

Analysis of Algorithms
CSCE – 629 (Spring 2019)
Course Project Report

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In this course project, I implemented a network routing protocol using the data structures and algorithms we have studied in class. This provided me with an opportunity to translate my theoretical understanding into a real-world practical computer program. Translating algorithmic ideas at a “higher level” of abstraction into real implementations in a programming language is not always trivial. The implementations forced me to work on more details of the algorithms, which indeed lead to much better understanding.

1. Graph Generation:

- We had to implement two graphs with 5000 vertices each– Sparse Graph with average Vertex degree of about 6 and a Dense Graph with an average vertex degree of 1000 (20%).
- I have created two classes, Node class and Edge Class. Node class has attribute as ‘node name’ and ‘adjacent list of edges’. Edge class has attribute as ‘source node’, ‘target node’ and ‘edge weight’. Edge weight is randomly generated between 1 – 9999.
- The Connection is done randomly. To implement this, I have randomly generated a number between 1 and 4999 for every possible neighbor vertex of a source vertex and if it less than 6 (for sparse graph) or 1000 (for dense graph), I have added an edge between the two vertices.
- To make sure that the generated graph is correct and follows the desired properties. I have defined a method to verify the graph. Another method outputs the average number of connections per node.

```
class Node:
    def __init__(self, name):
        self.name = name
        self.adjacent_list_of_edges = []
```

```
class Edge:
    def __init__(self, source, target, weight):
        self.source = source
        self.target = target
        self.weight = weight
```

```

import random
def generate_graph(num_nodes, probability):
    graph = []
    n_edges = 0
    for i in range(num_nodes):
        graph.append(Node(str(i)))

    #making sure that graph is connected by making a cycle
    for i in range(num_nodes-1):
        initial_node = graph[i]
        target_node = graph[i+1]
        edge_weight = random.randint(1,10000)
        initial_node.adjacent_list_of_edges.append(Edge(initial_node,target_node,
edge_weight))
        target_node.adjacent_list_of_edges.append(Edge(target_node, initial_node,
edge_weight))
        n_edges+=1
    edge_weight = random.randint(1,10000)
    graph[num_nodes-1].adjacent_list_of_edges.append(Edge(graph[num_nodes-1], graph[0],
edge_weight))
    graph[0].adjacent_list_of_edges.append(Edge(graph[0], graph[num_nodes-1], edge_weight))
    n_edges+=1

    #randomly adding other nodes in the graph
    for i in range(num_nodes):
        initial_node = graph[i]
        for j in range(i+1,num_nodes):
            if(random.randint(0,num_nodes) < (probability-1/num_nodes) * num_nodes):
                target_node = graph[j]
                edge_weight = random.randint(1,10000)
                initial_node.adjacent_list_of_edges.append(Edge(initial_node, target_node,
edge_weight))
                target_node.adjacent_list_of_edges.append(Edge(target_node, initial_node,
edge_weight))
                n_edges+=1

    return graph, n_edges

```

Average number of edges per node for dense graph is
1000.5832

Average number of edges per node for sparse graph is
5.9992

2. Heap Structure:

I have implemented a heap Class following the instructions provided for heap. My heap is implemented as follows:

- The vertices of a graph are named by integers 0, 1, . . . , 4999;
- The heap is given by an array H[5000], where each element H[i] gives the name of a vertex in the graph. It is named as 'heap'.
- The vertex "values" are given in another array D[5000]. Thus, to find the value of a vertex H[i] in the heap, we can use D[H[i]].
- In addition to above arrays, I have also kept an array node_to_heap_index which stores the index of a node in the heap.

```
import math
import sys
class heap:
    def __init__(self, max_size):
        self.max_size = max_size
        self.heap = [-1] * self.max_size
        self.D = [-1]* self.max_size
        self.node_to_heap_index = [-1] * self.max_size
        self.size = 0

    def Parent(self, i):
        if i%2 !=0:
            return math.floor((i-1)/2)
        else :
            return math.floor((i-2)/2)

    def Left_Child(self, i):
        return (2 * i + 1) if (2* i + 1)< self.size else -1

    def Right_Child(self, i):
        return (2 * i + 2) if (2* i + 2)< self.size else -1

    def Swap(self,i,k):

        self.heap[k],self.heap[i] = self.heap[i], self.heap[k]
        self.node_to_heap_index[self.heap[k]], self.node_to_heap_index[self.heap[i]] =
self.node_to_heap_index[self.heap[i]], self.node_to_heap_index[self.heap[k]]
```

```

def Max(self):
    return self.heap[0]

def Fix_Heap(self,i):
    L = self.Left_Child(i)
    R = self.Right_Child(i)
    maximum = i
    if L != -1:
        if L<= self.size-1 and self.D[self.heap[i]]< self.D[self.heap[L]]:
            maximum = L
        else:
            maximum = i
    if R!= -1:
        if R<= self.size-1 and self.D[self.heap[maximum]]<
self.D[self.heap[R]]:
            maximum = R

    if maximum != i:
        self.Swap(i,maximum)
        self.Fix_Heap(maximum)

def Extract_Max(self):
    if self.size < 1:
        print('error')

    max_element = self.heap[0], self.D[self.heap[0]]
    self.Swap(0, self.size-1)
    self.size= self.size - 1
    self.Fix_Heap(0)
    return max_element

def Insert(self, a, BW):
    if self.size>= self.max_size:
        print('error is here')
    self.heap[self.size] = int(a.name)
    self.D[int(a.name)] = BW
    i = self.size
    self.size = self.size+1

    self.node_to_heap_index[int(a.name)] = i
    while(i>0 and (self.D[self.heap[self.Parent(i)]] < self.D[self.heap[i]])):
        self.Swap(i,self.Parent(i))
        i = self.Parent(i)

```

```

def reset_node(self, node_number, value):
    index = self.node_to_heap_index[node_number]
    if(index == -1):
        raise IndexError('Node not in Heap')
    self.D[self.heap[index]] = value
    parent = self.Parent(index)
    while(self.D[self.heap[parent]] < self.D[self.heap[index]]):
        self.Swap(index, parent)
        index = parent
        if(parent == 0):
            break
        parent = parent = self.Parent(index)

    self.Fix_Heap(index)

def Delete(self, i):
    if i > self.size:
        print('error in there')
    self.D[self.heap[i]] = sys.maxsize
    while(i > 0 and (self.D[self.heap[self.Parent(i)]] < self.D[self.heap[i]])):
        self.Swap(i, self.Parent(i))
        i = self.Parent(i)

    self.heap[0] = self.heap[self.size-1]
    self.node_to_heap_index[self.heap[0]] = 0
    self.heap[self.size-1] = -1
    self.size = self.size - 1

    self.Fix_Heap(0)

def print_heap(self, node = 0, tab = ''):
    if(node == -1):
        return
    print(tab, str(self.D[self.heap[node]]))
    self.print_heap(self.Left_Child(node), tab + '\t')
    self.print_heap(self.Right_Child(node), tab + '\t')

```

3. Routing Algorithms:

General Algorithms:

1) Max Bandwidth path using Dijkstra's algorithm:

```
a) for each vertex v = 1 to n do
    status[v] = unseen
b) status[s] =intree
d[s] = 0
dad[s] = -1
c) for each edge[s,v] do
    status[v] = fringe
    dad[v] = s
    d[v] = wt(s,v)
d) while there are fringes do
    let v be the fringe with max(d[v])
    // For array it goes through all the d[v]
    //To find out the max for heap it does extract_max()
    status[v] =intree
    for each edge[v,t] do
        if status[t] == unseen
            then status[t] = fringe
                dad[t] = v
                d[t] = min(d[v], wt(v,t))
        else if status[t] == fringe and d[t] < min(d[v], wt(v,t))
            then d[t] = min(d[v], wt(v,t))
                dad[t] = v
```

2) Max Bandwidth path using Kruskal's Algorithm:

```
a) Sort all edges in decreasing order
b) T =  $\phi$ 
c) for each edge ei = [vi, wi] do
    r1 = find(vi)
    r2 = find(wi)
    if r1 != r2
        then T = T + ei
        Union(r1, r2)
d) return T
```

3) Union-Find

```
Union(s1, s2):
if rank[s1] > rank[s2]
    then dad[s2] = s1
```

```

        rank[s1] += rank[s2]
    else
        then dad[s1] = s2
        rank[s2] += rank[s1]

```

```

Find(i):
a) s = queue()
b) while (i != dad[i]) do
    s.push(i)
    i = dad[i]
c) while (s not empty) do
    dad[s.pop()] = i
d) return i

```

Time Complexity:

