

# Evolutionary Computation - Assignment 1

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## 1 Description of the problem

We are given a graph with  $n$  nodes, each of which is described by its  $x$  and  $y$  coordinates and a node cost. The goal is to select exactly  $\text{ceil}(n/2)$  nodes that form a Hamiltonian cycle (closed path) and minimize

$$\sum_{e \in E} \text{cost}(e) + \sum_{v \in V} \text{cost}(v)$$

Where  $E$  is a set of selected edges,  $\text{cost}(e)$  is Euclidean distance between two nodes rounded mathematically to integer value,  $V$  is set of selected nodes and  $\text{cost}(v)$  is node cost.

## 2 Pseudocode

### 2.1 Random Solution

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**Algorithm 1** Algorithm that randomly selects  $\text{ceil}(n/2)$  nodes assuming that every pair of nodes is connected by an arc

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**Input:** number of nodes  $n$

**Output:** list of  $\text{ceil}(n/2)$  random numbers

```
1: cycle  $\leftarrow []$ 
2: while  $\text{len}(\text{cycle}) < \text{ceil}(n/2)$  do
3:   candidate  $\leftarrow$  random integer in range  $[0, n]$ 
4:   if candidate is not in cycle then
5:     add candidate to the cycle
6:   end if
7: end while
8: return cycle
```

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### 2.2 Nearest Neighbour

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**Algorithm 2** Algorithm that starts from given node and iteratively adds edge and node with lowest cost

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**Input:**  $D$ : Distance matrix;  $C$ : List of costs;  $n$ : Starting node;

**Output:** *Cycle*: List of greedily selected nodes

```
1: cycle  $\leftarrow [n]$ 
2: for  $i = 1$  to  $\text{ceil}(n/2)$  do
3:   node  $\leftarrow \arg \min_{n \notin \text{cycle}} D[\text{cycle}[i], n] + C[n]$ 
4:   append node to the cycle
5: end for
6: return cycle
```

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## 2.3 Greedy Cycle

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**Algorithm 3** Algorithm that starts from given node and iteratively adds edge and node with lowest cost always creating a cycle

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**Input:**  $D$ : Distance matrix;  $C$ : List of costs;  $n$ : Starting node;

**Output:**  $Cycle$ : List of greedily selected nodes

```

1:  $cheapest \leftarrow \arg \min_{i \neq n} D[n, i] + C[i]$ 
2:  $cycle \leftarrow [n, cheapest]$ 
3: for  $i = 1$  to  $\text{ceil}(n/2)$  do
4:    $position, vertex \leftarrow \arg \min_{j=0,1,\dots,i,s \notin cycle} D[cycle[j], s] + D[s, cycle[j+1]] -$ 
     $D[cycle[j], cycle[j+1]] + C[s]$ 
5:   insert  $vertex$  at position  $j+1$  to the  $cycle$ 
6: end for
7: return  $cycle$ 
```

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## 3 Results

Problem	random			nn			
	Best	Worst	Average	Best	Worst	Average	
TSPA	238331.0	291490.0	265228.75	84471.0	90513.0	87679.135	
TSPB	283627.0	292368.0	266623.445	77448.0	82631.0	79282.58	
TSPC	193432.0	238238.0	215074.295	56304.0	63697.0	58872.68	
TSPD	192364.0	241940.0	218404.675	50335.0	59846.0	54290.68	
Problem	greedy						
	Best	Worst	Average	Best	Worst	Average	
TSPA	75666.0	80321.0	77064.415				
TSPB	68743.0	76324.0	70735.35				
TSPC	53226.0	58876.0	55842.07				
TSPD	50409.0	60964.0	54883.01				

