Evolutionary Computation Lab II

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1 Problem description

We are given three columns of integers with a row for each node. The first two columns contain x and y coordinates of the node positions in a plane. The third column contains node costs.

- 1. Select exactly 50% of the nodes (if the number of nodes is odd we round the number of nodes to be selected up).
- 2. Form a Hamiltonian cycle (closed path) through this set of nodes such that the sum of the total length of the path plus the total cost of the selected nodes is minimized. The distances between nodes are calculated as Euclidean distances rounded mathematically to integer values.

The distance matrix should be calculated just after reading an instance and then only the distance matrix (no nodes coordinates) should be accessed by optimization methods to allow instances defined only by distance matrices.

2 Corrections from report 1

Node costs in neighborhood

It's been pointed out that neighborhood in the previous report should have included node costs (which I intentionally left out when computing nearest neighbor). This led to higher (worse) results than what's possible when including the cost. Fixed in f4ca6c8.

Major error in greedy cycle

Furthermore, I have discovered a major error in my greedy cycle implementation, that renders all results incredulous. The error regarded pairwise node iteration, which is critical for computing deltas.

Fixed in d66d28c.

Updated results are available in the *Results* section. For updated visualizations, see here.

3 Pseudocode

3.1 RCL

Input set of graph nodes G, set of candidate nodes N, size S, percentage P

Output An RCL N' (subset of N) of size S, chosen by repeatedly sampling from P% best nodes

Steps

- 1. Initialize N' to an empty list.
- 2. While size(N') \neq S, do:
 - (a) Construct a subset M of N, which contains a random P% of nodes from N.
 - (b) If size(G) = 0, then:
 - i. Pick a random node from M and move it from N to N'.
 - (c) Else If size(G) = 1, then:
 - i. Pick a node from M which is the **nearest neighbor** to the only node in G and move it from N to N'.
 - (d) Else:
 - i. Pick a node from M which constitutes the smallest increase in the objective function (greedy cycle) for G; move it from N to N'.

3.2 2_regret

Input set of graph nodes G, candidate node n

Output 2-regret of n with respect to G.

Steps

- 1. Initialize d1 to the objective function delta if n was inserted between the *first* pair of nodes from G.
- 2. Initialize d2 to the objective function delta if n was inserted between the *second* pair of nodes from G ("second pair" = 2nd and 3rd, **not** 3rd and 4th!).
- 3. Ensure d1 > d2 (swap values if necessary).
- 4. For every remaining (3rd, 4th, ...) pair p of nodes in G, do:
 - (a) Compute the objective function delta d if n was inserted between p.
 - (b) If d > d1, then:

- i. Let d2 = d1.
- ii. Let d1 = d.
- (c) Else if d > d2, then:
 - i. Let d1 = d.
- 5. Return d1 d2.

3.3 2-regret solution

Input A set N of nodes.

Output An ordered subset N' of N, with half of N's cardinality.

Steps

- 1. Initialize N' to an empty list.
- 2. While len(N') < len(N)/2, do:
 - (a) Let rcl = RCL(N', 10, 0.04).
 - (b) Choose node n from N, such that 2 regret(N', n) is maximal.
 - (c) Remove n from N.
 - (d) Add n to N'.

3.4 Weighted sum criterion solution (50/50)

Input A set N of nodes.

Output An ordered subset N' of N, with half of N's cardinality.

Steps

- 1. Initialize N' to an empty list.
- 2. While len(N') < len(N)/2, do:
 - (a) Let rcl = RCL(N', 10, 0.04).
 - (b) Choose node n from N, such that 2_regret(N', n) delta is maximal, where delta denotes the best (minimal) objective function change attainable by inserting n anywhere in N'.
 - (c) Remove n from N.
 - (d) Add n to N'.