A. MaxBandwidthPath using Dijkstra's Algorithm(array implementation) : $O(n^2+m)$

The inner loop to find max runs n times and the outer loop can also run maximum of $(n-1)$ times. Hence, $O(n^2)$. The inner for loop runs for total of m times in whole program. Hence, $O(m)$. Thus overall time complexity is $O(n^2+m)$

B. MaxBandwidthPath using Dijkstra's Algorithm (heap implementation) : $O(m\log n + n\log n)$

The inner for loop in whole program runs for total number of edges, m , and the insert in the heap takes $\log n$ times. Hence, $O(m\log n)$. The outer loop can run maximum of $(n-1)$ times and the delete in heap takes $O(\log n)$. Hence $O(n\log n)$. Thus overall time complexity comes to $O(n\log n + m\log n)$

C. MaxBandwidthPath using Kruskal Algorithm : $O(m\log^*n + m\log m)$

Sorting m edges takes time $O(m\log m)$ and m find operations with path compression in the for loop takes $O(m\log^*n)$ time. Hence, $O(m\log m + m\log^*n)$

4.Implementation:

1. Dijkstra's without heap:

- This follows the normal Dijkstra's implementation where the largest fringe is chosen as the largest fringe element in the array.
- This requires traversing through entire array each time.
- Function `max_BW_fringe` calculates the largest fringe and return its index.
- 'BW', 'status' and 'dad' array stores the Bandwidth, status (0 : Unseen, 1: fringe, 2: intree) and Dad of a particular vertex indexed using its number.

```
def apply_Without_Heap(self):

    fringes_count = 0
    BW = [-1] * len(self.G)
    status = [-1] * len(self.G)
    dad = [-1] * len(self.G)

    for node in self.G:
        status[int(node.name)] = 0 # 0 --> Unseen

    status[int(self.S.name)] = 2 # 2 --> Intree
    BW[int(self.S.name)] = float('inf')

    for edge in self.G[int(self.S.name)].adjacent_list_of_edges:
        BW[int(edge.target.name)] = int(edge.weight)
        status[int(edge.target.name)] = 1 # 1 --> Fringe
        dad[int(edge.target.name)] = int(self.S.name)
        fringes_count += 1

    maximum_bw = sys.maxsize
    while(fringes_count > 0):
        current_max = self.max_BW_fringe(self.G, status, BW, len(self.G))
        status[current_max] = 2
        if(current_max == int(self.T.name)):
            maximum_bw = BW[current_max]
            return maximum_bw, dad
```

```

fringes_count -= 1
for edge in self.G[current_max].adjacent_list_of_edges:
    w = int(edge.target.name)
    if(status[w] == 0):
        status[w] = 1
        fringes_count += 1
        dad[w] = current_max
        BW[w] = min(BW[current_max], int(edge.weight))
    elif(status[w] == 1 and BW[w] < min(BW[current_max], int(edge.weight))):
        dad[w] = current_max
        BW[w] = min(BW[current_max], int(edge.weight))

```

```

def max_BW_fringe(self, graph, status_array, bandwidth_array, N):
    index = -1
    max_fringe_bw = -sys.maxsize - 1
    for i in range(0, N):
        if((status_array[i] == 1) and (bandwidth_array[i] >= max_fringe_bw)):
            index = i
            max_fringe_bw = bandwidth_array[i]
    return index

```

2. Dijkstra's with heap:

- This uses the heap class defined in step2.
- The largest fringe is calculated through heap operation – extract_max.
- Whenever a new node is identified as fringe, it is inserted in the heap.
- Whenever a new BW value for a fringe is generated, older fringe is deleted, and new fringe is inserted in heap using the reset_node command.

```

def apply_With_Heap(self):
    BW = [-1] * len(self.G)
    status = [-1] * len(self.G)
    dad = [-1] * len(self.G)
    self.max_heap = heap(len(self.G))
    for node in self.G:
        status[int(node.name)] = 0 # 0 --> Unseen

