

Electric Cars: Logistics For the Environment

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Data Structures













	0	1	2
0	-	 	 
1	 	-	 
2	 	 	-

Figure 1: Adjacency Matrix with weights of pairs Distance - Time. Indexes represent the ID of the respective nodes. The diagonal is empty due both IDs refer to the same node.


0	1	2	3
			

Figure 2: Array that contains the data of each node [Name of the ubication, x coordinate, y coordinate, type of node (either deposit, station, or client) and if is a station, the type of station (Fast, normal, or slow charging)]. Index represents the ID of the node.

Variables' dictionary

<i>Variable</i>	<i>Description</i>
N	Number of customers in the map. <i>Equals to $W \cdot R$</i>
S	Number of clients that have not been visited.
W	The maximum number of customers that a vehicle can visit.
R	Number of routes.

Table 1: Dictionary of variables used to express asymptotic complexity.

Algorithm and Complexity

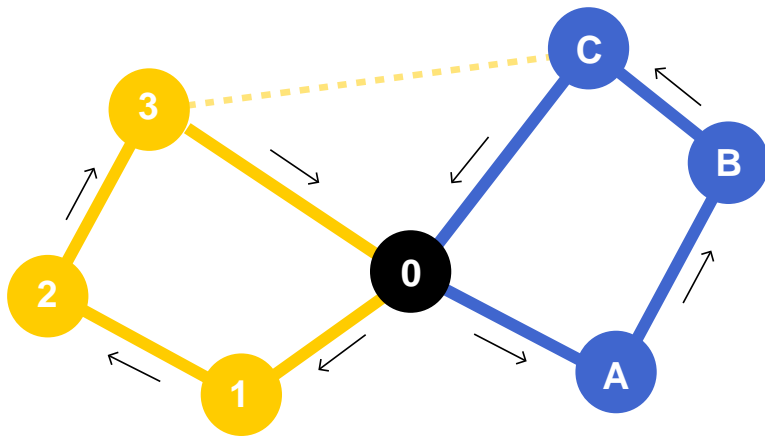


Figure 3: Procedure of the algorithm to calculate routes.

Steps	Asymptotic complexity
Find the best client	$O(S)$
Found a route	$O(S*W)$
Found all the routes	$O(S*W*R)$
Final algorithm	$O(S*N)$