4 Results

ALG.	FILE	min	avg	max
	TSPA.csv	241,510	266,062	308,034
random	TSPB.csv	241,731	266,549	293,093
Tailuoili	TSPC.csv	189,473	215,587	239,581
	TSPD.csv	195,876	218,919	250,422
	TSPA.csv	110,035 84,471	116,145 87,649	125,805 95,013
nn	TSPB.csv	106,815 77,448	116,181 79,304	124,675 82,669
nn	TSPC.csv	62,629 56,304	66,196 58,877	71,616 63,304
	TSPD.csv	62,788 50,335	66,847 54,406	71,396 59,846
	TSPA.csv	113,298 75,666	123,691 77,069	129,175 80,321
cycle	TSPB.csv	111,981 68,764	120,922 70,685	131,174 76,324
Cycle	TSPC.csv	67,077 53,226	72,771 55,845	75,763 58,876
	TSPD.csv	66,193 50,409	71,606 55,055	77,797 60,077
2-regret	TSPA.csv	86,224	103,385	119,577
	TSPB.csv	83,179	97,807	113,449
Z-regret	TSPC.csv	58,778	66,161	71,821
	TSPD.csv	53,396	63,387	69,441
	TSPA.csv	80,427	87,553	101,498
MCC	TSPB.csv	73,310	82,983	95,733
wsc	TSPC.csv	54,306	60,334	66,962
	TSPD.csv	51,420	58,478	66,263

5 Visualizations

5.1 TSPA.csv

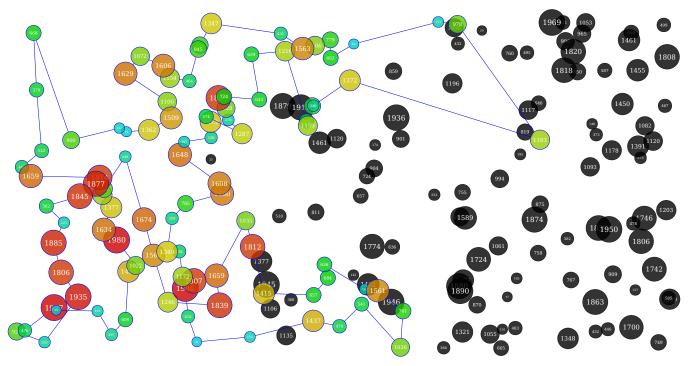


Figure 1: Best 2-regret solution to TSPA (119,577)

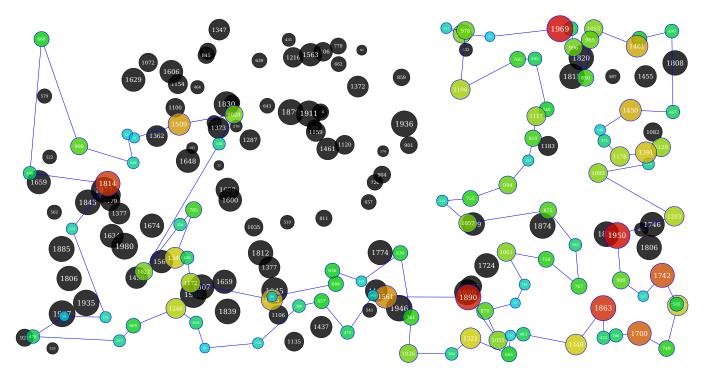


Figure 2: Best weighted-sum-criterion solution to TSPA (101,498)

5.2 TSPB.csv

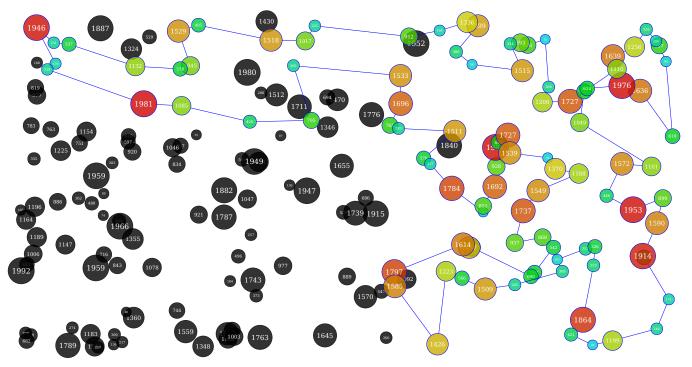


Figure 3: Best 2-regret solution to TSPB (113,449)

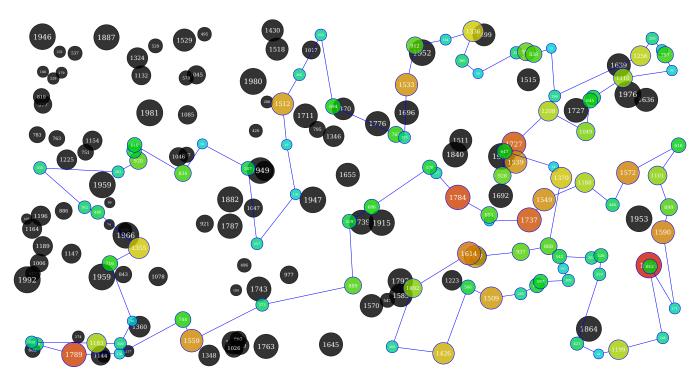


Figure 4: Best weighted-sum-criterion solution to TSPB (95,733)

