

PROJECT

Advanced Algorithmic and Programming

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1 Creating a graph from GTFS data

1.1 Importing relevant GTFS data

Our city of choice for the project is Phoenix, in Arizona.

We use the public data available here: <https://transitfeeds.com/p/valley-metro/68/latest>

The difficulty here was to import the relevant data and convert it to a graph. Some Python libraries seem to exist but none was convenient for the project, so we needed to transform the data manually. After reading documentation on GTFS, we needed only two files from the data feed: `stops.txt` and `stop_times.txt`.

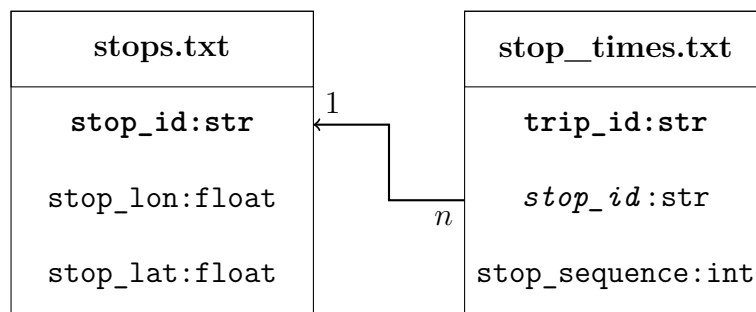


Figure 1 – Database representation of the relevant GTFS files

The nodes of the graph are directly given by the file `stops.txt`, so we can just parse the file to import them. This is done in the `import_stops` function of `gtfs.py`.

Whereas the edges need some operations:

1. Parse the file
2. Regroup in order the stops in the same trips
3. Create edges between consecutive stops in a trip

This is done in the `import_edges` function of `gtfs.py`.

1.2 Creating a graph

The `Graph` class is in `graph.py`.

The class supports weighted directed graphs, but unweighted graphs can also be created. The class constructor accepts an optional parameter which is a callback to compute weight from two

given nodes. In our case, this function takes two `Stop` (defined in [gtfs.py](#)) instances and returns the Euclidian distance between them:

$$\text{compute_weight} : (s, s') \mapsto \sqrt{(s'_{\text{lat}} - s_{\text{lat}})^2 + (s'_{\text{lon}} - s_{\text{lon}})^2}$$

If this parameter is not passed, all edge weights will be set to 0.

After testing the pathfinding methods (see [subsection 2.2](#)), we realized that our way of storing neighbors was not optimized.

Indeed we stored the neighbors in a huge adjacency list. So when we needed to compute some neighbors, we had to search in the whole list for neighbors.

So we optimized this aspect. We are now storing the neighbors of a node inside the `Node` object itself. This implementation has redundancy in memory, but fetching the neighbors of a node is now $\mathcal{O}(1)$ for the `neighbors_out` and `neighbors_in` methods.

But our graph was just too big at that time to be able to work efficiently on it.

So we restricted the graph to a district of the city, so the graph's order and size have been divided by 8.

1.3 Results

- 7982 lines from `stops.txt` were imported to 891 nodes (7863 before restricting the graph) in the graph.
- 1720661 lines from `stop_times.txt` were imported to 975 edges (8319 before the restriction) in the graph.

	import_stops	import_edges	Graph
Before optimizations	38.61	4211.02	78.64
	31.75	5036.29	49.05
	44.50	5085.64	68.83
	28.56	3786.03	43.43
	26.48	4070.86	43.62
After neighbors optimization	27.24	4931.54	155.09
	29.34	3610.81	74.15
	49.93	4096.88	115.51
	28.70	3279.42	86.15
	26.39	3685.92	82.94

Table 1 – Execution time (ms) for the graph creation over several tries

We see in this [Table 1](#) that after the neighbors optimization described in [subsection 1.2](#), the execution time for the `Graph` instantiation has increased from an average of 56.71 ms to 102.77 ms. This can be explained by the new storage method of neighbors: we need to add two different tuples into two different sets.

