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Heuristics & Approximations Algorithms

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Contents

1	Introduction	2
2	Nearest Neighbour	3
3	First Custom Local Search Algorithms	4
4	Performance Analysis & Boxplots	4
5	Process Analysis	4
	Appendices	5
A		
	Custom Cluster Algorithm to solve CVRP	5
B		
	LocalSearch Algorithm - Swap if any improvement	6
C		
	LocalSearch Algorithm - Move if any improvement	7
D		
	BOXPLOT FIGURES	7

1 Introduction

The goal of this project is to show the knowledge acquired by the student in regards to Heuristic- and Local Search Algorithms. The algorithms are developed to suit the Capacitated Vehicle Routing Problem (CVRP). CVRP is a combinatorial optimization problem, where a set of n customers has to be visited by m vehicles. Each customer can only be visited one vehicle. The sum of demands by the customers assigned to a vehicle must not exceed the capacity of the vehicle. All vehicles start and end their route at the depot (the starting point). The objective is to minimise the travelling cost and try to use as few vehicles as possible. Heuristics and Local Search algorithms are applied to CVRP, where the heuristics are laying the foundation of possible solutions of routes, and the Local Search algorithms try to improve the solutions. Three heuristic algorithms have been developed:

- Nearest Neighbour - Pick nearest customer from the depot or current customer without breaking the capacity
- Furthest Neighbour (with nearest clustering) - Pick customer furthest away from the depot, find the closest customers until the capacity is reached, and generate a route by using Nearest Neighbour
- K-Nearest Neighbour - Pick k nearest customers and randomly pick one of the k options. Proceed until no more customers without breaking the capacity

Two local search algorithms have been developed:

- Two Opt - Pick a route generated by some heuristic. Pick two edges, swap the edges, and compare the distance from before and after the swap. If better, make the swap.
- Three Opt - Pick a route generated by some heuristic. Pick three edges, swap the edges, and compare the distance from before and after the swap. If better, make the swap.

At the end of the report, comparison of the heuristics and results are presented.

2 Nearest Neighbour

The Nearest Neighbour algorithm is a simple, yet powerful, approach to tackle CVRP. The algorithm goes starts at the depot and picks the nearest customer by iterating through all of the customers. It marks the customer and visited, adds it to a vehicle, and continues until the capacity of the vehicle is reached (or nearly reached). It returns to the depot and start over with the remaining customers.

The algorithm is described in pseudocode:

Algorithm 1 Custom CVRP Heuristic - Nearest Neighbour Approach

Require: $Point_1 \dots Point_N$

Ensure: *Solution* (solution to the CVRP instance)

```

1: function ALGORITHM( $Points[]$ )
2:    $capacity \leftarrow 0$ 
3:    $data \leftarrow$  CVRP instance
4:    $visited \leftarrow$  empty array containing visited nodes
5:    $shortestdistance \leftarrow 0$ 
6:    $current \leftarrow$  keep track of next node to visit
7:   for  $i \leftarrow 0$  to length of data-1 do
8:     for  $j \leftarrow 1$  to length of data do
9:        $temp \leftarrow$  euclidean distance( $current, j$ )
10:      if  $j$  not in  $visited$  then
11:        if  $temp \leq shortestdistance$  then
12:           $shortestdistance \leftarrow temp$ 
13:           $current \leftarrow j$ 
14:           $visited \leftarrow j$ 
15:       $\triangleright$  checking if capacity requirements are met
16:      if current total capacity  $\leq$  max capacity then
17:         $capacity \leftarrow$  capacity of current node
18:      else
19:        Add starting point to end of route and
20:        add capacity of current node to variable( $capacity$ )
21:       $shortestdistance \leftarrow 0$   $\triangleright$  reset shortestdistance
22:   return solution

```

3 First Custom Local Search Algorithms

Having executed the heuristics, providing us with solutions to the initial problem instances, we can continue optimizing. This is where the Local Search Optimization algorithms comes into play. Given a canonical solution, we want to optimize it to reduce the total cost of the solution. We have provided three custom Local Search (LS) algorithms, trying to optimize the canonical solution. All of our localsearches optimize solution by make exchanges of visiting node between 2 routes in solution set. Thus they can be used in combination with original Traveling salesman problem's localsearches such as 2-opt and 3-opt, where improvements are done by swapping and reversing order of visiting nodes in single route.