Algorithm design criteria

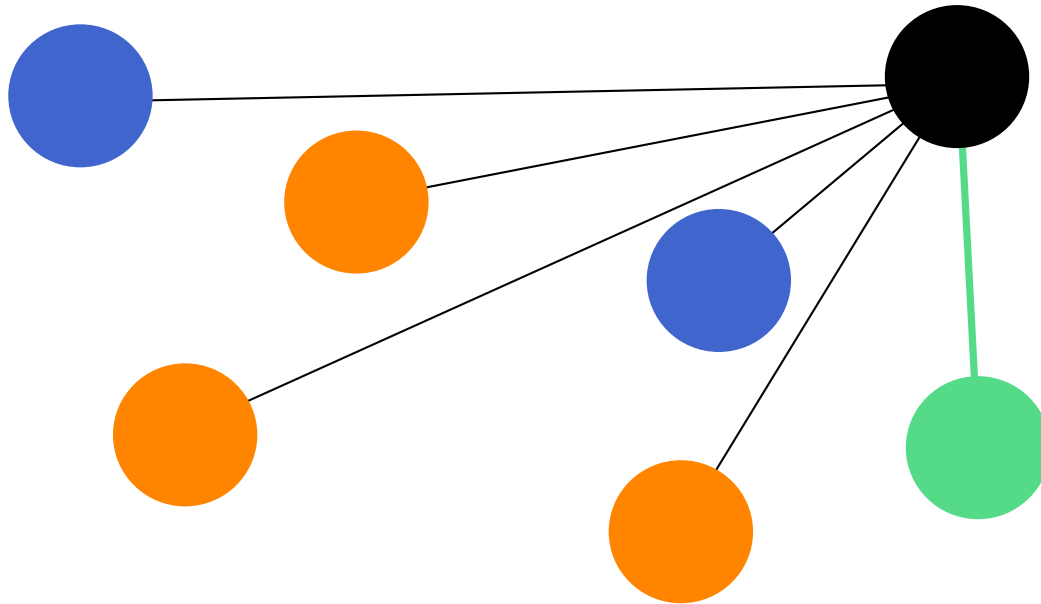


Figure 4:
*Procedure of the
algorithm to
calculate routes.*

After researching of all the different approaches that can be done, we decided to focus in a greedy one, trying constantly to find a random nearest client by the time of calculating a route. This nearest client is selected randomly in a k selected range. This has several optimization benefits since it groups the clients by zones, maximizing the time of the routes and ending with a less use of vehicles thanks to the random search. Besides its good complexity, which is $O(S*W)$ to find each of the different routes, the algorithm also has shown a great performance in the given datasets.

Time and Memory Consumption

Dataset	Best case (sec)	Average case (sec)	Worst case (sec)
1	3.81128	3.87553	3.92024
2	3.09814	3.10244	3.12267
3	2.87452	2.92764	2.993
4	3.68958	3.79476	4.066
5	2.25342	2.3123	2.41406
6	2.67286	2.74125	2.77312
7	3.4899	3.65148	4.17748
8	2.15364	2.23134	2.35968
9	2.59232	2.7165	2.7948
10	3.44148	3.52738	3.69084
11	2.29312	2.37526	2.44696
12	2.50348	2.55964	2.63098

Table 2: Execution time of the algorithm for different datasets.

Dataset	Memory consumption (MB)
1	79.63
2	82.10
3	76.23
4	78.09
5	73.91
6	75.68
7	80.32
8	72.82
9	74.98
10	77.75
11	74.21
12	75.96

Table 3: Memory consumption of the algorithm for different datasets.

