, v, weight});
```

```
    }
```

```

int find(vector<int>& parent, int i)
{
    if (parent[i] == i)
return i;

    return parent[i] = find(parent, parent[i]);
}

```

```

void unionSets(vector<int>& parent, vector<int>& rank, int x, int y) {
int rootX = find(parent, x);

```

```

    int rootY = find(parent, y);

    if (rootX != rootY) {
        if
(rank[rootX] < rank[rootY])
parent[rootX] = rootY;
        else if
(rank[rootX] > rank[rootY])
parent[rootY] = rootX;
        else
        {
parent[rootY] = rootX;
rank[rootX]++;
        }
    }
}

```

```

void kruskal() {
sort(edges.begin(), edges.end());
vector<int> parent(V);
vector<int> rank(V, 0);

    for (int i = 0; i < V; i++)
parent[i] = i;

```



```

        cout << "Minimum Spanning Tree (MST) for U lity
Networks:\n";
        for (const Edge& edge : edges) {
            int u =
find(parent, edge.u);
            int v = find(parent, edge.v);
            if (u !=
v) {
                cout << edge.u << " - " << edge.v << " (Cost: " << edge.weight << ")\n";
unionSets(parent, rank, u, v);
            }
        }
    }
};

int main() {
    Graph g(4);
    g.addEdge(0, 1, 5);
    g.addEdge(1, 2, 4);
    g.addEdge(2, 3, 6);
    g.addEdge(0, 3, 3);

    g.kruskal();
    return 0;
}

```

2. Union-Find for Land Ownership:

```
#include <iostream>
```

```
#include <vector>
```

```
using namespace std;
```

```

class UnionFind {
    vector<int> parent, rank;

public:
    UnionFind(int n) : parent(n), rank(n, 0)
    {
        for (int i = 0; i < n; i++)
            parent[i] = i;
    }

    int find(int x) {
        if (parent[x] != x)
            parent[x] = find(parent[x]);
        return parent[x];
    }

    void unionSets(int x, int y)
    {
        int rootX = find(x);
        int rootY = find(y);

        if (rootX != rootY) {
            if (rank[rootX] < rank[rootY])
                parent[rootX] = rootY;
            else if (rank[rootX] > rank[rootY])
                parent[rootY] = rootX;
            else {
                parent[rootY] = rootX;
                rank[rootX]++;
            }
        }
    }
}

```

```
};
```

```
int main() {
```

```
    UnionFind uf(5);
```

```
    uf.unionSets(0, 1);
```

```
    uf.unionSets(1, 2);
```

```
    cout << "Land ownership tracking:\n";    for (int i = 0; i < 5; i++) {        cout << "Owner of parcel " <<
i << ": " << uf.find(i) << endl;
    }
```

```
    return 0;
```

```
}
```

3. Prim's Algorithm

```
#include <iostream>
```

```
#include <vector>
```

```
#include <limits>
```

```
using namespace std;
```

```
int findMinKey(vector<int>& key, vector<bool>& mstSet, int V)
```

```
{    int min = INT_MAX, minIndex;
```

```
    for (int v = 0; v < V; v++) {
```

```
    if (!mstSet[v] && key[v] < min) {
```

```
    min = key[v];        minIndex = v;
```

```
    }
```

```
}
```

```
    return minIndex;
```

```
}
```

```

void printMST(vector<int>& parent, vector<vector<int>>& graph, int V) {
    cout << "Edge \tWeight\n";
    for (int i = 1; i < V; i++) {        cout << parent[i] << " - " << i <<
"\t" << graph[i][parent[i]] << endl;
    }
}

```

```

void primMST(vector<vector<int>>& graph, int V) {    vector<int> parent(V);
// Array to store constructed MST    vector<int> key(V, INT_MAX); // Key
values used to pick minimum weight edge    vector<bool> mstSet(V, false); // To
represent the set of ver ces included in MST

```

```

    key[0] = 0;    // Make the first vertex as root
parent[0] = -1; // First node is always the root of the MST

```

```

    for (int count = 0; count < V - 1; count++) {
int u = findMinKey(key, mstSet, V);

```