2 Finding the shortest paths

2.1 Common interface

To ease the usage of pathfinding for future computations, we created a common interface which is the `Pathfinder` class of `pathfinding.py`.

We used the *Strategy design pattern* here with the `method` property, which is assigned in the class constructor to a function with the following signature:

$$\text{method} : (\text{pathfinder} : \text{Pathfinder}, \text{start} : \text{int}) \rightarrow \text{None}$$

This function is to compute all shortest paths from `start` and store them in `pathfinder`, using its `previous` and `distance` dictionaries.

2.2 Pathfinding methods

We implemented two methods for the `common interface`: `bfs` and `dijkstra`, defined in `pathfinding.py`.

Both use a queue to define the next nodes to be traversed, and proceed until this queue is empty. For `dijkstra`, this queue is actually a priority queue, sorted by ascending distance to `start`.

For each outward neighbor n of the current traversed node c , we initialize the relevant sections of `pathfinder` if needed, and we update them with the current neighbor traversed:

bfs if $d(\text{start}, n) = d(\text{start}, c) + 1$, then we add c to the backtrack list of n .

dijkstra if $d(\text{start}, n) \leq d(\text{start}, c) + w$ with w the weight of the edge (c, n) , then we add c to the backtrack list of n . And if $d(\text{start}, n) < d(\text{start}, c) + w$, then we reset the backtrack list and distance.

To optimize these methods, at the time when our graph was not restricted, we replaced the queues (which were simple lists) by heap queues using the `heapq` module of Python. This also improved the correctness of our methods, since it allows to pop nodes from the queue in a consistent order.

2.3 Results

	bfs	dijkstra
Before optimizations	1.54	2.05
	1.48	2.26
	2.72	3.51
	2.37	2.88
	2.15	2.33
After heaps optimization	1.92	4.09
	3.06	5.40
	2.97	5.77
	3.36	8.08
	2.71	4.38

Table 2 – Execution time (ms) for the pathfinding methods over several tries

We see in this [Table 2](#) that after the heaps optimization, the execution times have increased from an average of 2.05 ms to 2.80 ms for `bfs` and from 2.61 ms to 5.54 ms for `dijkstra`. Even though this optimization helped a lot when our graph was not restricted to a single district, it seems that using heap queues for little graphs does not help.

3 Finding clusters

Clustering described in this section is implemented in the `clustering` function, defined in [clustering.py](#).

3.1 Edge betweenness

Edge betweenness is a good indicator of clusters. Indeed, the edges with the highest betweenness can be seen as the bridges linking the clusters. Thus, destroying a number of these edges can separate these clusters, highlighting them.

Thus, the main component of the clustering algorithm is a loop (lines 13 to 117) that, at each iteration, identifies the clusters in the graph, computes the betweenness of each edge, spots the highest ones and deletes them. The loop repeats that operation until the good number of clusters is obtained (usually 5 to 15 iterations).

The betweenness of an edge is the number of shortest paths connecting nodes that use this edge. So, in order to obtain it, the algorithm must find every shortest path connecting nodes in the graph. To do so, the algorithm first create an empty betweenness list. Then, it uses two nested loops:

- The first one (lines 24 to 96) iterates over every node contained in the graph. Each retrieved node is used as a starting point.

- Then, the algorithm uses the second loop (lines 49 to 96) to iterate over every other node contained in the graph.

At each iteration the algorithm use a `Pathfinder` instance (described earlier in [subsection 2.1](#)) to check if there is a path between the starting and target nodes (line 52). If there is at least one path, the program iterates on the list of paths retrieved (lines 81 to 96). For each path the algorithm browses the list of nodes and, for each pair of neighboring nodes, computes increments the betweenness of the edge in the betweenness list.

3.2 Isolating clusters

A cluster is a list of linked nodes. Identifying them can be done during the computing of the betweenness.