Table 1: Comparison of TSP algorithms

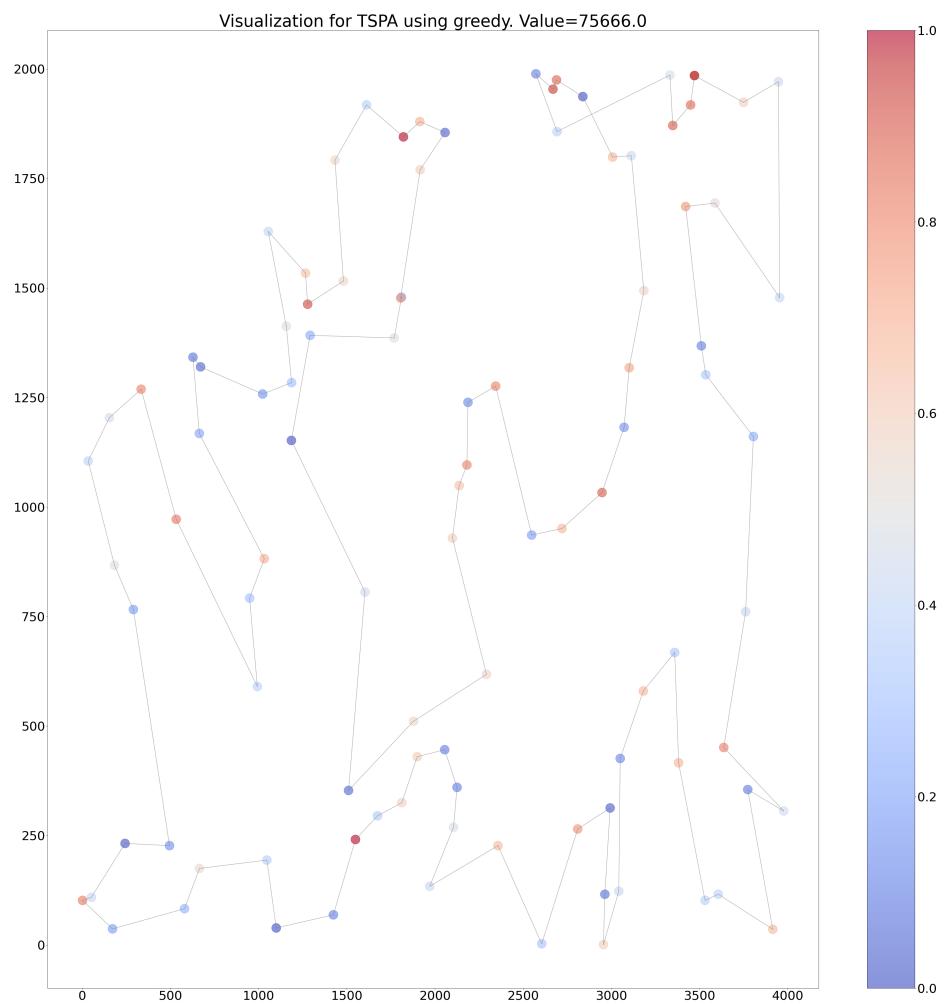


Figure 1: TSPA: Greedy

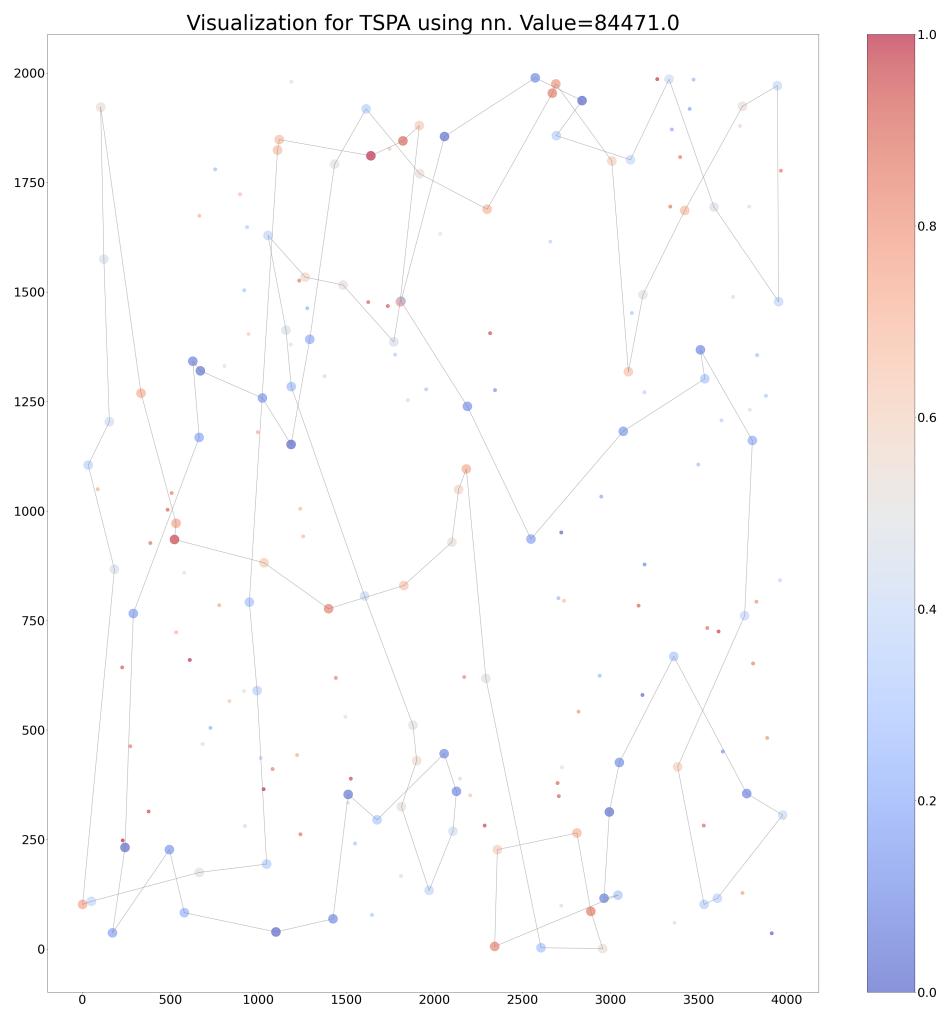


Figure 2: TSPA: Nearest Neighbor

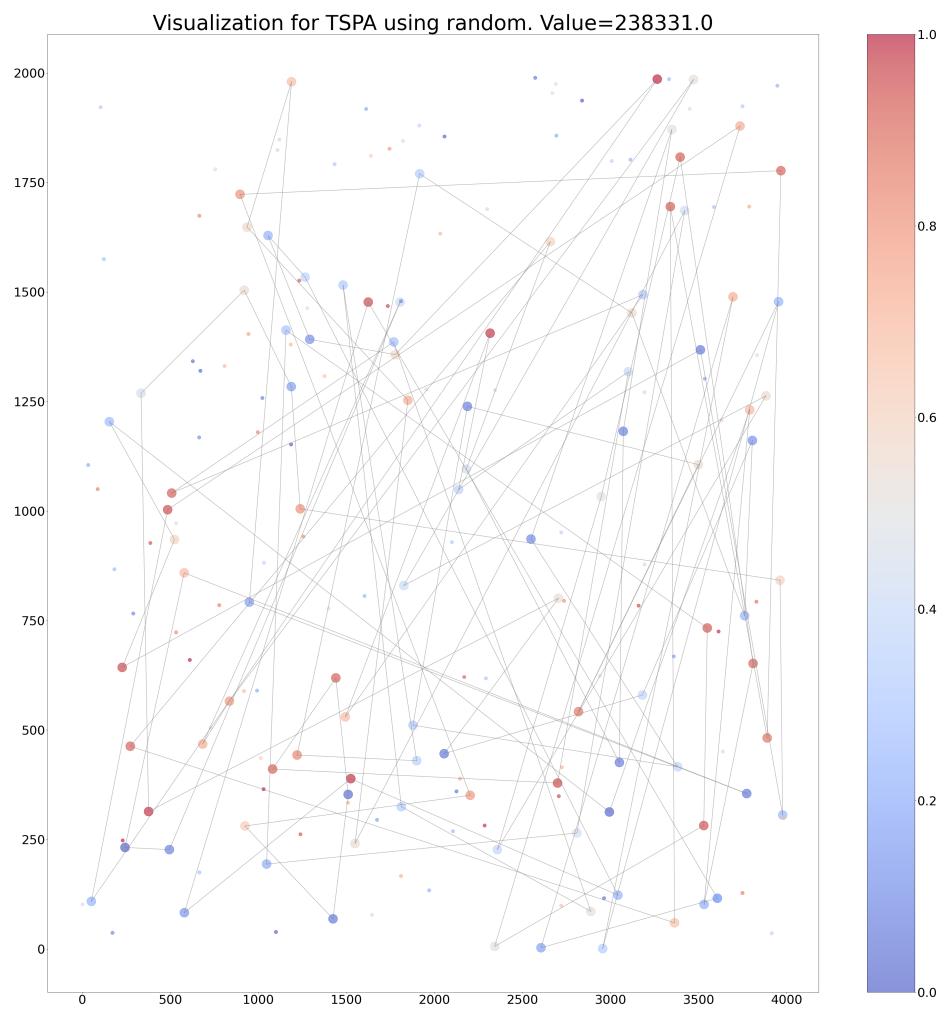


