min-path

February 14, 2024

```
[2]: # !pip install networkx
# !pip install matplotlib
# !pip install tqdm
```

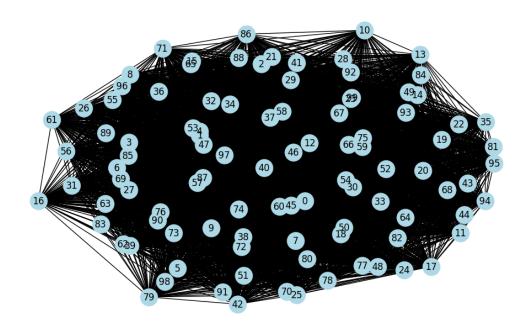
```
[3]: import random
import networkx as nx
import matplotlib.pyplot as plt
from itertools import combinations, groupby
```

0.1 Generating graph

```
[4]: # You can use this function to generate a random graph with 'num_of_nodes' nodes
     # and 'completeness' probability of an edge between any two nodes
     # If 'directed' is True, the graph will be directed
     # If 'draw' is True, the graph will be drawn
     def gnp_random_connected_graph(num_of_nodes: int,
                                     completeness: int,
                                     directed: bool = False,
                                     draw: bool = False):
         11 11 11
         Generates a random graph, similarly to an Erdős-Rényi
         graph, but enforcing that the resulting graph is conneted (in case of \Box
      \negundirected graphs)
         11 11 11
         if directed:
             G = nx.DiGraph()
         else:
             G = nx.Graph()
         edges = combinations(range(num_of_nodes), 2)
         G.add_nodes_from(range(num_of_nodes))
         for _, node_edges in groupby(edges, key = lambda x: x[0]):
             node_edges = list(node_edges)
             random_edge = random.choice(node_edges)
             if random.random() < 0.5:</pre>
```

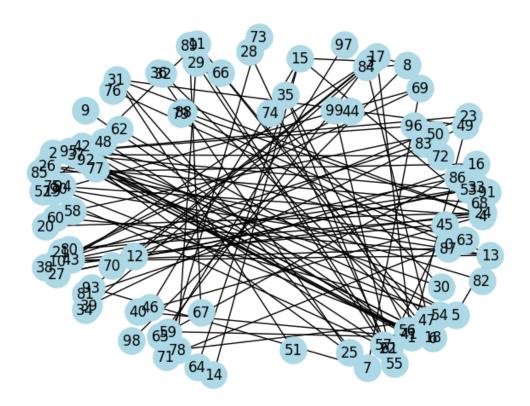
```
random_edge = random_edge[::-1]
    G.add_edge(*random_edge)
    for e in node_edges:
        if random.random() < completeness:</pre>
            G.add_edge(*e)
for (u,v,w) in G.edges(data=True):
    w['weight'] = random.randint(-5, 20)
if draw:
    plt.figure(figsize=(10,6))
    if directed:
        # draw with edge weights
        pos = nx.arf_layout(G)
        nx.draw(G,pos, node_color='lightblue',
                with_labels=True,
                node_size=500,
                arrowsize=20,
                arrows=True)
        labels = nx.get_edge_attributes(G,'weight')
        nx.draw_networkx_edge_labels(G, pos,edge_labels=labels)
    else:
        nx.draw(G, node_color='lightblue',
            with_labels=True,
            node_size=500)
return G
```

```
[5]: G = gnp_random_connected_graph(100, 1, False, True)
```