Each time a starting node is chosen the algorithm checks if it has already been used as a target node. If it has not, a potential new cluster has been found, and a new cluster list is created (lines 43 to 47). If it has, the algorithm search the number of the cluster in which the node is (lines 34 to 41). In both cases the algorithm saves the number of the cluster in which is the starting node.

Each time a linked node is found the algorithm can determine that it is in the same cluster as the first. Before doing so the algorithm checks if the node has already been found. If it has, it is the sign that a connected cluster has been found, it is merged into the current one (lines 57 to 72). If it has not, the target node is added to the cluster being explored (lines 73 to 78).

3.3 Results

Our clustering algorithm completes in an average of 24.189s.

And for the number of clusters found, our algorithm breaks down the huge central cluster from a number of requested clusters of 147. Below, it only finds tiny clusters.

These clusters are densely connected subgraphs. This means that these clusters are zones in which one can easily navigate from one point to another. Or alternatively, this means that the edges kept in clusters are the least used, since the edges crossed by the highest numbers of shortest paths are removed.

Appendix

This program is meant to be run by invoking Python on the `gtfs.py` script and adding the path to the GTFS data as argument.

For example, you may execute `python src/gtfs.py ./data/` from the workspace root.

A `gtfs.py`

```
1 from typing import Tuple
2 from sys import argv
3 from os import getcwd
4 from os.path import join
5 from time import perf_counter
6 from math import sqrt
7 from timing import timing
8 from pathfinding import *
9 from clustering import clustering
10
11 Position = Tuple[float, float]
12 class Stop:
13     """Stop representation
14
15     # Properties
16     - 'id' - Unique identifier
17     - 'position' - Position of the stop
18     """
19     def __init__(self, id: str, lat: float, lon: float):
20         self.__id: str = id
21         self.__position: Position = (lat, lon)
22     def __repr__(self) -> str:
23         return "{0} {1}".format(self.__id, self.__position)
24     def __lt__(self, other: 'Stop') -> bool:
25         return self.id < other.id
26     def __eq__(self, other: 'Stop') -> bool:
27         return self.id == other.id and self.position == other.position
28     def __hash__(self) -> int:
29         return hash((self.__id, self.__position))
30     @property
31     def id(self) -> str:
32         return self.__id
33     @property
34     def position(self) -> Position:
35         return self.__position
36
37 def import_stops(path: str) -> Tuple[List[Stop], Dict[str, int]]:
38     """Import stops from GTFS 'stops.txt'
39
40     # Arguments
41     - 'path' - Path to the file
42
43     # Return value
```



```

44 Tuple '(stops, id_map)' where 'stops' is a list of 'Stop' instances,
45 and 'id_map' is a dictionary 'stop.id => node_id' where 'node_id' is the ←
    index of the node in 'stops'
46 """
47 stops: List[Stop] = []
48 id_map: Dict[str, int] = dict()
49 with open(path, "rt") as file:
50     for i, line in enumerate(file):
51         if i > 0:
52             data = line.strip().split(",")
53             if data[10] == "TE" and int(data[8]) < 2 and len(data[9]) == 0: # ←
                Filter stops
54                 stop = Stop(data[0], float(data[4]), float(data[5]))
55                 stops.append(stop)
56                 id_map[stop.id] = len(stops) - 1
57 return stops, id_map
58 def import_edges(path: str, id_map: Dict[str, int]) -> Set[Tuple[int, int]]:
59     """Import edges from GTFS 'stop_times.txt'
60
61     # Arguments
62     - 'path' - Path to the file
63     - 'id_map' - 'id_map' returned from 'import_stops'
64
65     # Return value
66     Set of ordered tuples of stop IDs
67     """
68     trips: Dict[str, List[Tuple[int, int]]] = dict()
69     # Import raw data
70     with open(path, "rt") as file:
71         for i, line in enumerate(file):
72             if i > 0:
73                 data = line.strip().split(",")
74                 if data[3] in id_map:
75                     if data[0] not in trips:
76                         trips[data[0]] = []
77                     heappush(trips[data[0]], (int(data[4]), id_map[data[3]]))
78     # Transform data
79     edges: Set[Tuple[int, int]] = set()
80     for trip in trips.values():
81         while len(trip) > 1:
82             edges.add((heappop(trip)[1], trip[0][1]))
83     return edges
84
85 if __name__ == "__main__":
86     # Get data path
87     """The path to the data files can be set using a script argument.
88
89     For example, if you execute Python from the workspace root, you can enter: '←
        python src/gtfs.py ./data/'.
90     Or, if you execute Python from the 'src/' directory: 'python gtfs.py ./data←
        /'.
91     """
92     DATAPATH = argv[1] if len(argv) > 1 else "../data/"
93     # Import data
94     (stops, id_map), exetime = timing(import_stops)(join(DATAPATH, "stops.txt"))
95     print("Imported {0} stops in {1}ms".format(len(stops), exetime * 1e3))