```

    mstSet[u] = true;

```

```

        for (int v = 0; v < V; v++) {            if (graph[u][v] &&
!mstSet[v] && graph[u][v] < key[v]) {
            parent[v] = u;
key[v] = graph[u][v];
        }
    }
}

```

```

    printMST(parent, graph, V);
}

```

```

int main() {
vector<vector<int>> graph = {
    {0, 2, 0, 6, 0},
    {2, 0, 3, 8, 5},
    {0, 3, 0, 0, 7},
    {6, 8, 0, 0, 9},
    {0, 5, 7, 9, 0}
};

    int V = graph.size();    cout << "Minimum Spanning Tree
using Prim's Algorithm:\n";    primMST(graph, V);

    return 0;
}

```

4.Bellman-Ford Algorithm

```

#include <iostream>
#include <vector>
#include <climits>

using namespace std;

void bellmanFord(int V, vector<vector<int>>& edges, int
src) {    vector<int> dist(V, INT_MAX);    dist[src] = 0;

    for (int i = 0; i < V - 1; i++) {        for (auto& edge :
edges) {            int u = edge[0], v = edge[1], weight =
edge[2];            if (dist[u] != INT_MAX && dist[u] +
weight < dist[v]) {                dist[v] = dist[u] + weight;
            }
        }
    }
}

```

```

    }

    cout << "Minimum costs from source:\n";
    for (int i = 0; i < V; i++) {
        cout << "To node " << i << ": " << dist[i] << endl;
    }
}

```

```

int main() {    int V = 4;
vector<vector<int>> edges = {
    {0, 1, 5},
    {1, 2, -2},
    {2, 3, 3},
    {0, 3, 10}
};

```

```

    bellmanFord(V, edges, 0);
return 0;
}

```

5.Ford-Fulkerson Algorithm

```

#include <iostream>
#include <vector>
#include <queue>
#include <limits>

```

```

using namespace std;

```

```

bool bfs(vector<vector<int>>& rGraph, int s, int t, vector<int>&
parent) {    int V = rGraph.size();    vector<bool> visited(V, false);
queue<int> q;

```

```

    q.push(s);
visited[s] = true;
parent[s] = -1;

    while (!q.empty()) {
int u = q.front();
    q.pop();

        for (int v = 0; v < V; v++) {
if (!visited[v] && rGraph[u][v] > 0) {
parent[v] = u;          visited[v] = true;
if (v == t) return true;
            q.push(v);
        }
    }
}

return false;
}

int fordFulkerson(vector<vector<int>>& graph, int s, int
t) {    int V = graph.size();    vector<vector<int>> rGraph
= graph; // Residual graph    vector<int> parent(V);    int
maxFlow = 0;

    while (bfs(rGraph, s, t, parent)) {
int pathFlow = INT_MAX;

        for (int v = t; v != s; v = parent[v]) {
int u = parent[v];          pathFlow =
min(pathFlow, rGraph[u][v]);

```

```

    }

    for (int v = t; v != s; v =
parent[v]) {
        int u = parent[v];
        rGraph[u][v] -= pathFlow;
        rGraph[v][u] += pathFlow;
    }

    maxFlow += pathFlow;
}

return maxFlow;
}

int main() {
vector<vector<int>>> graph = {
    {0, 10, 5, 15},
    {0, 0, 4, 0},
    {0, 0, 0, 10},
    {0, 0, 0, 0}
};

