Project 3 Objective 2: Investigation of Dijkstra's and Bellman-Ford Implementations for the

Shortest Path Algorithm for a graphical network between the Middelburg railway station

and UCR

Shaivan Bhagat

SCI COMP 302 Algorithms and Data Structures

This is all my own work. I have not knowingly allowed others to copy my work. This work has not

been submitted for assessment in any other context.

1. Description

For this project, I am investigating the graph algorithm with the Shortest Path (SP) of the street

networks between the Middelburg railway station and University College Roosevelt. For the

Shortest Path, the chosen implementations are Dijkstra and Bellman Ford. They are both used to

find the shortest paths in a weighted graph. Dijkstra's algorithm is a greedy algorithm that finds the

shortest path from a source node to all other nodes in the graph. It iteratively selects the node with

the smallest tentative distance and updates the distances of its neighbors. Bellman-Ford algorithm

works for graphs with both positive and negative weights. It iterates through all edges multiple

times, relaxing the edges to find the shortest paths. It handles negative weights (unlike Dijkstra's)

but can detect negative cycles in the graph. If there are no negative cycles, it provides the shortest

paths.

2. Experimental Platform

The algorithms were run on a MacBook Air running on a M2 Chip. General specifications of the

hardware are:

Total Number of Cores: 8 (4 performance & 4 efficiency)

Memory: 8 GB 0

Chipset Model: Apple M2

Type: CPU

Bus: Built-In

The algorithms were run completely through Python. Python files were run via Spyder,

accessed through Anaconda-Navigator 2.4.2. The specific versions of the software are as

follows:

o Spyder version: 5.4.3 (conda)

o Python version: 3.11.4 64-bit

o Qt version: 5.15.2

o PyQt5 version: 5.15.7

o Operating System: Darwin 22.5.0

Whilst the algorithms were being run, all cores were exploited, and CPU load was dedicated

specifically towards enhancing the performance of the algorithms.

3. Experimental Components

a. Null Hypothesis

The implementation Dijkstra for the algorithm of the Shortest Path problem will perform better

(time complexity) than the Bellman-Ford implementation for the same problem.

b. Time & Space Complexity

Number of runs for the two implementations: 10.

Big-Oh (Complexity):

Dijkstra's Algorithm: O((V + E) * log(V)) with a binary heap as the priority queue; when

implemented with a binary heap as the priority queue, has a time complexity of O((V + E) *

log(V)), where V is the number of vertices and E is the number of edges. This complexity arises

from the need to update the priority queue and extract the minimum element during each iteration.

Bellman-Ford Algorithm: O(V * E) (worst case); where V is the number of vertices and E is the

number of edges. The worst-case time complexity is often considered to be O(V * E) because the

algorithm iterates over all edges for a maximum of V-1 passes. In practice, the algorithm often

terminates earlier if no negative cycles are present.

Space Complexity: O(V) for both Dijkstra's and Bellman-Ford; primarily used for storing distances and predecessor information.

Time complexity in python was measured using the following pseudocode:

```
import time

start_time = time.time()

end_time = time.time()

elapsed_time = end_time - start_time

print(f"Elapsed time: {elapsed_time} seconds")
```

c. Experiment & Code

Independent Variable(s):

- Graph Structure (G): The structure of the graph, including nodes, edges, and their attributes.
 This is the input to the algorithm and is independent because you can choose or generate different graphs to analyze.
- o Source and Target nodes: The nodes between which the shortest path is calculated.

Dependent Variables(s):

- Shortest Path: The sequence of nodes representing the shortest path from the source to the target. This is the main output of the algorithm, and it depends on the input graph structure and the selected source and target nodes.
- Elapsed Time: The time taken by the algorithm to calculate the shortest path. This time is dependent on factors such as the size and complexity of the graph.

Time was measured in <u>microseconds (µs)</u> as the amount taken to find and print the shortest possible path between the Middelburg railway station and UCR. This was then used as the basis for comparison of performance and complexity.