Figure 3: TSPA: Random

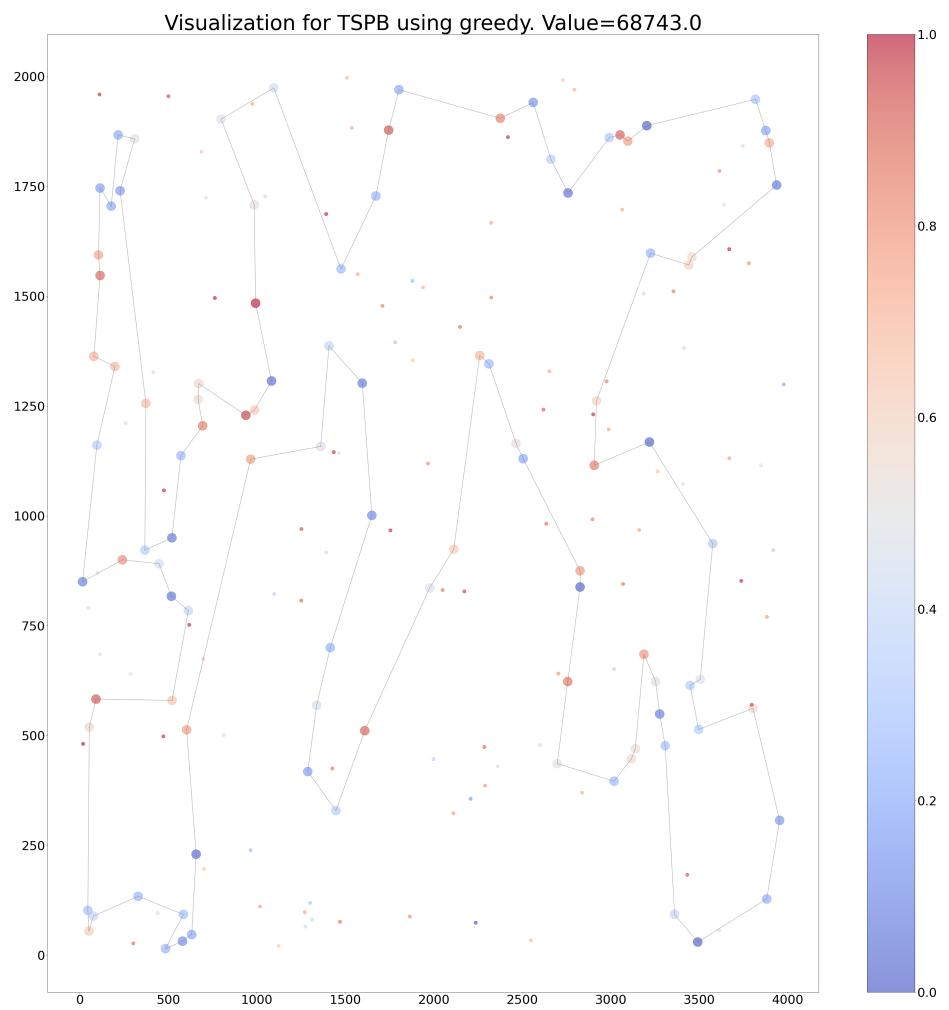


Figure 4: TSPB: Greedy

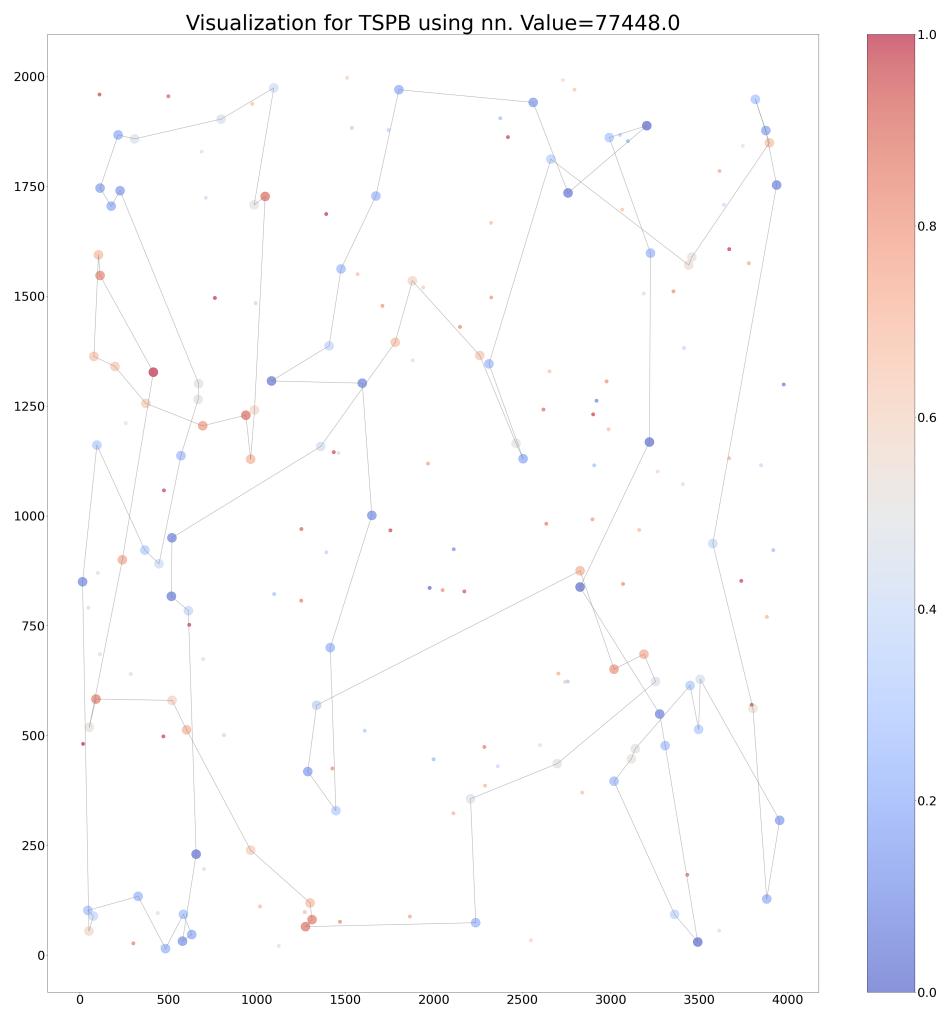


Figure 5: TSPB: Nearest Neighbor

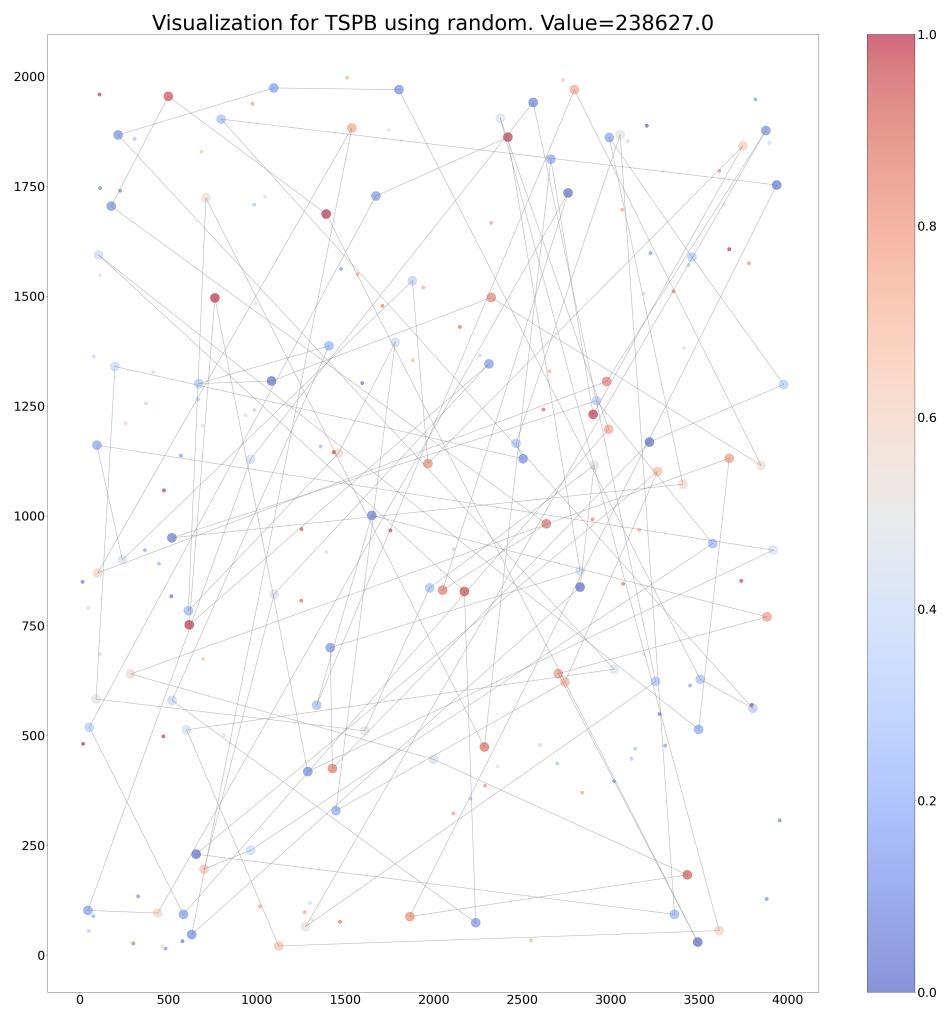