The first LS algorithm looks into the generated routes done by the heuristics. It takes two routes and compare them with each other, checking if any swaps are possible in the two routes. A swap is possible if the distance is being reduced in at least one route, while still preserving the maximum capacity limit of both routes. The second LS algorithm looks like the first but with a minor change. It takes one route and compares it with every other route to find a better solution. It then continues doing this with every other route.

4 Performance Analysis & Boxplots

5 Process Analysis

Appendices

A

Custom Cluster Algorithm to solve CVRP

Algorithm 2 Cluster CVRP Localsearch

Require: *closest_point*(*A, points*): returns point in points which has minimum distance from A

Require: *elim_empty_routes*: returns list of routes where the empty routes are removed

Ensure: *routes*: Better Solution to the CVRP instance

```

1: function ALGORITHM(routes, maxcap)
2:   route_convexes is an array storing convex_hull for each route
3:   improvements  $\leftarrow$  True
4:   while improvements do
5:     for all i, j such that i, j are routes index combination do
6:       central_point_i  $\leftarrow$  average of point in convex i
7:       central_point_j  $\leftarrow$  average of point in convex j
8:       closest_from_i_j  $\leftarrow$  closest_point(mid_point_i, convex_routes[j])
9:       closest_from_j_i  $\leftarrow$  closest_point(mid_point_j, convex_routes[i])
10:      if closest_from_i_j and closest_from_j_i is Found then
11:        if closest_from_i_j fits in route j then
12:          add move to route improvements j
13:          make the swap
14:        else if closest_from_j_i fits in route i then
15:          add move to route improvements i
16:          make the swap
17:        else if capacity is not reached by swapping customers then
18:          swap the route-points between i and j
19:        if improvements then
20:          recalculate convex hull of route i and j
21:      if not improvements then
22:        routes = elim_empty_routes(routes) return routes

```

B**LocalSearch Algorithm - Swap if any improvement**

Algorithm 3 swap_if_improvement

Require: $dist(A, B)$ returns distance from point A to B

```

1: function SWAP_IF_IMPROVEMENT( $route\_i, route\_j, j\_to\_i, i\_to\_j$ )
2:    $A, B, C \leftarrow route\_i[i\_to\_j-1], route\_i[i\_to\_j],$ 
       $route\_i[i\_to\_j+1 \bmod(\% \ len(route\_i))]$ 
3:    $D, E, F \leftarrow route\_j[j\_to\_i-1], route\_j[j\_to\_i],$ 
       $route\_j[j\_to\_i+1 \bmod(\% \ len(route\_j))]$ 
4:    $old\_distance\_i \leftarrow dist(A, B) + dist(B, C)$ 
5:    $old\_distance\_j \leftarrow dist(D, E) + dist(E, F)$ 
6:    $new\_distance\_i \leftarrow dist(A, E) + dist(E, C)$ 
7:    $new\_distance\_j \leftarrow dist(D, B) + dist(B, F)$ 
8:   if  $old\_distance\_i + old\_distance\_j >$ 
       $new\_distance\_i + new\_distance\_j$  then
      swap point B of route i with point E of route j
      return True

```

C

LocalSearch Algorithm - Move if any improvement

Algorithm 4 move_if_improvement

Require: $\text{dist}(A, B)$ returns distance from point A to B

```

1: function MOVE_IF_IMPROVEMENT(route_i, route_j, i_to_j, j_to_i)
2:   A, B, C  $\leftarrow$  route_i[i_to_j - 1], route_i[i_to_j],
      route_i[i_to_j + 1 % len(route_i)]
3:   D, E, F  $\leftarrow$  route_j[j_to_i - 1], route_j[j_to_i],
      route_j[j_to_i + 1 % len(route_j)]
4:   old_distance_i  $\leftarrow$   $\text{dist}(A, B) + \text{dist}(B, C)$ 
5:   old_distance_j  $\leftarrow$   $\text{dist}(D, E) + \text{dist}(E, F)$ 
6:   new_dist_before_B  $\leftarrow$   $\text{dist}(A, E) + \text{dist}(E, B) + \text{dist}(B, C)$ 
7:   new_dist_after_B  $\leftarrow$   $\text{dist}(A, B) + \text{dist}(B, E) + \text{dist}(E, C)$ 
8:   new_distance_j  $\leftarrow$   $\text{dist}(D, F)$ 
9:   if new_dist_before_B < new_dist_after_B then
10:    if old_distance_i + old_distance_j >
      new_dist_before_B + new_distance_j then
      move point E from route j to route i before B
11:  else
12:    if old_distance_i + old_distance_j >
      new_dist_after_B + new_distance_j then
      move point E from route j to route i after B

```

D

BOXPLOT FIGURES