```

```

96 edges, exetime = timing(import_edges)(join(DATAPATH, "stop_times.txt"), id_map←
    )
97 print("Imported {0} edges in {1}ms".format(len(edges), exetime * 1e3))
98 # Construct graph
99 exetime = perf_counter()
100 GRAPH = Graph(stops, compute_weight=lambda u, v: sqrt(
101     (v.position[0] - u.position[0]) ** 2 + (v.position[1] - u.position[1]) ** ←
        2))
102 for start, end in edges:
103     GRAPH.add_edge(start, end)
104 print("Constructed graph in {0}ms".format((perf_counter() - exetime) * 1e3))
105 # Construct pathfinders
106 BFS = Pathfinder(GRAPH, bfs)
107 DIJKSTRA = Pathfinder(GRAPH, dijkstra)
108 # Create clustering
109 _, exetime = timing(clustering)(DIJKSTRA, set(id_map.values()), 147)
110 print("Finished clustering in {0}ms".format(exetime * 1e3))

```

B graph.py

```

1 from typing import Callable, Dict, Generic, Iterable, Iterator, List, TypeVar
2
3 T = TypeVar("T")
4 class Node(Generic[T]):
5     """Graph node representation
6
7     # Generic
8     - 'T' - Type of the node value
9
10    # Properties
11    - 'value' - Value of the node
12    - 'neighbors_out' - Dictionnary 'node => weight' of outward neighbors
13    - 'neighbors_in' - Dictionnary 'node => weight' of inward neighbors
14    """
15    def __init__(self, value: T):
16        self.__value = value
17        self._neighbors_out: Dict[int, float] = dict()
18        self._neighbors_in: Dict[int, float] = dict()
19    def __repr__(self) -> str:
20        return repr(self.__value)
21    def __lt__(self, other: 'Node') -> bool:
22        return self.value < other.value
23    def __eq__(self, other: 'Node') -> bool:
24        return self.value == other.value
25    def __hash__(self) -> int:
26        return hash(self.__value)
27    @property
28    def value(self) -> T:
29        return self.__value
30    @property
31    def neighbors_out(self) -> Dict[int, float]:
32        return self._neighbors_out
33    @property
34    def neighbors_in(self) -> Dict[int, float]:

```

```

35         return self._neighbors_in
36 class Graph(Generic[T]):
37     """Graph (weighted directed) representation
38
39     # Generic
40     - 'T' - Type of the nodes
41
42     # Properties
43     - 'nodes' - List of nodes
44     - 'size' - Size of the graph
45     - 'compute_weight' - Function to compute edge weight from two nodes
46     """
47     def __init__(self, nodes: Iterable[T], compute_weight: Callable[[T, T], float] ←
         = None):
48         self.__nodes: List[Node[T]] = []
49         for node in nodes:
50             self.add_node(node)
51         self.__size: int = 0
52         self.__compute_weight = compute_weight
53     def __iter__(self) -> Iterator[Node[T]]:
54         return iter(self.__nodes)
55     def __getitem__(self, key: int) -> Node[T]:
56         return self.__nodes[key]
57     @property
58     def order(self) -> int:
59         return len(self.__nodes)
60     @property
61     def size(self) -> int:
62         return self.__size
63     def add_node(self, node: T):
64         """Add a node to the graph
65
66         # Arguments
67         - 'node' - Node value to add
68         """
69         self.__nodes.append(Node(node))
70     def add_edge(self, start: int, end: int):
71         """Add an edge 'u-(weight)->v' to the graph
72
73         Will compute the weight using the 'compute_weight' property.
74         If 'compute_weight' is 'None', the weight will be 0.
75
76         # Arguments
77         - 'start' - Key of the first node
78         - 'end' - Key of the second node
79
80         # Errors thrown
81         - 'ValueError' if both keys are equal
82         """
83         if start == end:
84             raise ValueError("start={0} and end={0} are equal".format(start, end))
85         else:
86             weight = 0 if self.__compute_weight is None else self.__compute_weight ←
                 (self.__nodes[start].value, self.__nodes[end].value)
87             self.__nodes[start].neighbors_out[end] = weight
88             self.__nodes[end].neighbors_in[start] = weight

```

```
89 | self.__size += 1
```

C pathfinding.py

```
1 from typing import Callable, Dict, Generic, List, Set, Tuple, TypeVar
2 from math import inf
3 from heapq import heapify, heappop, heappush
4 from graph import Graph
5
6 T = TypeVar("T")
7 class Pathfinder(Generic[T]):
8     """Wrapping class to allow pathfinding in graphs
9
10     # Generic
11     - 'T' - Type of the graph nodes
12
13     # Properties
14     - 'graph' - Graph used to compute pathfinding
15     - 'previous' - Dictionary 'from => to => previous' where 'previous' is a list↵
16       of the previous nodes of 'to' when searching from 'from'
17     - 'distance' - Dictionary 'from => to => distance'
18     - 'method' - Function to compute shortest paths from a node
19     """
20     def __init__(self, graph: Graph[T], method: Callable[['Pathfinder[T]', int], ↵
21       None]):
22         self.graph = graph
23         self._previous: Dict[int, Dict[int, Set[int]]] = dict()
24         self._distance: Dict[int, Dict[int, float]] = dict()
25         self._method = method
26     def reset(self):
27         """Reset the pathfinding results"""
28         Pathfinder.__init__(self, self.graph, self._method)
29     def compute(self, start: int):
30         """Execute the pathfinding method from a certain node
31
32         Save the newly computed data to the save file
33
34         # Arguments
35         - 'start' - Key of the starting node
36         """
37         if start not in self._previous:
38             self._method(self, start)
39     def has_path(self, start: int, end: int) -> bool:
40         """Check if a path exist between two nodes
41
42         # Arguments
43         - 'start' - Key of the starting node
44         - 'end' - Key of the ending node
45
46         # Errors thrown
47         - 'ValueError' if both keys are equal
48
49         # Return value
50         'True' if a path exists; 'False' otherwise
```

```

49     """
50     if start == end:
51         raise ValueError("start={0} and end={0} are equal".format(start, end))
52     else:
53         if start not in self._previous:
54             self.compute(start)
55         return end in self._previous[start]
56 def get_paths(self, start: int, end: int) -> List[List[int]]:
57     """Get the shortest path between two nodes
58
59     # Arguments
60     - 'start' - Key of the starting node
61     - 'end' - Key of the ending node
62
63     # Errors thrown
64     - 'ValueError' if both keys are equal
65
66     # Return value
67     List of paths, with a path being a list of node indexes
68     """
69     if start == end:
70         raise ValueError("start={0} and end={0} are equal".format(start, end))
71     else:
72         if start not in self._previous:
73             self.compute(start)
74         paths: List[List[int]] = []
75         def __recurse(path: List[int], pos: int):
76             if path[pos] == start:
77                 paths.append(path[:pos + 1][::-1])
78             else:
79                 for previous in self._previous[start][path[pos]]:
80                     if len(path) < pos + 2:
81                         path.append(previous)
82                     else:
83                         path[pos + 1] = previous
84                         __recurse(path, pos + 1)
85             if self.has_path(start, end):
86                 __recurse([end], 0)
87         return paths
88 def get_distance(self, start: int, end: int) -> float:
89     """Get the distance between two nodes
90
91     # Arguments
92     - 'start' - Key of the starting node
93     - 'end' - Key of the ending node
94
95     # Return value
96     Distance from 'start' to 'end', in edges
97     """
98     if start not in self._previous:
99         self.compute(start)
100     return self._distance[start][end] if end in self._distance[start] else inf
101
102 def bfs(self: Pathfinder[T], start: int):
103     """Breadth-First Search method for 'Pathfinder'"""
104     if start not in self._previous:

```

```

105     self._previous[start] = dict()
106     self._distance[start] = dict()
107     self._distance[start][start] = 0
108     queue = [start]
109     heapify(queue)
110     while len(queue) > 0:
111         current = heappop(queue)
112         for u in self.graph[current].neighbors_out:
113             if u not in self._distance[start]:
114                 self._previous[start][u] = set()
115                 self._distance[start][u] = self._distance[start][current] + 1
116                 heappush(queue, u)
117             if self._distance[start][u] == self._distance[start][current] + 1:
118                 self._previous[start][u].add(current)
119 def dijkstra(self: Pathfinder[T], start: int):
120     """Dijkstra method for 'Pathfinder'"""
121     if start not in self._previous:
122         self._previous[start] = dict()
123         self._distance[start] = dict()
124         self._distance[start][start] = 0
125         marked: Set[int] = set()
126         queue: List[Tuple[float, int]] = [(0, start)]
127         heapify(queue)
128         while len(queue) > 0:
129             _, current = heappop(queue)
130             marked.add(current)
131             for (u, weight) in self.graph[current].neighbors_out.items():
132                 tentative_distance = self._distance[start][current] + weight
133                 if u not in self._distance[start] or tentative_distance < self._distance[start][u]:
134                     self._previous[start][u] = set()
135                     self._distance[start][u] = tentative_distance
136                 if self._distance[start][u] == tentative_distance:
137                     self._previous[start][u].add(current)
138                 if u not in marked:
139                     heappush(queue, (self._distance[start][u], u))

```

D clustering.py

```

1 from pathfinding import Pathfinder
2
3 def clustering(DIJKSTRA: Pathfinder, nodes, n, display_all=False):
4     """
5     Clustering method (first try)
6     """
7     print("\nCreating", n, "clusters...")
8     clusters = [] # Content of the clusters
9     n_nodes = len(nodes) # Number of nodes
10    iteration = 0
11    while len(clusters) < n: # As long as the right number of clusters has not ←
12        been constituted
13        iteration += 1
14        print("\nIteration", iteration)
15        edge_betweenness = {} # Betweenness of each edge