```

```

status[int(self.S.name)] = 2 # 2 --> Intree
BW[int(self.S.name)] = float('inf')

for edge in self.G[int(self.S.name)].adjacent_list_of_edges:
    status[int(edge.target.name)] = 1 # 1 --> Fringe
    BW[int(edge.target.name)] = int(edge.weight)
    self.max_heap.Insert(edge.target, BW[int(edge.target.name)])
    dad[int(edge.target.name)] = int(self.S.name)

while (self.max_heap.size != 0):
    max_element = self.max_heap.Extract_Max()
    current_max = max_element[0]
    status[current_max] = 2
    if current_max == int(self.T.name):
        return dad, BW[current_max]
    for edge in self.G[current_max].adjacent_list_of_edges:
        w = int(edge.target.name)
        if status[w] == 0:
            status[w] = 1
            BW[w] = min(BW[current_max], int(edge.weight))

            self.max_heap.Insert(edge.target, BW[w])

            dad[w] = current_max

        elif (status[w] == 1 and BW[w] < min(BW[current_max], int(edge.weight)) ):
            dad[w] = current_max
            BW[w] = min(BW[current_max], int(edge.weight))
            self.max_heap.reset_node(w, BW[w])

```

3. Kruskal's using Heap Sort:

- To implement heap sort for getting the edge with highest weight each time. I have implemented another heap specific to edges. It contains 3 arrays: 'S', 'T' and 'D'. S array contains the source of edge, T array contains the target of edge and D contains the value of edge weight. It implements similar procedure like the Dijkstra's heap to implement heap sort.

- This new heap gives out an iterable number of extract_max operations (using Yield Operation). Thus, allowing us to receive all edges in non-increasing order.
- After that we do number of union-find operations following the below mentioned implementations.

```

from queue import Queue
import sys
class Kruskal:
    def __init__(self, Graph, Source, target, n_edges):
        self.G = Graph
        self.S = Source
        self.T = target
        self.num_of_edges = n_edges

    def sort_and_iterate_edges(self):
        edge_heap = EdgeHeap(self.num_of_edges)
        for i in range(0, len(self.G)):
            for edge in self.G[i].adjacent_list_of_edges:
                if(int(edge.source.name) < int(edge.target.name)):
                    edge_heap.Insert(int(edge.source.name), int(edge.target.name),
int(edge.weight))
        for _ in range(0, self.num_of_edges):
            yield edge_heap.extract_max()

    def get_maximum_spanning_tree(self):
        N= len(self.G)
        maximum_spanning_tree = []
        for i in range(0, N):
            maximum_spanning_tree.append(Node(i))
        dad_array = [-1] * N
        rank_array = [0] * N
        for edge_source, edge_target, edge_weight in self.sort_and_iterate_edges():
            r1 = self.find(edge_source, dad_array)
            r2 = self.find(edge_target, dad_array)
            if(r1 != r2):
                maximum_spanning_tree[edge_source].adjacent_list_of_edges.append(Edge(maximum_spanning_tree[edge_target], maximum_spanning_tree[edge_source], int(edge_weight)))
                maximum_spanning_tree[edge_target].adjacent_list_of_edges.append(Edge(maximum_spanning_tree[edge_source], maximum_spanning_tree[edge_target], int(edge_weight)))

```

```

        self.union(r1, r2, rank_array, dad_array)
    return maximum_spanning_tree

def union(self, rank1, rank2, rank_array, dad_array):
    if(rank_array[rank1] > rank_array[rank2]):
        dad_array[rank2] = rank1
    elif (rank_array[rank1] < rank_array[rank2]):
        dad_array[rank1] = rank2
    else:
        dad_array[rank1] = rank2
        rank_array[rank2] += 1

def find(self, v, dad_array):
    w = v
    q = Queue()
    while(dad_array[w] != -1):
        q.put(w)
        w = dad_array[w]
    while not q.empty():
        dad_array[q.get()] = w

    return w

def get(self,maximum_spanning_tree, i, j):
    for edge in maximum_spanning_tree[i].adjacent_list_of_edges:
        if(int(edge.target.name) == j):
            return int(edge.weight)
    return -1

def apply_dfs(self, graph, node_number, color_array, path_array, target):
    if (node_number == target):
        return True
    found = False
    color_array[node_number] = 2
    for edge in graph[node_number].adjacent_list_of_edges:
        if(color_array[int(edge.target.name)] == 1):
            path_array[int(edge.target.name)] = int(edge.source.name)
            found = self.apply_dfs(graph, int(edge.target.name), color_array, path_array,
target)

            if found:
                break
    color_array[node_number] = 3
    return found

def get_maximum_bandwidth(self, maximum_spanning_tree, source, target, N):