Software prototype

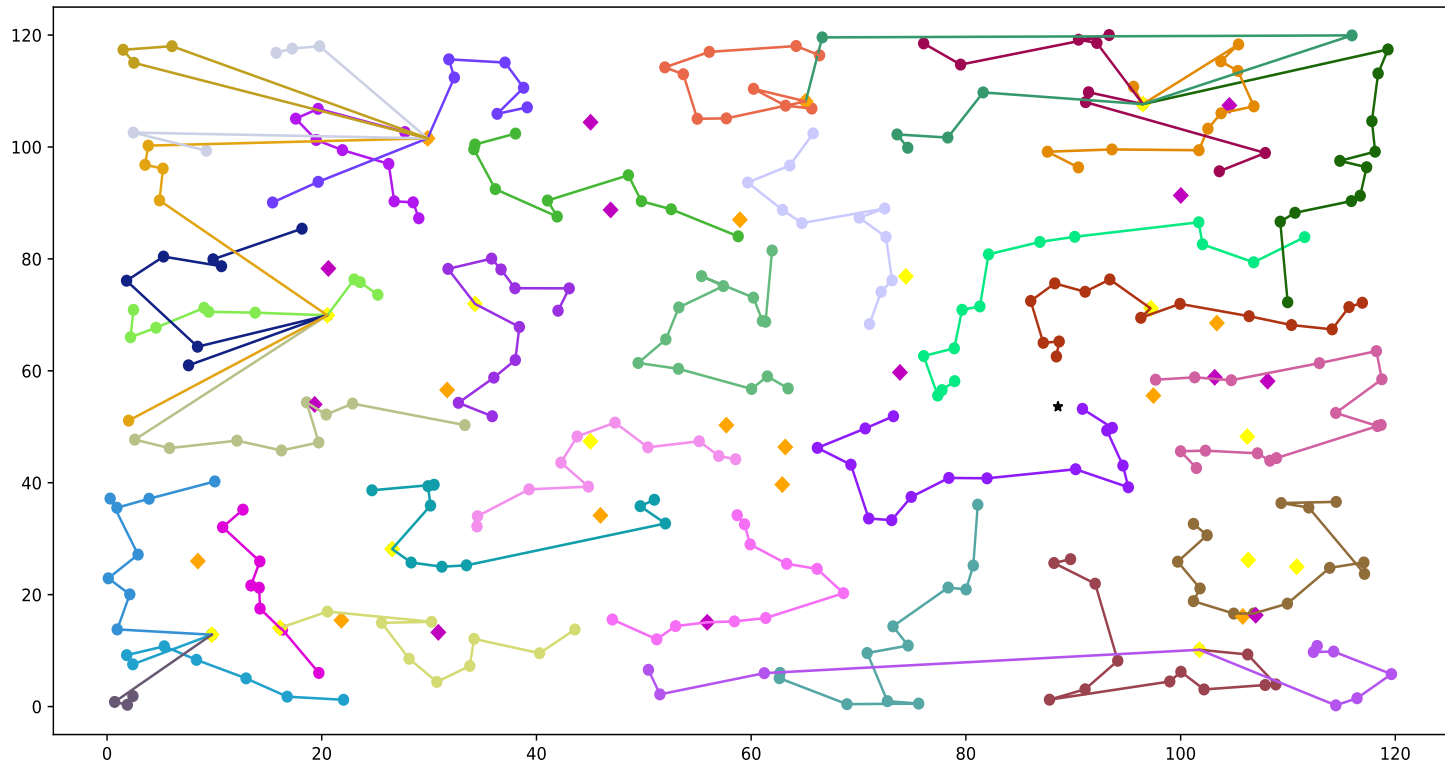


Figure 6: Illustration designed with matplotlib. It represents each of the routes calculated for one of the datasets. Keep in mind that each color represents a route, and every route starts and ends in the black star, which represents the deposit.

Software prototype

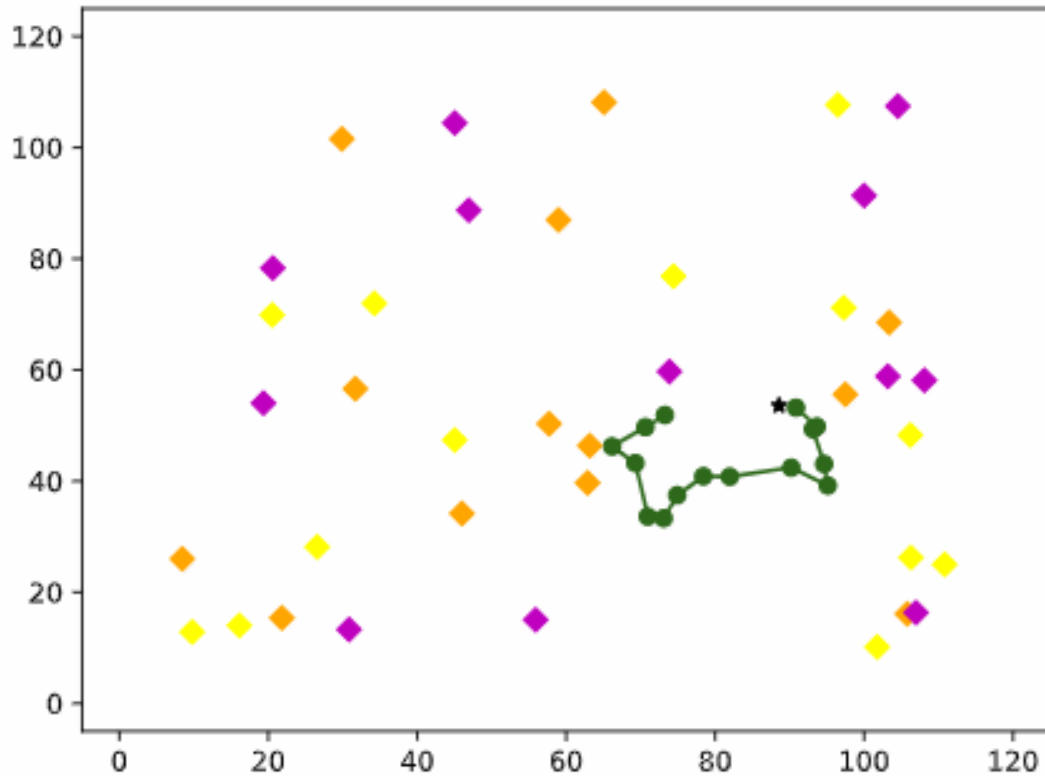


Figure 6: Illustration designed with matplotlib. It represents each of the routes calculated for one of the datasets. Keep in mind that each color represents a route, and every route starts and ends in the black star, which represents the deposit.