Performance comparison of Kruskal's and Prim's MST algos

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Task and experiment description

This experiment's purpose is to demonstrate work of Prim's and Kruskal's algorithms, show one of their implementations and measure their performance depending on number of nodes and edges in graph.

The experiment measures mean time of 1000 finding MST of graphs in different configurations.

For each algo was performed 1000 tries for each combination of (10, 20, 50, 100, 200, 500) nodes and (0.1, 0.3, 0.5, 0.7, 0.9) probabilities of certain edge being in graph.

Then performance of every graph configuration (num of nodes and probability) is compared between our observed algorithms. Based on this information the conclusion about performance and scenarios of usage are made.

Computer Specifiction

- processor: AMD Ryzen 7 5700U

- number of cores: 8

- frequency: 1.8 - 4.3 GHz

- memory: 15.08 GiB

- os: Arch Linux

- kernel: Linux 5.15.7-arch1-1

Kruskal's Algorithm

```
In [ ]: def kruskal(graph):
            edges = graph[1]
            nodes = graph[0]
            components = [set(node) for node in list(map(lambda x: [x], nodes))]
            edges.sort(key=lambda x: x[2])
            component1 = 0
            component2 = 1
            for edge in edges:
                component1 = component2 = set()
                for component in components:
                     if edge[0] in component:
                         component1 = component
                     elif edge[1] in component:
                         component2 = component
                 if component1 != component2 and component1!=set() and component2!=set()
                     component1.update(component2)
                     components.remove(component2)
                     tree.append(edge)
                 if len(tree) == len(nodes) - 1:
                    break
            return tree
```

Prim's Algorithm

```
In [1]: from heapq import heappop, heappush
        def prim_heaps(graph):
            mst = []
            used_verts = set()
            num_verts = len(graph[0])
            edges = [[] for _ in range(num_verts)]
            for edge in graph[1]:
                 if edge[0] == edge[1]: continue
                 heappush(edges[edge[0]], (edge[2], edge[1]))
                 heappush (edges[edge[1]], (edge[2], edge[0]))
            cost, dest = 0, 1
            while len(used_verts) < num_verts:</pre>
                 smallest_edge_vert = 0
                 for vert in used_verts:
                     while(len(edges[vert])) > 0 and edges[vert][0][dest] in used_verts
                         heappop (edges [vert])
                     if len(edges[vert]) == 0: continue
                     if len(edges[smallest_edge_vert]) == 0 or edges[vert][0][cost] < ed</pre>
                         smallest_edge_vert = vert
                 edge = heappop(edges[smallest_edge_vert])
                mst.append((smallest_edge_vert, edge[dest], edge[cost]))
                used_verts.add(smallest_edge_vert)
                 used_verts.add(edge[dest])
             return mst
```

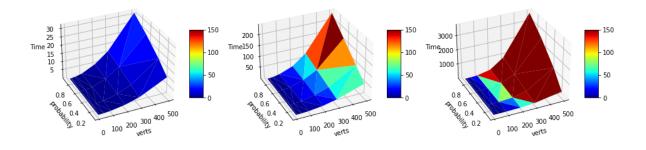
```
In [ ]: def prim(graph):
            edges = graph[1]
            nodes = graph[0]
            tree = []
            connected = {edges[0][0]}
            for i in range(len(nodes) - 1):
                min_edges = []
                for edge in edges:
                    if (edge[0] in connected) and (edge[1] not in connected):
                        min_edges.append(edge)
                    elif (edge[1] in connected) and (edge[0] not in connected):
                        min_edges.append(edge)
                min_edge = sorted(min_edges, key = lambda x: x[2])[0]
                tree.append(min_edge)
                 connected.add(min_edge[2])
            return tree
```

Experiments environment

```
In [6]: import random
        import networkx as nx
        import matplotlib.pyplot as plt
        import time
        from itertools import combinations, groupby
        from tqdm import tqdm
        def gnp_random_connected_graph(num_of_nodes: int,
                                        completeness: int,
                                        draw: bool = False) -> list[tuple[int, int]]:
            Generates a random undirected graph, similarly to an Erdős-Rényi
            graph, but enforcing that the resulting graph is conneted
            edges = combinations(range(num_of_nodes), 2)
            G = nx.Graph()
            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)
                 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(0, 10)
            if draw:
                plt.figure(figsize=(10, 6))
                nx.draw(G, node_color='lightblue',
                with_labels=True,
                         node_size=500)
            return G
```

```
NUM_OF_ITERATIONS = 1000
        def performance_test(num_vert, probability, method):
            time_taken = 0
            for i in tqdm(range(NUM_OF_ITERATIONS)):
                 # note that we should not measure time of graph creation
                G = gnp_random_connected_graph(num_vert, probability, False)
                edges = list(map(lambda x: (x[0], x[1], x[2]['weight'])), G.edges.data()
                nodes = list(G.nodes)
                my_graph = (nodes, edges)
                if method == 'kruskal':
                    f = kruskal
                elif method == 'prim':
                    f = prim
                elif method == 'prim_heaps':
                    f = prim_heaps
                start = time.time()
                f (my_graph)
                end = time.time()
                time_taken += end - start
            time_taken / NUM_OF_ITERATIONS
            return time_taken
In []: verts = [10, 20, 50, 100, 200, 300, 500]
        probs = [0.1, 0.3, 0.6, 0.9]
```