```

```

color_array = [1] * N # 1 -->White
path_array = [-1] * N

self.apply_dfs(maximum_spanning_tree, source, color_array, path_array, target)

path = str(target)
k = target
maximum_bandwidth = sys.maxsize
while(k != source):
    path = str(path_array[k]) + "->" + path
    maximum_bandwidth = min(maximum_bandwidth, self.get(maximum_spanning_tree, k,
path_array[k]))
    k = path_array[k]
return maximum_bandwidth, path

def apply_Kruskal(self):

    maximum_spanning_tree = self.get_maximum_spanning_tree()
    max_BW, path = self.get_maximum_bandwidth(maximum_spanning_tree, self.S, self.T,
len(self.G))
    return max_BW, path

```

5. Testing:

- For testing, we must generate 5 different graphs (both sparse and dense). For each graph we must randomly generate 5 different Source Target Pair.
- We then apply all three algorithms on each S-T pair.
- We calculate the time taken to generate the path by each algorithm.
- We then evaluate and compare the results.
- Following testing is done using the under mentioned code.

```

import time
def apply_different_algorithms(graph, n_times):
    for _ in range(n_times):
        source, target = get_random_ST_pair(graph[0])
        MWP = Dijkstra(graph[0], source, target)
        MBW = Kruskal(graph[0], int(source.name), int(target.name), graph[1])
        start_with_heap = time.time()
        value = MWP.apply_With_Heap()

```

```

end_with_heap = time.time()

start_without_heap = time.time()
value2 = MWP.apply_Without_Heap()
end_without_heap = time.time()

start_kruskal = time.time()
value3 = MBW.apply_Kruskal()
end_kruskal = time.time()

path = str(target.name)
k = int(target.name)
while(k != int(source.name)):
    path = str(value[0][k]) + "->" + path
    k = value[0][k]

print(path)
print('max BW with heap=', value[1])
print('time taken with heap=', (end_with_heap-start_with_heap))

without_heap_path = str(target.name)
k1 = int(target.name)
while(k1 != int(source.name)):
    without_heap_path = str(value2[1][k1]) + "->" + without_heap_path
    k1 = value2[1][k1]

print(without_heap_path)
print('Max BW without heap =', value2[0])
print('time taken without heap=', (end_without_heap-start_without_heap))

print(value3[1])
print('max BW path using Kruskal=', value3[0])
print('time taken with kruskal=', (end_kruskal-start_kruskal))

```

```

for i in range(5):
    sparse_graph = generate_graph(5000, float(6/5000))
    print('applying on sparse graph----- /n ---
-----')
    apply_different_algorithms(sparse_graph,5)
    dense_graph = generate_graph(5000, 0.20)
    print('applying on dense graph----- /n ---
-----')
    apply_different_algorithms(dense_graph,5)

```

6.Results:

The following example depicts the procedure using one example. The rest of the results will be depicted using the table provided.

Source = 406; Target = 4593 (Sparse Graph 1)

Path = 406->1277->3171->1888->75->74->3082->3083->124->4725->1556->1358->3617->4304->1875->1876->1350->3001->3000->4196->4197->2225->1364->455->2457->777->778->779->1522->1521->2560->2548->2549->2550->656->1748->1373->4410->4409->4408->4407->1433->4939->4940->4941->2435->1289->1290->3946->3947->3401->3402->1363->4028->1960->1961->3197->337->1513->161->2792->4419->3431->3432->953->4311->4593

Max BW without heap = 6832

time taken without heap= 1.4002 secs

Path = 406->1277->3171->1888->75->74->3082->3083->124->4725->1556->1358->3617->4304->1875->1876->1350->3001->3000->4196->2454->4346->143->3676->4013->1461->1078->4580->4579->3790->3789->3788->3787->4130->2687->2688->2980->351->350->4938->4939->4940->4941->2435->1289->1290->3946->3947->3401->3402->1363->4028->1960->1961->3197->337->1513->161->2792->4419->3431->3432->953->4311->4593

max BW with heap= 6832

time taken with heap= 0.1745 secs

Path = 406->1277->3171->1888->75->74->3082->3083->124->4725->1556->1358->3617->4304->1875->1876->1350->3001->3000->4196->4197->2225->1364->455->11->1343->2596->2040->3735->3736->1074->1576->298->297->2367->824->2624->465->2580->1348->993->3675->3674->4908->2464->2365->753->1458->2577->2674->2864->351->350->4938->4939->4940->4941->2435->1289->1290->3946->3947->3401->3402->1363->4028->1960->1961->3197->337->1513->161->2792->4419->3431->3432->953->4311->4593