- O In the following code, the nx.shortest_path function is used with the specified method set to "dijkstra." This function computes the shortest paths in the graph using Dijkstra's algorithm. The paths are then stored in the paths variable, and the elapsed time is measured for performance evaluation.
- The shortest_path function takes a graph (G), source and target nodes (source and target), and an optional weight parameter.
- The function calls nx.shortest_path with the specified method set to 'dijkstra'. This function computes the shortest paths in the graph using Dijkstra's algorithm.
- The calculated shortest path is stored in a variable and the elapsed time is measured for performance evaluation. The results, including the elapsed time and the shortest path, are printed.

```
/Users/shaivan/Downloads/Project3/Project3_ShaivanBhagat.py
× jkl.py
                  X Project3_ShaivanBhagat.py
                                                       × QWE.py
                                                                     × xyz.py
           # Implementation of Djikstra's algorithm
           def has_path(G, source, target):
                    nx.shortest_path(G, orig, dest)
                except nx.NetworkXNoPath:
                    return False
                return True
          def shortest_path(G, source=None, target=None, weight=None, method="dijkstra"):
    if method not in ("dijkstra", "bellman-ford"):
        # so we don't need to check in each branch later
   49
50
51
52
53
54
55
56
57
58
               raise ValueError(f"method not supported: {method}")
method = "unweighted" if weight is None else method
               if source is None:
                    if target is None:
                         # Find paths between all pairs.
                         if method == "unweighted":
                             paths = dict(nx.all_pairs_shortest_path(G))
   59
60
                         elif method == "dijkstra":
                              paths = dict(nx.all_pairs_dijkstra_path(G, weight=weight))
                         else: # method == 'bellman-ford
   62
63
                              paths = dict(nx.all_pairs_bellman_ford_path(G, weight=weight))
                    else:
                         # Find paths from all nodes co-accessible to the target.
                         if G.is_directed():
                              G = G.reverse(copy=False)
   67
68
                         if method == "unweighted":
                             paths = nx.single_source_shortest_path(G, target)
   69
70
71
72
73
74
                         elif method == "dijkstra":
                         paths = nx.single_source_dijkstra_path(G, target, weight=weight)
else: # method == 'bellman-ford':
                              paths = nx.single_source_bellman_ford_path(G, target, weight=weight)
                         # Now flip the paths so they go from a source to the target.
                         for target in paths:
   75
76
77
78
79
80
                              paths[target] = list(reversed(paths[target]))
                else:
                    if target is None:
                         # Find paths to all nodes accessible from the source.
if method == "unweighted":
                              paths = nx.single_source_shortest_path(G, source)
                         elif method == "dijkstra":
                              paths = nx.single_source_dijkstra_path(G, source, weight=weight)
   83
84
                                                'bellman-ford
                         else: # method ==
                              paths = nx.single_source_bellman_ford_path(G, source, weight=weight)
                    else:
                         # Find shortest source-target path.
                         if method == "unweighted":
   88
89
                         paths = nx.bidirectional_shortest_path(G, source, target)
elif method == "dijkstra":
                         _, paths = nx.bidirectional_dijkstra(G, source, target, weight) else: # method == 'bellman-ford':
   91
92
                             paths = nx.bellman_ford_path(G, source, target, weight)
                return paths
   94
95
           # Timing the shortest path calculation
           start_time = time.time()
           shortest_path(G, source=orig, target=dest, weight="travel_time", method="dijkst|ra")
           elapsed_time = time.time() - start_time
print(f"Shortest path time Djikstra: {elapsed_time} seconds")
```

The bellman_ford_shortest_path function is defined, which uses the NetworkX function nx.shortest_path with the method='bellman-ford' parameter to find the shortest path between the specified source and target nodes in the graph G. The timing of the Bellman-Ford algorithm is measured by recording the start time (start_time_bellman_ford) before

calling the function and then calculating the elapsed time (elapsed_time_bellman_ford) after the function has been executed. The shortest path and the time taken by the Bellman-Ford algorithm are printed.

```
# Implementation of Bellman-Ford algorithm

def bellman_ford_shortest_path(G, source, target, weight=None):
    return nx.shortest_path(G, source=source, target=target, weight=weight, method='bellman-ford')

# Timing the shortest path calculation using Bellman-Ford algorithm

start_time_bellman_ford = time.time()

bellman_ford_path = bellman_ford_shortest_path(G, source=orig, target=dest, weight='travel_time')

elapsed_time_bellman_ford = time.time() - start_time_bellman_ford

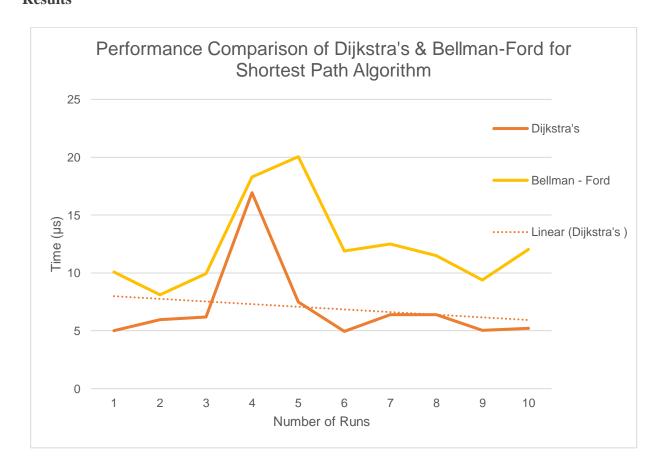
print(f"Bellman-Ford Shortest path time: {elapsed_time_bellman_ford} seconds")

print("Bellman-Ford Shortest path:", bellman_ford_path)

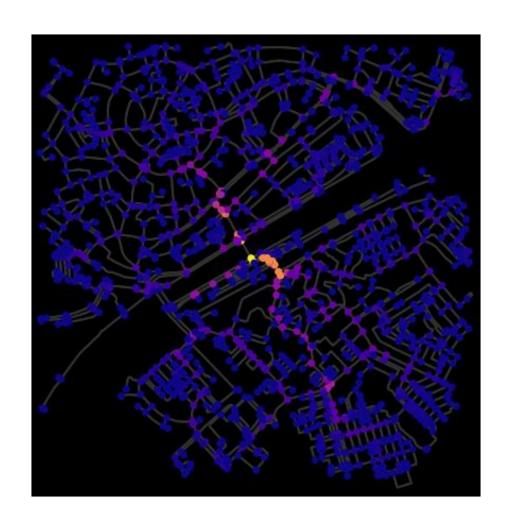
def shortest_path(G, source=None, target=None, weight=None, method="bellman-ford"):
    if method not in ("dijkstra", "bellman-ford"):
    # so we don't need to check in each branch later
```

