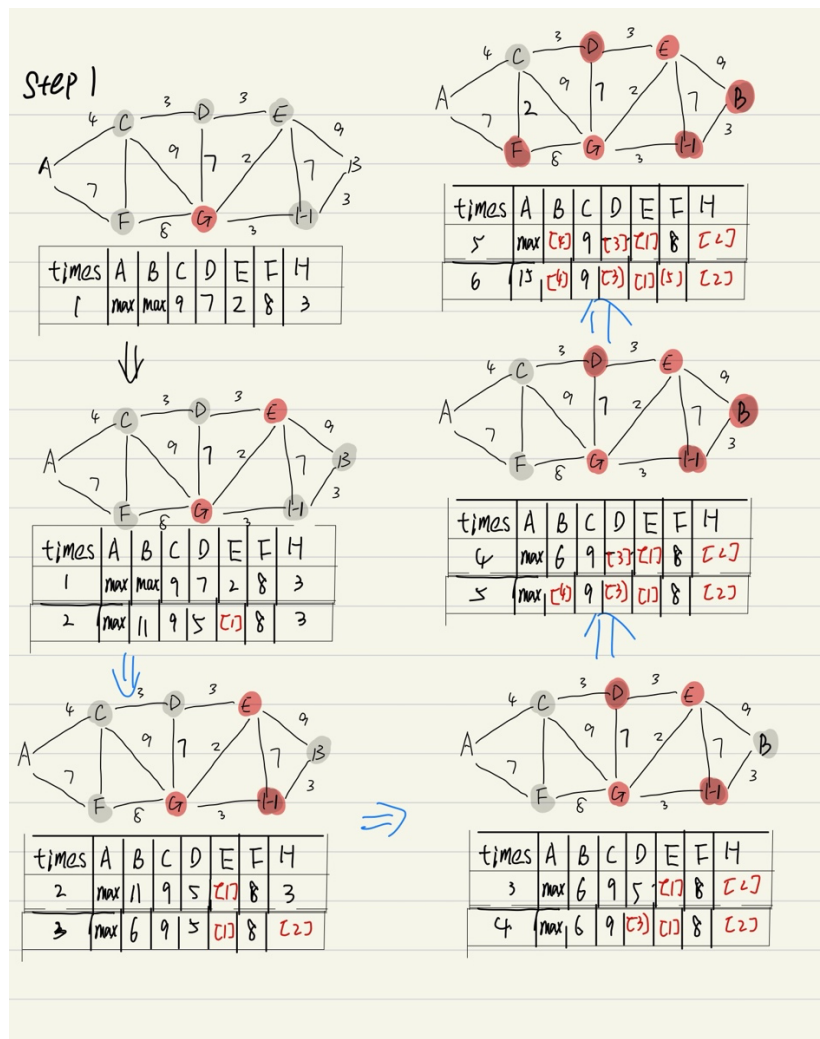
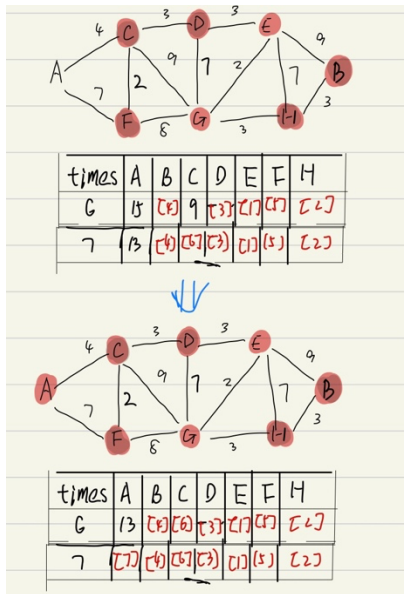


Problem 1.

1. Dijkstra's Algorithm can solve this problem. This algorithm could work on both directed and undirected graphs. By using greedy way find the shortest paths in the graph

Demonstrate:





Resulting Routes:

Destination	Distance	Path
G -> A:	12	G->E->D->C->A
G -> B	6	G->H->B
G -> C	8	G->E->D->C
G -> D	5	G->E->D
G -> E	2	G -> E
G -> F	8	G -> F
G -> H	3	G -> H

2. Find one “optimal” location:

For this question I will use Dijkstra’s Algorithm. Firstly, I need to find distance than each point as a starting point to the farthest point. Secondly, use this starting point as a key, place the longest distance as a value in a tuple, and then add the tuple to the list. Until joining so the

point. Finally, the tuples are sorted by value, returning the minimum key, which is the optimal point.