max BW path using Kruskal= 6832

time taken with kruskal= 0.8916 secs

Final Results: -

Sparse Graph			Dense Graph		
Dijkstra's with Heap (s)	Dijkstra's without Heap (s)	Kruskal's with Heap Sort (s)	Dijkstra's with Heap (s)	Dijkstra's without Heap (s)	Kruskal's with Heap Sort (s)
0.1745	1.4002	0.8916	1.6914	2.3587	158.8112
0.0528	0.3879	0.8696	0.7061	2.0694	159.8894
0.09678	0.7988	0.8935	1.6834	1.4561	160.5163
0.1286	0.8447	0.8856	3.266	4.0900	154.6394
0.0159	0.0249	0.8765	1.4980	1.7702	154.7042
0.1655	1.3294	0.8297	2.5791	1.6406	154.0190
0.0030	0.0149	0.8407	2.1255	2.8942	154.4708
0.1785	1.4770	0.8308	3.1965	4.5697	154.699
0.1775	1.4341	0.8317	3.3061	4.7054	153.6011
0.1336	1.0404	0.8397	2.2280	3.2543	154.5047
0.1725	1.4780	0.8148	0.0837	2.2938	154.3382
0.1545	1.2137	0.8178	0.8498	3.5784	153.8811
0.1276	0.9265	0.8178	2.4344	2.2590	154.9146
0.1077	0.8561	0.8198	0.3850	2.4354	155.7454
0.1047	0.2323	0.8178	0.9096	3.6541	155.0713
0.1186	0.8686	0.8207	1.8699	2.6877	154.6155
0.0917	0.6343	0.8277	0.4677	0.5615	154.3073
0.1685	1.3823	0.8218	3.3121	4.6904	154.5208
0.0977	0.6522	0.8267	1.6735	3.4258	155.4981

0.1126	1.1012	0.8237	1.6495	0.9195	153.8216
0.1017	0.7379	0.8258	1.5240	4.6665	154.9186
0.0678	0.4587	0.8337	3.3560	4.8161	153.7528
0.1805	1.4421	0.8387	0.6782	1.6086	154.0022
0.0279	1.0113	0.8267	3.4198	4.8290	154.4908
0.1156	1.1530	0.8258	2.2718	3.1765	154.6923

7. Conclusion:

- We can see that the Dijkstra's with heap outperforms the performance of Dijkstra's without heap and Kruskal's for both sparse graph as well as dense graph. This is because it sorts using heap sort and it stops as soon as we find the target fringe.
- Kruskal's performance is good for sparse graph as the maximum spanning tree is also sparse and the number of edges is less. That is why, it is feasible to run Kruskal's in sparse graph. It runs faster than Dijkstra's without heap on an average. This makes sense as the run time for Kruskal is dependent on sorting which is $O(m \log m)$. In sparse graph, the 'm' is comparatively small and thus, the sorting is fast. Where as for dense graph, we can see that the performance of Kruskal degrades a lot. This is because 'm' is inherently large for dense graph.
- It does make sense to run Kruskal once to get the maximum spanning tree if you have a static graph and we have a lot of S-T pair. Once we have the maximum spanning tree in hand, we can calculate the Maximum Bandwidth Path between any two vertices using linear time algorithm such as DFS ($O(M + E)$). But as we know that these algorithms are designed for dynamic graph used in wireless communication systems, it is safe to assume that Kruskal is not feasible for dense dynamic graph.

- The performance of Dijkstra's algorithm without heap is average for both sparse as well as dense graph.
- **For sparse graph:**
Time taken by all 3 algorithms are as follows:
Dijkstra's with heap <<<< Kruskal's with heap sort <<<< Dijkstra's without heap (Average case).
- **For Dense Graph:**
Time taken by all 3 algorithms are as follows:
Dijkstra's with heap <<<< Dijkstra's without heap (Average case) <<<< Kruskal's with heap sort.

8.Future Improvement:

Merge sort can be used instead of heap sort because Merge sort on arrays has considerably better data cache performance, often outperforming heapsort on modern desktop computers because merge sort frequently accesses contiguous memory locations (good locality of reference); heapsort references are spread throughout the heap.