raise ValueError(f"method not supported: {method}")
method = "unweighted" if weight is None else method

Results



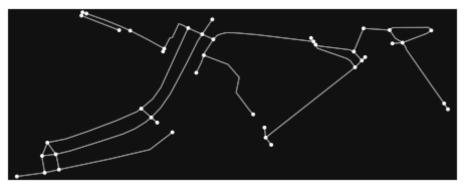
Run	Dijkstra's	Bellman - Ford
1	5,0067	10,0804
2	5,9604	8,1106
3	6,1988	9,948
4	16,9380	18,29
5	7,4997	20,048
6	4,9508	11,89579
7	6,408	12,4897
8	6,4038	11,4809
9	5,048	9,4039
10	5,209	12,049











d. Static Analysis & Refactoring

Through Source > Run code analysis, static analysis can be performed for the python file using pylint.

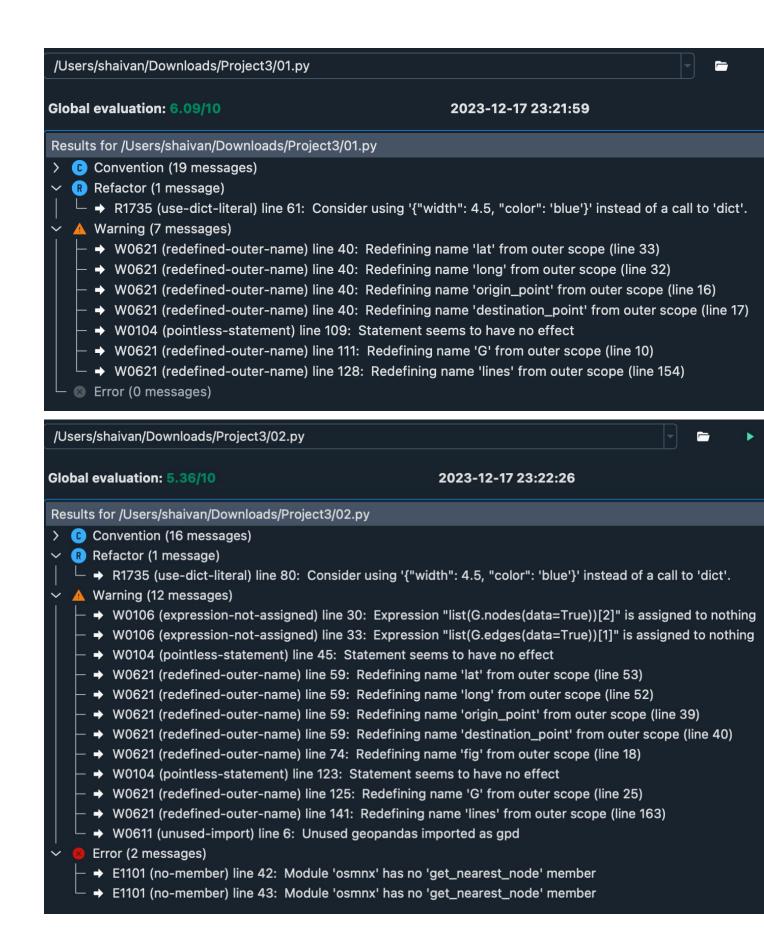
Most of the errors are convention errors which requires to name the attributes, variables, and arguments according to the python naming convention. The refractor error can be ignored. However, the error message can lead to issues with implementation.