cout << "Maximum utility allocation: " << fordFulkerson(graph, 0, 3) << endl;
return 0;
}

```

Conclusion:

- Kruskal's Algorithm helps optimize utility networks, minimizing the cost of providing basic services like water, electricity, and internet.
- Union-Find Data Structure efficiently tracks land ownership and lease agreements, ensuring transparency and reducing disputes.
- Ford-Fulkerson: Maximizes utility allocation to urban zones efficiently.

- . Bellman-Ford: Computes accessibility costs accurately, even with varying or negative weights.
- prim's Algorithm is a powerful tool for building cost-effective and resilient infrastructure networks, ensuring minimal total cost while connecting all urban areas. It incrementally selects the least costly connections, making it ideal for designing transportation routes, pipelines, and utility systems. This approach supports sustainable urban planning, aligns with SDG Goal 11.3, and fosters inclusive development by enabling efficient and scalable infrastructure for cities like Vatsalya Nagar.

Business case: City Management

Sub-task: Smart City Services

SDG Goal: 11

Target: 11.2

Indicator: 11.2

Overview: Smart city services are integral to Vatsalya Nagar's vision of a technologically advanced and efficient urban ecosystem. By integrating IoT, data management, and digital infrastructure, the city aims to enhance governance, improve service delivery, and promote transparency. These services include smart infrastructure management, e-governance platforms, and real-time monitoring systems that optimize resource use. With an annual revenue potential of ₹1,200 crore, smart city services also foster innovation and create opportunities for businesses specializing in technology, making Vatsalya Nagar a model for modern urban living.

[illegible]

- o Use Case: Store and retrieve smart city data such as sensor readings, service logs, or citizen reports efficiently.
- o Steps:
 1. Insert each data point (e.g., timestamped sensor readings) into the BST.
 2. Search for specific data points or ranges of data quickly.
 3. Perform in-order traversal for sorted data analysis.
- o Implementation Scenario: Enables fast data access for real-time decision-making and analytics.

3. Bubble Sort: Low-Priority Task Management

- o Use Case: Handle low-priority tasks such as ranking citizen feedback forms or sorting non-critical notifications.
- o Steps:
 1. Iterate through the list of tasks, comparing adjacent items.
 2. Swap items if they are out of order.
 3. Repeat until the entire list is sorted.
- o Implementation Scenario: Simple sorting for tasks that do not require advanced algorithms or real-time performance.

4. Dijkstra's Algorithm: Optimize Public Transport Routes

- o Use Case: Find the shortest and safest routes for public transport, ensuring reduced travel time and better connectivity across the city.
- o Steps:
 1. Initialize distances from the source to all nodes as infinity, except the source itself.
 2. Use a priority queue to explore the nearest node and update distances for its neighbors.
 3. Repeat until all nodes are visited.
 4. Output the shortest paths.

5. A* Search Algorithm*: Optimize Emergency Services

Use Case: Efficiently navigate emergency vehicles through the city by considering real-time traffic and heuristic distances.

1. Maintain open and closed sets for nodes.
2. Use a heuristic (e.g., straight-line distance) to estimate cost to the target.
3. Prioritize nodes in the open set by total estimated cost.
4. Expand nodes and update costs iteratively.

Code Implementation:

Code Implementation:

1. Heap Sort for Task Prioritization:

```
import heapq

def prioritize_tasks(tasks):
    # Convert task list to a max-heap by negating priorities
    max_heap = [(-priority, task) for task, priority in tasks]
    heapq.heapify(max_heap)

    print("Task Priority Order:")
    while max_heap:
        priority, task = heapq.heappop(max_heap)
        print(f"Task: {task}, Priority: {-priority}")

# Example tasks with priorities
tasks = [("Garbage Collection", 3), ("Energy Distribution",
5), ("Emergency Response", 10)]
prioritize_tasks(tasks)
```

2. Binary Search Tree for Data Management:

```
class Node:
```

```
    def __init__(self, key):
```

```
        self.key = key
```

```
self.left = None
```

```
self.right = None
```

```
class BST:    def
```

```
    __init__(self):
```

```
self.root = None
```

```
def insert(self, key):
```

```
if not self.root:
```

```
self.root =
```

```

Node(key)

else:
    self._insert(self.root, key)

    def _insert(self, node, key):
    if key < node.key:
        if
    node.le is None:
    node.le = Node(key)
        else:
    self._insert(node.le , key)
    else:
        if node.right is None:
    node.right = Node(key)
        else:
            self._insert(node.right, key)

    def in_order_traversal(self, node):
        if node:
            self.in_order_traversal(node.le )
    print(node.key,          end="          ")
    self.in_order_traversal(node.right)