Figure 6: TSPB: Random

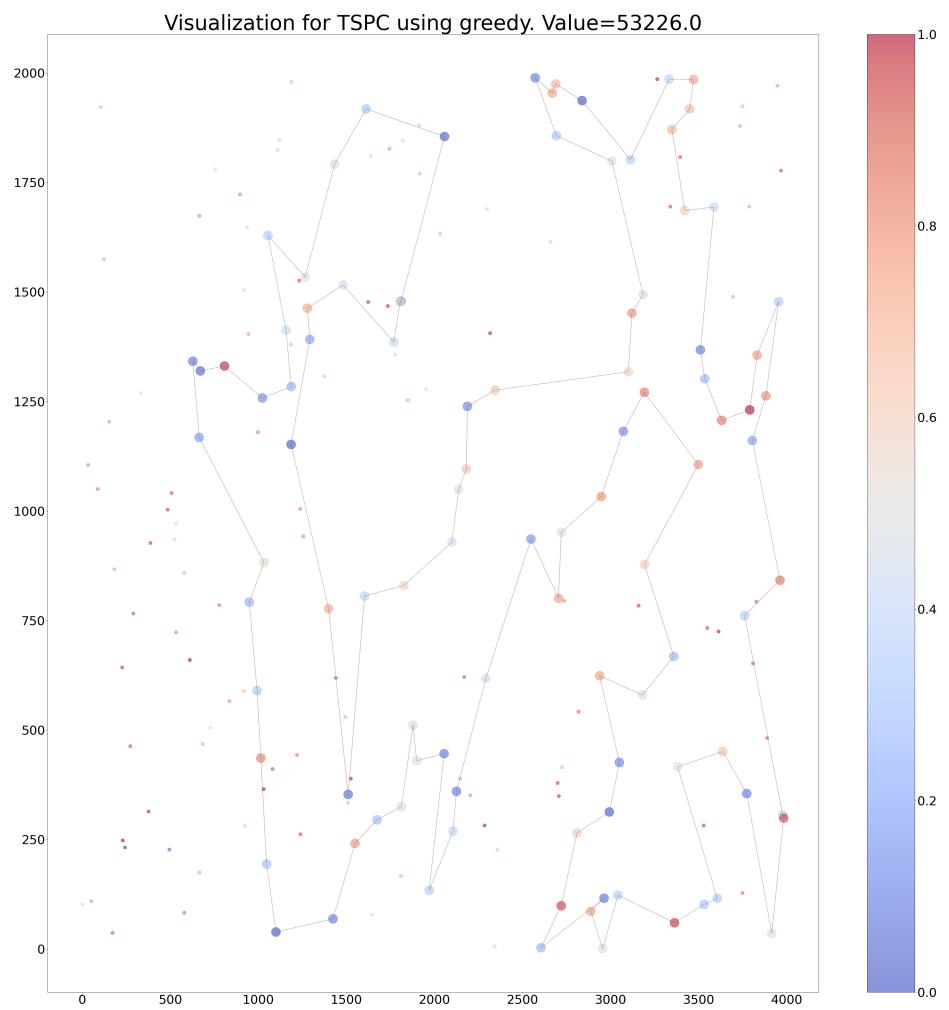


Figure 7: TSPC: Greedy

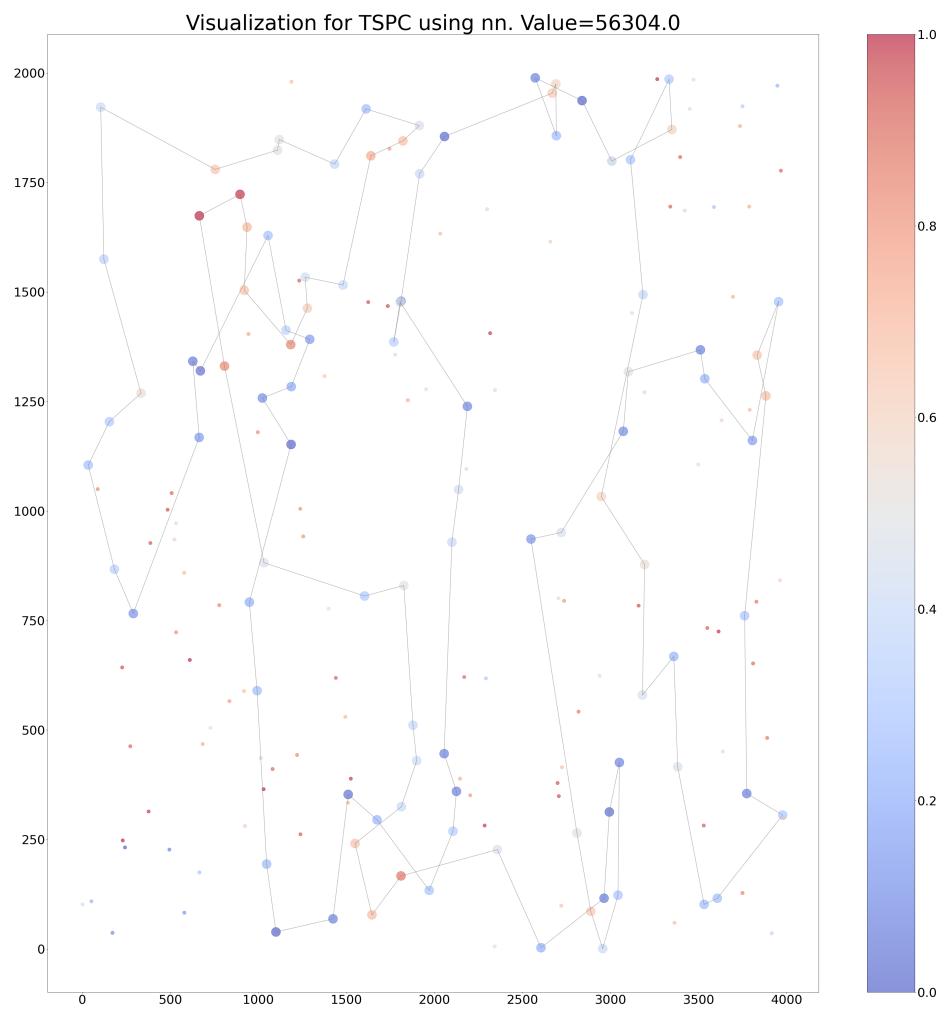


Figure 8: TSPC: Nearest Neighbor

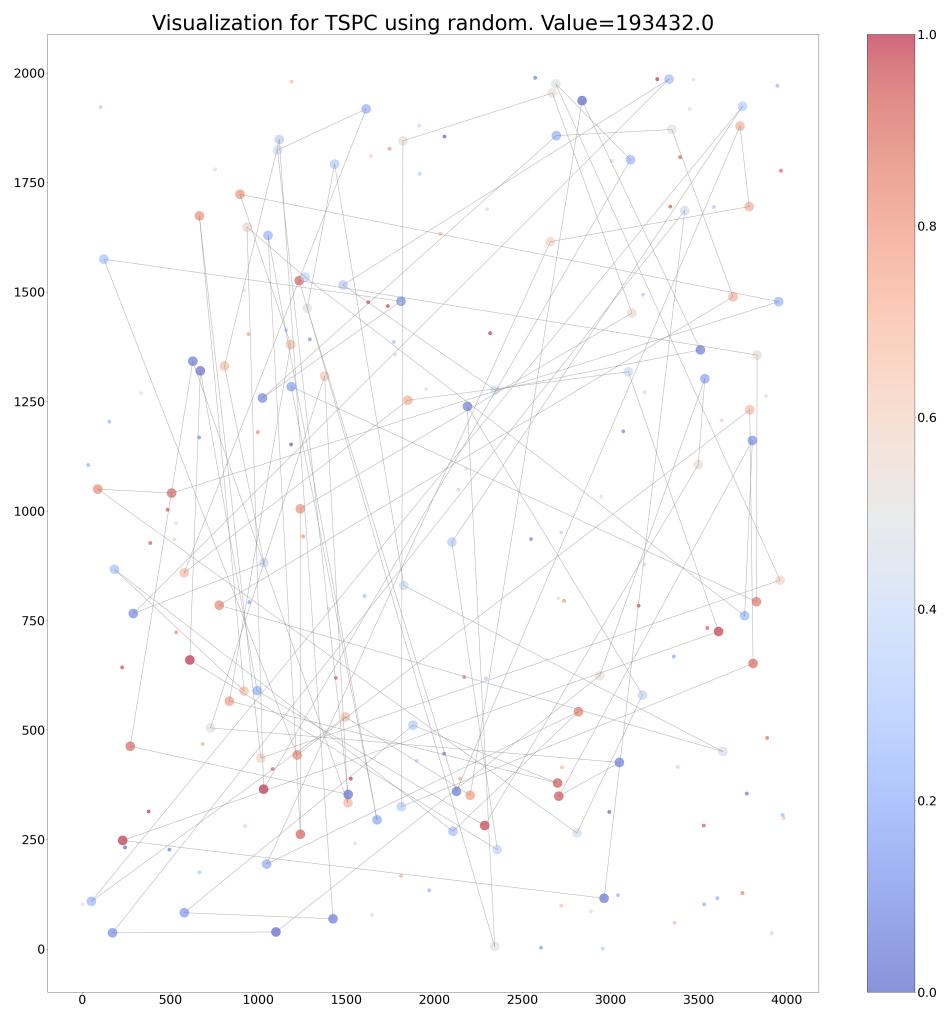