```

```

15 DIJKSTRA.reset()
16 nodes_found = set() # List of discovered nodes
17 nodes_to_explore = nodes.copy() # List of nodes to explore
18 progress = 0
19 clusters = []
20 current_cluster = -1
21 while nodes_to_explore: # As long as there is still unexplored nodes
22     # Displaying the progress
23     new_progress = int(round(100 * (1 - len(nodes_to_explore) / n_nodes), ←
24                             -1))
25     if new_progress > progress:
26         progress = new_progress
27         print(progress, "%", sep='')
28     starting_node = nodes_to_explore.pop() # Take a node to explore
29     # Update the list of clusters
30     if starting_node in nodes_found: # If the node has already been ←
31         discovered...
32         # Search the node in every cluster list
33         for i in range(len(clusters)):
34             if starting_node in clusters[i]:
35                 current_cluster = i
36     else: # If the node hasn't been discovered report the discovery of a ←
37         new cluster
38         clusters.append({starting_node}) # Create a new entry for the new ←
39         cluster containing its first node
40         current_cluster = len(clusters) - 1 # Set the new cluster as the ←
41         current one
42     nodes_found.add(starting_node) # Add the node to the discovered nodes ←
43     list
44     for target_node in nodes_to_explore: # Search every other node
45         # (paths), exetime = timing(DIJKSTRA.get_paths)(starting_node, ←
46         target_node)
47         # print("Dijkstra of {0} in {1}ms".format(starting_node, exetime * ←
48         1e3))
49         paths = DIJKSTRA.get_paths(starting_node, target_node) # Search a ←
50         path between the two nodes
51         if len(paths) > 0: # If there is a path...
52             # Generating list of contents of each cluster
53             if target_node in nodes_found: # If the node has already been ←
54                 found
55                 # Search the node in the lists of the others clusters
56                 for other_cluster in range(len(clusters)):
57                     if current_cluster != other_cluster:
58                         if target_node in clusters[other_cluster]:
59                             # Add the content of the other cluster to the ←
60                             current cluster
61                             clusters[current_cluster].update(clusters[←
62                                 other_cluster])
63                             del clusters[other_cluster] # Delete the ←
64                             other cluster
65                             # If the deleted cluster was older take into ←
66                             account the shift of keys in the list
67                             if other_cluster < current_cluster:
68                                 current_cluster -= 1
69                             break
70             else: # If the node hasn't been found...

```

```

57         # Set the second node as discovered at add it to the same ←
           cluster as the first one
58         nodes_found.add(target_node)
59         clusters[current_cluster].add(target_node)
60         n_paths = len(paths) # Number of paths of equal length found
61         for path in paths: # For each path...
62             last_node = path.pop() # Get the last node of the path ←
           between the two nodes
63         while path: # For each edge of the path...
64             previous_node = path.pop() # Get the previous node
65             edge_name = (previous_node, last_node) # Compute the ←
           name of the edge between them
66             # Compute the new edge betweenness
67             if edge_name in edge_betweenness:
68                 edge_betweenness[edge_name] += 1 / n_paths
69             else:
70                 edge_betweenness[edge_name] = 1 / n_paths
71             last_node = previous_node
72         n_clusters = len(clusters) # Count the number of clusters
73         print(n_clusters, "clusters found")
74         # Delete as many edges as there are clusters to create
75         for i in range(n - n_clusters):
76             # Get the edge with the highest betweenness
77             highest_betweenness = max(edge_betweenness, key=edge_betweenness.get)
78             print("Deleting the edge between ", highest_betweenness[0], " and ", ←
           highest_betweenness[1],
79                   " (Betweenness: ", edge_betweenness[highest_betweenness], ")", ←
           sep='')
80             del edge_betweenness[highest_betweenness] # Delete it from the list ←
           of edges betweenness
81             # Deleting the edge by deleting its name from the neighbors_out list ←
           of the start node
82             del DIJKSTRA.graph[highest_betweenness[0]].neighbors_out[←
           highest_betweenness[1]]
83             """
84             Doing so should delete de facto the edge from the graph.
85             A ghost edge will still be listed in neighbors_in but it shouldn't be ←
           used by the program
86             """
87         # Displaying the clusters created
88         for i in range(len(clusters)):
89             size = len(clusters[i])
90             if size > 1 or display_all:
91                 print("Cluster ", i + 1, ": size: ", size, " (", round(100 * len(←
           clusters[i]) / n_nodes), "%), content: ",
92                       sorted(clusters[i]), sep='')

```