Algorithmic implementation:

```

69 MAX= float('inf')
70 matrix = [[MAX, MAX, 4, MAX, MAX, 7, MAX, MAX],
71           [MAX, MAX, MAX, MAX, 9, MAX, MAX, 3],
72           [4, MAX, MAX, 3, MAX, 2, 9, MAX],
73           [MAX, MAX, 3, MAX, 3, MAX, 7, MAX],
74           [MAX, 9, MAX, 3, MAX, MAX, 2, 7],
75           [7, MAX, 2, MAX, MAX, MAX, 8, MAX],
76           [MAX, MAX, 9, 7, 2, 8, MAX, 3],
77           [MAX, 3, MAX, MAX, 7, MAX, 3, MAX]]
78
79 def dijkstra(matrix, start_node):
80     matrix_length = len(matrix)
81     used_node = [False] * matrix_length
82     distance = [MAX] * matrix_length
83     distance[start_node] = 0
84     while used_node.count(False):
85         min_value = float('inf')
86         min_value_index = 999
87         for index in range(matrix_length):
88             if not used_node[index] and distance[index] < min_value:
89                 min_value = distance[index]
90                 min_value_index = index
91
92         used_node[min_value_index] = True
93         for index in range(matrix_length):
94             distance[index] = min(distance[index], distance[min_value_index] + matrix[min_value_index][index])
95
96     return distance
97
98 list1 = []
99 for i in range(8):
100     result = dijkstra(matrix,i)
101     result.sort()
102     list1.append((chr(65+i), result[7]))
103
104 list1.sort(key = lambda x : x[1])
105 print(list1)
106 print(list1[0])

```

The Dijkstra's Running time is $O(E \log V)$

If the n is the number of vertices, the time complexity is $n * O(E \log V)$

3. According to the output of the code, the optimal town is E

```

[('E', 10), ('D', 11), ('G', 12), ('C', 14), ('F', 14), ('H', 15), ('A', 18), ('B', 18)]
('E', 10)

```

4. For this problem, I still use Dijkstra's Algorithm. First, we need to find all the cases where any two-point combination is. Second, in the current combination, the Dijkstra's algorithm is used to get their distance to each point. Then compare the distance between the two points

to the same point and take out a smaller distance to join the list. Third, take the maximum value from the list as the distance from this combination to the furthest. Finally, compare the distance from all combinations to the farthest point. The smallest combination is the optimal solution.

Algorithmic implementation:

```
1 MAX= float('inf')
2 matrix = [[MAX, MAX, 4, MAX, MAX, 7, MAX, MAX],
3           [MAX, MAX, MAX, MAX, 9, MAX, MAX, 3],
4           [4, MAX, MAX, 3, MAX, 2, 9, MAX],
5           [MAX, MAX, 3, MAX, 3, MAX, 7, MAX],
6           [MAX, 9, MAX, 3, MAX, MAX, 2, 7],
7           [7, MAX, 2, MAX, MAX, MAX, 8, MAX],
8           [MAX, MAX, 9, 7, 2, 8, MAX, 3],
9           [MAX, 3, MAX, MAX, 7, MAX, 3, MAX]]
10
11 def dijkstra(matrix, start_node):
12     matrix_length = len(matrix)
13     used_node = [False] * matrix_length
14     distance = [MAX] * matrix_length
15     distance[start_node] = 0
16     while used_node.count(False):
17         min_value = float('inf')
18         min_value_index = 999
19         for index in range(matrix_length):
20             if not used_node[index] and distance[index] < min_value:
21                 min_value = distance[index]
22                 min_value_index = index
23
24         used_node[min_value_index] = True
25         for index in range(matrix_length):
26             distance[index] = min(distance[index], distance[min_value_index] + matrix[min_value_index][index])
27
28     return distance
```

```
30 def comparelist(tuple1, tuple2):
31     resultlist = []
32     for i in range(len(tuple1)):
33         for j in range(len(tuple2)):
34             if (tuple1[i][0] == tuple2[j][0]):
35                 if (tuple1[i][1] < tuple2[j][1]):
36                     resultlist.append(tuple1[i])
37                 else:
38                     resultlist.append(tuple2[j])
39     return resultlist
40
41 def creatlist(n):
42     list1 = []
43     for i in range(n):
44         list1.append(i)
45     return list1
```

```

47 def findpoints(matrix):
48     alldis_list = []
49     x,y = 0, 0
50     for i in range(len(matrix)-1):
51         y = x+1
52         for _ in range(len(matrix)-1-i):
53             val_listx = creatlist(len(matrix))
54             del val_listx[y]
55             listx = dijkstra(matrix, x)
56             del listx[y]
57             tuplex = list(zip(val_listx,listx))
58
59
60             val_listy = creatlist(len(matrix))
61             del val_listy[x]
62             listy = dijkstra(matrix, y)
63             del listy[x]
64             tupley = list(zip(val_listy,listy))
65
66
67             finallist = comparelist(tuplex, tupley)
68             finallist.sort(key = lambda x : x[1])
69             maxd = finallist[5][1]
70             alldis_list.append((chr(x+65), chr(y+65), maxd))
71             y += 1
72         x += 1
73     alldis_list.sort(key= lambda x : x[2])
74     return alldis_list
75
76 print(findpoints(matrix))
77