# Example usage
bst = BST() data_points = [15, 10, 20,
8, 12, 16, 25] for data in data_points:
    bst.insert(data)

print("In-Order Traversal (Sorted Data):")
bst.in_order_traversal(bst.root)
06330
066

```

3. Bubble-sort

Algorithm def

```
bubble_sort(arr):    n =
len(arr)    for i in
range(n - 1):
    # Last i elements are already sorted, no need to compare them
    for j in range(n - i - 1):
if arr[j] > arr[j + 1]:
    # Swap if the elements are in the wrong order
    arr[j], arr[j + 1] = arr[j + 1], arr[j]    print(f"Itera on
{i + 1}: {arr}") # Debugging to show progress

# Example usage: Sorting ci zen feedback scores
feedback_scores = [8, 5, 7, 2, 9, 3] print("Original
Feedback Scores:", feedback_scores)

bubble_sort(feedback_scores)

print("Sorted Feedback Scores:", feedback_scores)
```

4. Dijkstra's Algorithm

```
import heapq

def dijkstra(graph, start):    n = len(graph)
distances = [float('inf')] * n    distances[start] =
0    priority_queue = [(0, start)] # (distance,
node)    while priority_queue:
        current_distance, current_node = heapq.heappop(priority_queue)

        if current_distance > distances[current_node]:
            continue
```

```

        for neighbor, weight in graph[current_node]:
            distance = current_distance + weight            if distance
< distances[neighbor]:            distances[neighbor] =
distance            heapq.heappush(priority_queue, (distance,
neighbor))

```

```

return distances

```

```

# Example: Graph as an adjacency list

```

```

# Nodes represent city areas; edges represent public transport routes with weights as travel mes

```

```

graph = [
    [(1, 5), (2, 10)], # Node 0
    [(0, 5), (2, 3)], # Node 1
    [(0, 10), (1, 3)] # Node 2
]

```

```

start_node = 0 print("Shortest distances from node 0:",
dijkstra(graph, start_node)

```

```

5. A Search Algorithm from heapq

```

```

import heapq

```

```

def a_star(graph, start, goal, h):    open_set = []
    heapq.heappush(open_set, (0, start))    g_scores =
{node: float('inf') for node in graph}
    g_scores[start] = 0    came_from = {}

```

```

while open_set:

```

```

    current = heapq.heappop(open_set)[1]

```

```

    if current == goal:

```

```

        path = []
        while current
in came_from:
        path.append(current)
        current
= came_from[current]
        return
path[::-1]

        for neighbor, weight in graph[current]:
            tenta ve_g_score = g_scores[current] + weight
        if tenta ve_g_score < g_scores[neighbor]:
            came_from[neighbor] = current
            g_scores[neighbor] = tenta ve_g_score
            f_score = tenta ve_g_score + h[neighbor]
            heappush(open_set, (f_score, neighbor))

        return []

# Example graph and heuristic graph
= {
    0: [(1, 2), (2, 4)],
    1: [(0, 2), (3, 7)],
    2: [(0, 4), (3, 1)],
    3: []
}
heuristic = {0: 5, 1: 3, 2: 2, 3: 0}
start_node = 0 goal_node
= 3

print("Optimal path:", a_star(graph, start_node, goal_node, heuristic))

```

Conclusion:

- Heap Sort ensures smart city services are prioritized based on urgency and resource availability, improving operational efficiency.

- Binary Search Tree (BST) provides fast and organized storage for smart city data, supporting real-time analytics and decision-making.

Bubble Sort is intuitive and easy to implement. It works well for small datasets or cases where simplicity is preferred over performance. However, for larger datasets, more efficient algorithms like Quick Sort or Merge Sort are recommended.

Dijkstra's Algorithm:

This algorithm is highly effective for finding the shortest path in weighted graphs, making it ideal for optimizing public transportation routes and traffic management in smart cities. Its guaranteed accuracy ensures reliable pathfinding in various applications, including emergency response and infrastructure planning.

A Search Algorithm

A* combines the benefits of Dijkstra's algorithm and heuristic techniques, making it faster and more efficient for goal-oriented pathfinding. It is particularly suitable for dynamic smart city environments, where real-time navigation and adaptive planning are required, such as autonomous vehicles and delivery systems.