1 For Task 1

1.1 Kruskal's algorithm



```
[9]: mstk.edges(), len(mstk.edges())
```

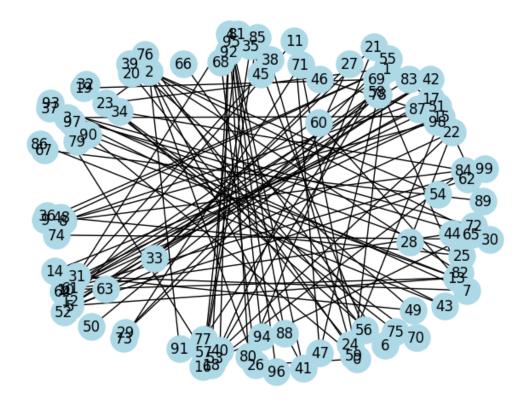
```
[9]: (EdgeView([(0, 26), (0, 71), (0, 76), (1, 9), (1, 17), (1, 23), (1, 40), (1, 59), (1, 77), (1, 48), (2, 23), (2, 26), (2, 30), (2, 41), (2, 54), (2, 56), (2, 82), (2, 91), (3, 15), (3, 43), (3, 65), (3, 93), (4, 31), (4, 40), (4, 80), (4, 88), (4, 94), (5, 13), (5, 42), (5, 58), (5, 62), (6, 75), (6, 32), (6, 66), (7, 31), (7, 69), (7, 93), (8, 17), (8, 84), (8, 98), (10, 13), (10, 44), (10, 45), (10, 69), (10, 83), (10, 87), (11, 47), (11, 52), (12, 17), (13, 60), (13, 79), (13, 90), (14, 15), (15, 24), (15, 64), (15, 46), (16, 22), (16, 65), (16, 81), (16, 85), (18, 37), (18, 84), (18, 92), (19, 69), (19, 75), (20, 24), (21, 33), (21, 80), (22, 28), (22, 52), (22, 89), (23, 70), (24, 38), (25, 36), (25, 37), (26, 68), (27, 33), (27, 53), (29, 55), (29, 78), (30, 74), (31, 49), (31, 51), (34, 84), (35, 56), (35, 57), (37, 72), (39, 49), (41, 95), (42, 61), (42, 63), (43, 86), (47, 78), (50, 58), (59, 97), (61, 99), (67, 89), (71, 73), (95, 96)]),
```

```
[10]: def kruskal(graph):
    if isinstance(graph, nx.DiGraph):
        G = nx.DiGraph()
```

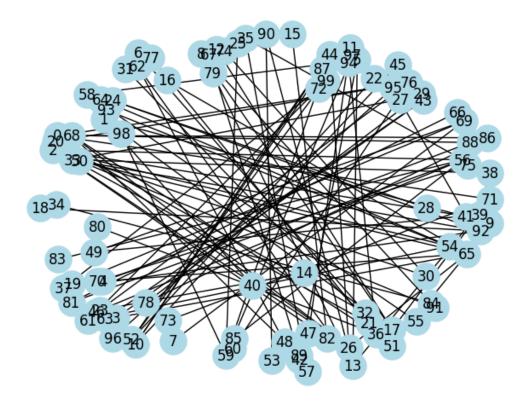
```
else:
    G = nx.Graph()
G.add_nodes_from(graph.nodes)
edges = sorted(graph.edges(data=True), key=lambda x: x[2]['weight'])
parent = {node: node for node in graph.nodes}
def find(node):
    if parent[node] != node:
        parent[node] = find(parent[node])
    return parent[node]
def union(u, v):
    root_u = find(u)
    root_v = find(v)
    parent[root_u] = root_v
for u, v, w in edges:
    if find(u) != find(v):
        G.add_edge(u, v, weight=w['weight'])
        union(u, v)
nx.draw(G, node_color='lightblue',
            with_labels=True,
            node_size=500)
return G
```

```
[11]: kruskal(G)
```

[11]: <networkx.classes.graph.Graph at 0x12151e3d0>



1.2 Prim's algorithm



```
[14]: mstp.edges(), len(mstp.edges())

[14]: (EdgeView([(0, 26), (0, 71), (0, 76), (1, 17), (1, 9), (1, 40), (2, 26), (2, 26), (2, 26), (3, 26), (4, 27), (4, 27), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4, 28), (4
```