```

Theoretical running time: $O(V^3)$

5.

```

[('C', 'H', 5), ('A', 'G', 6), ('B', 'C', 6), ('C', 'G', 6), ('F', 'G', 6), ('F', 'H', 6), ('A', 'H', 7), ('B', 'D', 7), ('D', 'G', 7), ('D', 'H', 7), ('A', 'B', 8), ('A', 'E', 8), ('B', 'F', 8), ('C', 'E', 8), ('D', 'E', 8), ('E', 'F', 8), ('B', 'E', 10), ('E', 'G', 10), ('E', 'H', 10), ('A', 'D', 11), ('C', 'D', 11), ('D', 'F', 11), ('B', 'G', 12), ('G', 'H', 12), ('A', 'C', 14), ('A', 'F', 14), ('C', 'F', 14), ('B', 'H', 15)]

```

In the output of the code, we can see the “C” and “H” are the optimal towns to place the two distribution centres.

Problem 2.

1. Verbal description:

I used Kruskal Algorithm to solve this MST problem. This algorithm is seeming like greedy algorithms, which find the optimum in the hopes of finding a global optimum. In the Kruskal algorithm, 1. need sort each edge according to the distance. 2. Take the minimum distance and add it to the spanning tree. If adding the edge created a cycle, then reject this edge. 3. Keep adding edges until all the points are connected. In the main function, we can just need add the distance of edge which in the spanning tree together.

2. Pseudocode:

```

1  def findParent(parent, n):
2      if parent[n] == n:
3          return n
4      return findParent(parent, parent[n])
5
6  def union(parent, x, y):
7      parent[findParent(parent,x)] = findParent(parent,y)
8
9  def kruskal_algo(g, n):
10     g <- sort according x[2]
11     for node in range(n):
12         parent[] <- node
13     for i in range(len(g)):
14         u, v, w = g[i]
15         x = findParent(parent, u)
16         y = findParent(parent, v)
17         if (x != y):
18             result += w
19             edge += 1
20             union(parent, x, y)
21     if (edge == n-1):
22         return result
23
24  def handlefile(filename):
25     f = open(filename)
26     lines <= f.readlines()
27     test_cases <-lines[0]
28     for i in range(test_cases):
29         n_vertex <- lines[lines_idx]
30         for _ in range(n_vertex):
31             point[] <- lines[lines_idx]
32             all_points.append(point)
33             lines_idx += 1
34         g = creat_graph(all_points)
35         print("Test case {0}: MST weight {1}".format(i+1, kruskal_algo(g, n_vertex)))

```

3. Theoretical running time:

$O(E \log V)$

1. For loop each vertex, put them into a list: $O(V)$
2. Sort the edge according to the weight: $O(E \log E)$
3. $O(E)$ calls to Findparent() and union(). Each take $O(\log V)$
4. Because $|E| < V*V \Rightarrow \log E = 2\log V$

The total running time we can write $O(E \log V)$