General plot for performnace depending on graphs' size



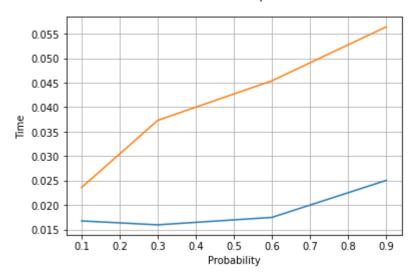
In the left we see Kruskal's algorithm. In the center - Prim's algorithm using heaps. In the right - primitive prim's Algorithm.

Performance of algorithms for graphs with 10 nodes

On each plot orange line is the performance of Kruskal's algo, and the blue one is Prim's algo

---plot for 2 algos where x - probability and y is time

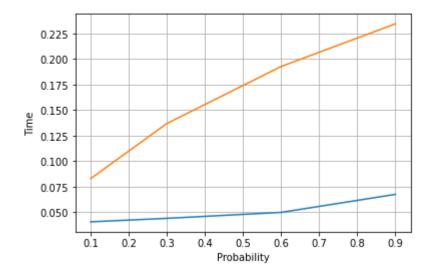




Performance of algorithms for graphs with 20 nodes

---plot for 2 algos where x - probability and y is time

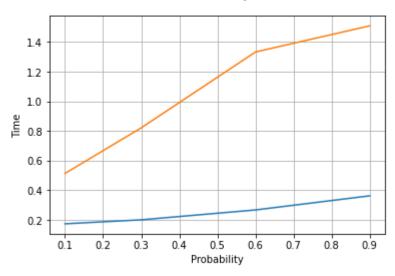
20 verts minimal span tree



Performance of algorithms for graphs with 50 nodes

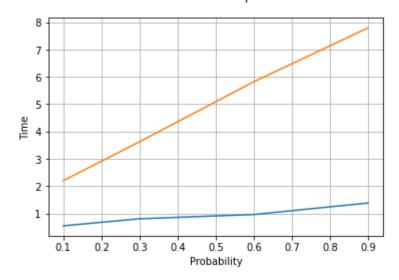
---plot for 2 algos where x - probability and y is time

50 verts minimal span tree



Performance of algorithms for graphs with 100 nodes

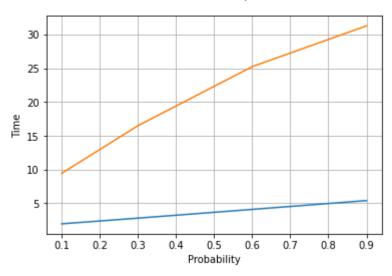
---plot for 2 algos where x - probability and y is time 100 verts minimal span tree



Performance of algorithms for graphs with 200 nodes

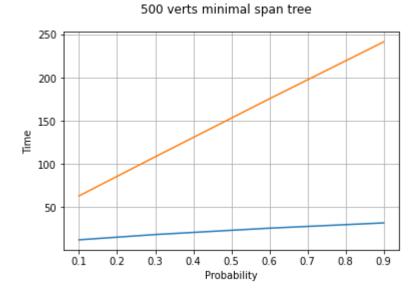
---plot for 2 algos where x - probability and y is time

200 verts minimal span tree



Performance of algorithms for graph with 500 nodes

---plot for 2 algos where \boldsymbol{x} - probability and \boldsymbol{y} is time



Result

In general, it can be seen that each algorithm coped with the task in a reasonable amount of time, this is because the ammount of vertices was not huge, also Kruskal's algorithm did its job way faster, this is due to appropriate types of data storage and manipulation available in python, algorithm's realisation utilizes language's features. The same on the other hand can not be told about Prim's algorithm. In order for it to work properly we had to use structure called binary heaps. Without them time complexity of this algorithm scales abnormally and firstly it took us about 2 hours to run all tests for Prim's algorithm, with the heap implementation all tests took about 20 minutes.