Figure 9: TSPC: Random

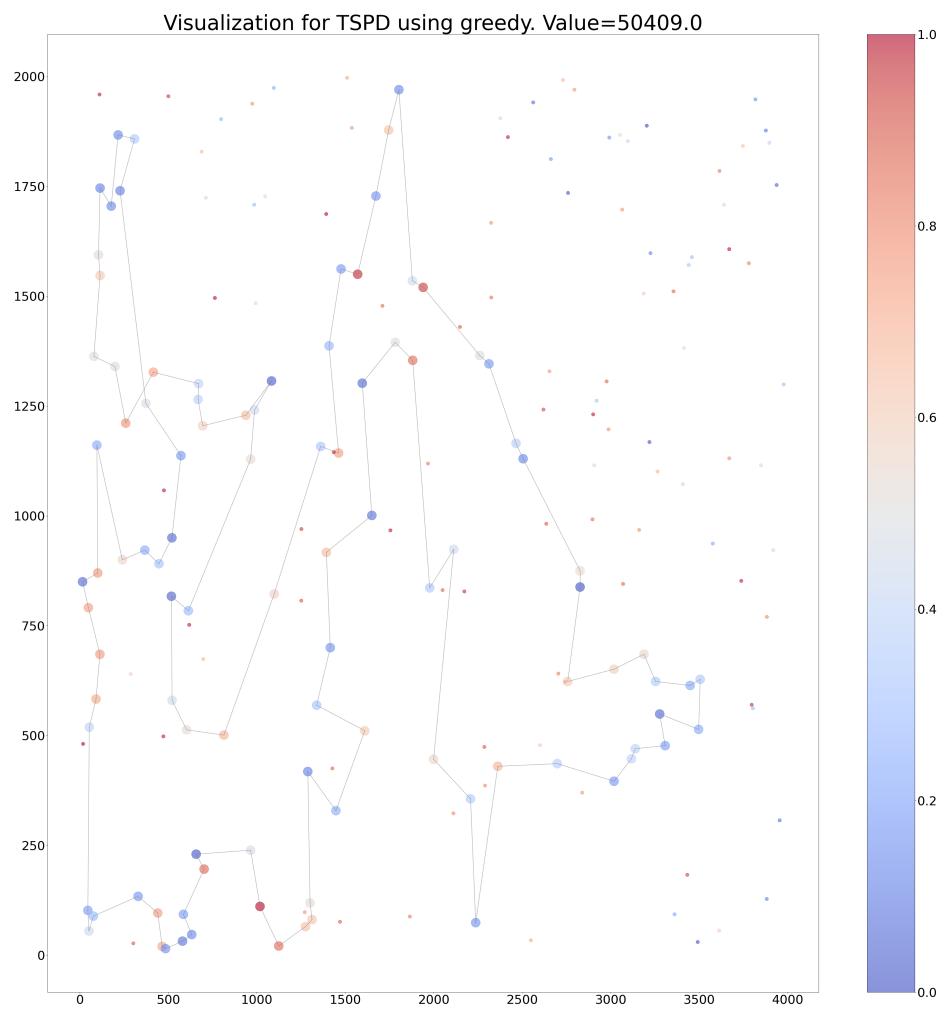


Figure 10: TSPD: Greedy

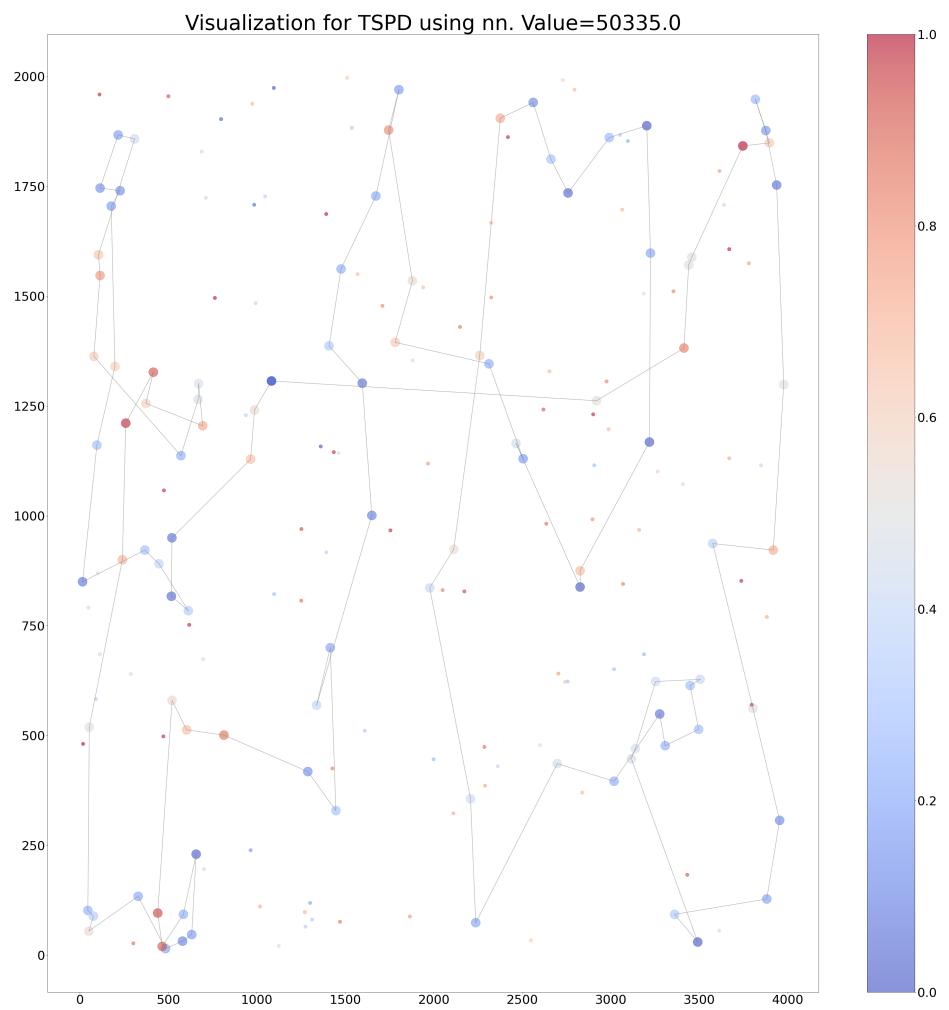


Figure 11: TSPD: Nearest Neighbor

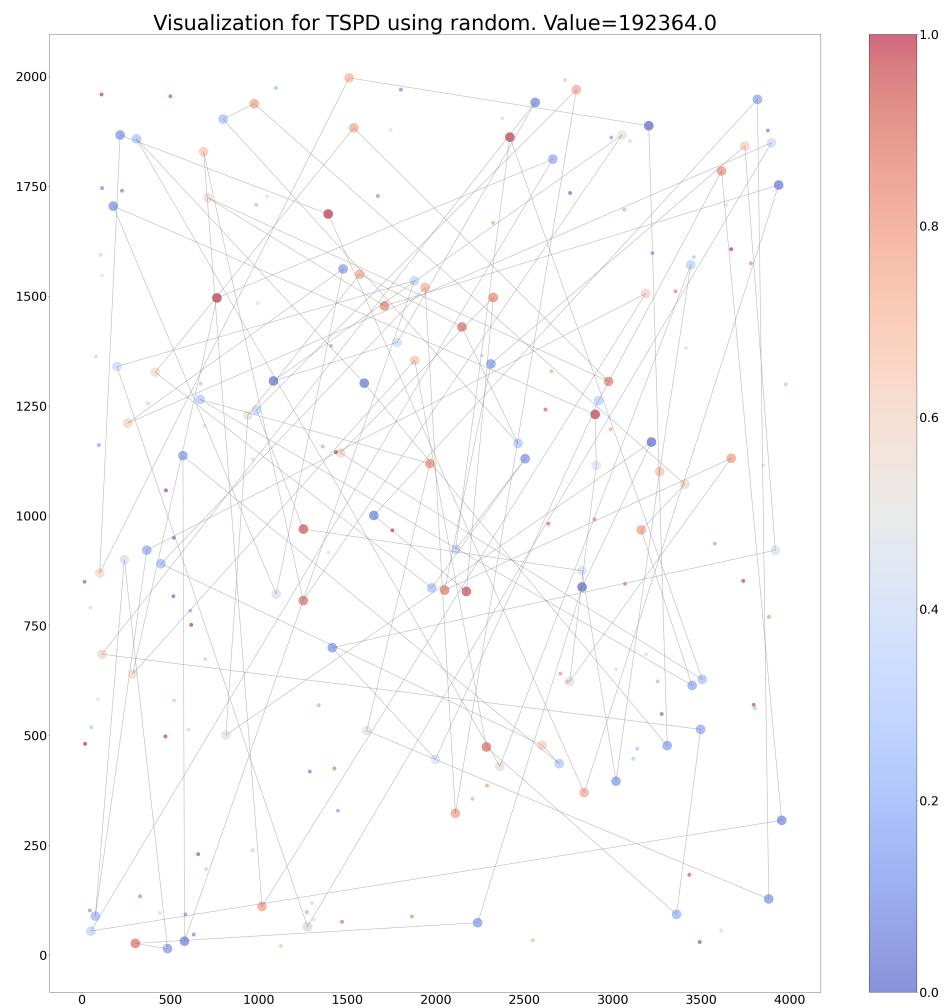


Figure 12: TSPD: Random

## **4 Code**

Implementation of algorithms and visualizations is available here

## **5 Conclusions**

In every problem, the greedy cycle method has the best results. On the visualizations we can see that paths created by the greedy cycle look mostly reasonable, the solution could be improved in several small areas, but generally, it is the best among others. The nearest neighbor method has slightly worse results, but the solutions often look strange when we look at them as humans. As expected, the random method ha the worst results with no intuitive paths and solutions far from being optimal.