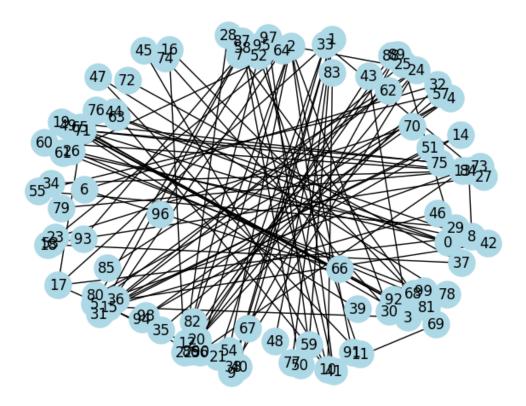
```
[14]: (EdgeView([(0, 26), (0, 71), (0, 76), (1, 17), (1, 9), (1, 40), (2, 26), (2, 30), (2, 41), (2, 54), (2, 56), (2, 82), (2, 91), (3, 65), (3, 43), (4, 88), (5, 13), (5, 42), (5, 58), (6, 32), (7, 69), (8, 17), (8, 84), (10, 44), (10, 45), (10, 87), (11, 52), (11, 47), (12, 17), (13, 71), (13, 62), (13, 79), (13, 90), (14, 23), (15, 82), (16, 65), (17, 71), (18, 92), (19, 69), (20, 76), (20, 32), (21, 33), (22, 52), (23, 76), (24, 92), (24, 28), (25, 59), (26, 44), (26, 65), (26, 68), (27, 33), (29, 78), (30, 74), (31, 51), (33, 65), (33, 38), (34, 84), (35, 56), (35, 57), (36, 68), (36, 77), (36, 80), (36, 94), (37, 92), (37, 72), (39, 49), (41, 95), (43, 63), (44, 52), (44, 59), (46, 56), (48, 62), (49, 69), (50, 92), (51, 65), (53, 90), (54, 61), (54, 64), (55, 58), (56, 81), (56, 83), (56, 85), (59, 97), (59, 98), (60, 76), (61, 99), (65, 70), (65, 93), (66, 81), (67, 89), (68, 86), (68, 88), (68, 89), (69, 71), (70, 75), (71, 73), (71, 92), (78, 86), (95, 96)]),
```

```
[15]: def prim(graph):
    U = set()
    V_U = set(graph.nodes)
```

```
start_vertex = next(iter(V_U))
U.add(start_vertex)
V_U.remove(start_vertex)
if isinstance(graph, nx.DiGraph):
    G = nx.DiGraph()
else:
    G = nx.Graph()
while V_U:
    min_weight = float('inf')
    min_edge = None
    for u in U:
        for v in V_U:
            if graph.has_edge(u, v):
                weight = graph[u][v]['weight']
                if weight < min_weight:</pre>
                    min_weight = weight
                    min_edge = (u, v)
    G.add_edge(*min_edge, weight=min_weight)
    U.add(min_edge[1])
    V_U.remove(min_edge[1])
nx.draw(G, node_color='lightblue',
            with_labels=True,
            node_size=500)
return G
```

```
[16]: prim(G)
```

[16]: <networkx.classes.graph.Graph at 0x126b60e10>



1.3 Example on time measuring

Read more on this: https://realpython.com/python-timer/

Recall that you should measure times for 5, 10, 20, 50, 100, 200, 500 nodes 1000 times (and take mean of time taken for each node amount).

Then you should build the plot for two algorithms (x - data size, y - mean time of execution).

```
[17]: import time from tqdm import tqdm
```