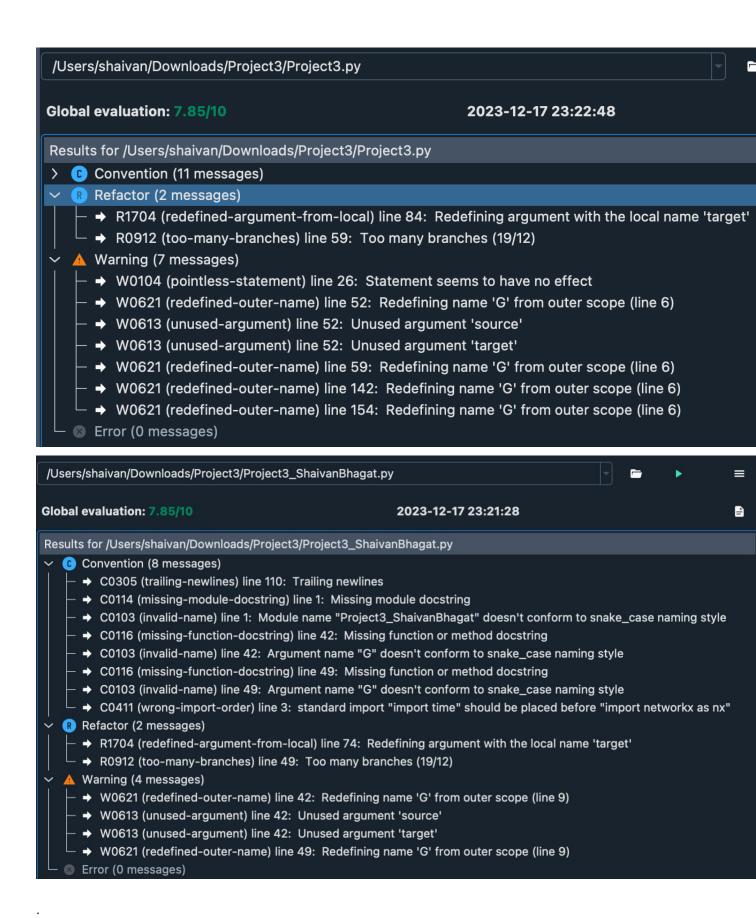
Refactoring would involve restructuring the code to enhance its readability, maintainability, and/or performance without modifying its behavior. Redundant code could be made more modular.

Techniques that could be applied to refactor code could possibly include simplifying conditional statements, modularization, adhering to naming conventions and PEP 8 standards, reducing complexity, and improving readability.

- o Error handling: 'has path' function is defined but not used. Removing this is suggested.
- Code duplication: The graph is being saved in different formats, which can be avoided by using only one specific format to avoid duplication.
- Time measurement: instead of manually converting the time taken for the implementations to return the answers, the 'timeit' module can and should be used for more accurate and convenient timing measurements
- Function parameter: The shortest_path function takes parameters named source and target, but the function body uses orig and dest. It would be more consistent and clear to use source and target throughout.
- O Although there are errors present in the code, as seen from the sensitivity analysis, most errors can be neglected. The only one which needs to be taken into consideration is for the second file of code with 2 errors where the module does not recognize the required function.

However, a solution for that is presented below as well.





```
origin_point = (51.495446, 3.618127)
destination_point = (51.499318, 3.610658)

origin_node = ox.distance.nearest_nodes(G, X=origin_point[1], Y=origin_point[0])
destination_node = ox.distance.nearest_nodes(G, X=destination_point[1], Y=destination_point[0])
```

4. Conclusion

Based on the results that I have obtained; I can verify the null hypothesis and conclude that Dijkstra's algorithm for the Shortest Path (SP) problem performs better compared to the Bellman-Ford implementation. Sensitivity and static code analysis has been presented as well, along with some possible options for code refactoring. Results along with the graph show that a shortest path between the Middelburg railway station and UCR is indeed present and can be calculated using the two implementation algorithms used in this case. Techniques like edge contraction, node ordering, and other preprocessing steps can be applied to simplify the graph structure and improve algorithm efficiency. For scenarios where the graph is dynamic (edges or nodes change over time), algorithms designed to handle dynamic graphs efficiently can be beneficial. On current hardware, leveraging GPU or specialized hardware can accelerate graph algorithms. Although here only the shortest path (SP) was considered, utilizing Minimum Spanning Tree (MST) to solve similar problems and to create an optimal graph network between the Middelburg railway station and UCR should be considered.

References

- https://apurv.page/plotthepath.html
- https://osmnx.readthedocs.io/en/stable/
- https://geoffboeing.com/2016/11/osmnx-python-street-networks/
- $\bullet \quad https://github.com/gboeing/osmnx-examples/blob/main/notebooks/00-osmnx-features-demo.ipynb$
- $\verb| https://networkx.org/documentation/stable//auto_examples/geospatial/plot_osmnx.html \# sphx-glr-download-auto-examples-geospatial-plot-osmnx-py \\$
- https://networkx.org/
- https://www.pylint.org/