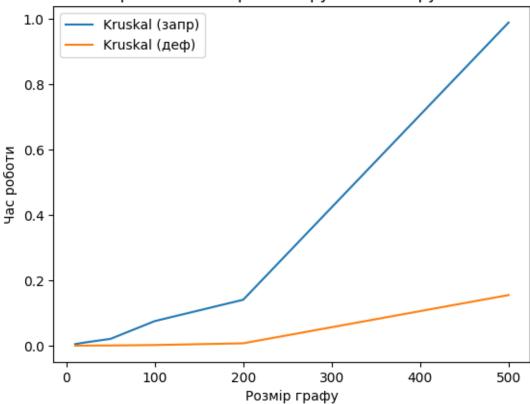
```
[18]: NUM_OF_ITERATIONS=100
def count_time(function, size):
    time_taken = 0
    for _ in tqdm(range(NUM_OF_ITERATIONS)):

    # note that we should not measure time of graph creation
    G = gnp_random_connected_graph(size, 0.4, False)

    start = time.time()
    function(G)
```

```
end = time.time()
              time_taken += end - start
          return time_taken / NUM_OF_ITERATIONS
     Kruscal's algorithm
     1)
                                                               find()
                                              Union
                                                                                    find
     union,
                                                   100
[20]: def kruskal_def(G):
          return tree.minimum_spanning_tree(G, algorithm="kruskal")
      graph_sizes = [10, 20, 50, 100, 200, 500]
      kruskal_mine = [count_time(kruskal, size) for size in graph_sizes] #
                           (
      kruskal_defined = [count_time(kruskal_def, size) for size in graph_sizes]
      plt.clf()
      plt.plot(graph_sizes, kruskal_mine, label='Kruskal ( )')
      plt.plot(graph_sizes, kruskal_defined, label='Kruskal ( )')
      plt.xlabel('
                         ')
      plt.ylabel('
                        ')
      plt.title('
                                          ')
      plt.legend()
      plt.show()
       0%1
                     | 0/100 [00:00<?, ?it/s]100%|
                                                        | 100/100 [00:00<00:00,
     211.66it/s]
     100%
                | 100/100 [00:00<00:00, 107.34it/s]
                | 100/100 [00:02<00:00, 46.76it/s]
     100%|
                | 100/100 [00:07<00:00, 12.93it/s]
     100%|
     100%|
                | 100/100 [00:18<00:00, 5.38it/s]
     100%|
                | 100/100 [01:43<00:00, 1.04s/it]
                | 100/100 [00:00<00:00, 12655.17it/s]
     100%|
                | 100/100 [00:00<00:00, 4502.69it/s]
     100%|
                | 100/100 [00:00<00:00, 921.07it/s]
     100%|
                | 100/100 [00:00<00:00, 258.62it/s]
     100%|
     100%|
                | 100/100 [00:01<00:00, 66.76it/s]
                | 100/100 [00:20<00:00, 4.90it/s]
     100%|
```

Порівняння алгоритмів Крускала та Крускала



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Prim's algorithm

2)

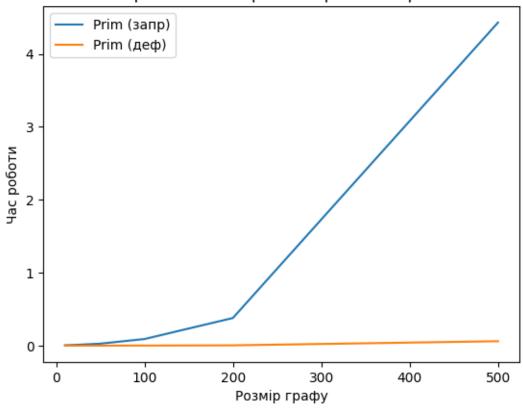
2 , - , - ,

, , , , ,

```
plt.ylabel(' ')
plt.title(' ')
plt.legend()
plt.show()
```

```
100%|
          | 100/100 [00:00<00:00, 213.51it/s]
          | 100/100 [00:00<00:00, 110.12it/s]
100%|
100%|
          | 100/100 [00:02<00:00, 36.15it/s]
          | 100/100 [00:09<00:00, 10.75it/s]
100%|
100%|
          | 100/100 [00:41<00:00, 2.43it/s]
100%|
          | 100/100 [07:28<00:00, 4.48s/it]
          | 100/100 [00:00<00:00, 14790.55it/s]
100%|
          | 100/100 [00:00<00:00, 5366.85it/s]
100%|
100%|
          | 100/100 [00:00<00:00, 1110.12it/s]
100%|
          | 100/100 [00:00<00:00, 325.13it/s]
100%|
          | 100/100 [00:01<00:00, 81.07it/s]
100%|
          | 100/100 [00:12<00:00, 8.14it/s]
```

Порівняння алгоритмів Прима та Прима



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