5.3 TSPC.csv

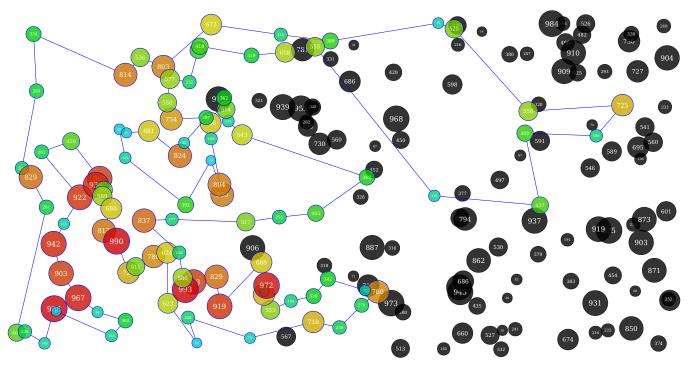


Figure 5: Best 2-regret solution to TSPC (71,821)

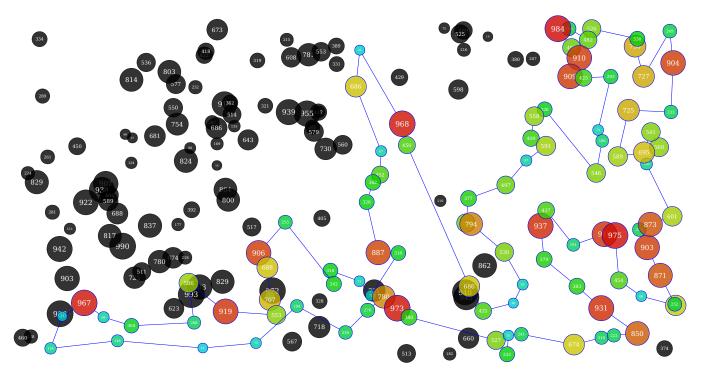


Figure 6: Best weighted-sum-criterion solution to TSPC (66,962)

5.4 TSPD.csv

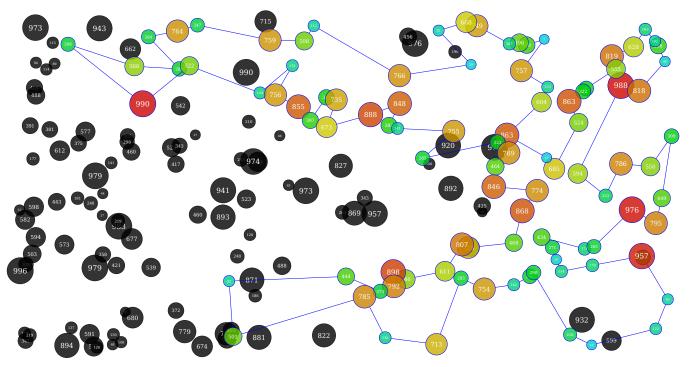


Figure 7: Best 2-regret solution to TSPD (69,441)

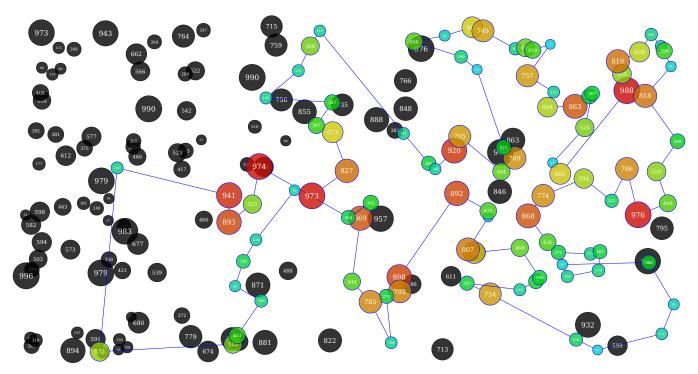


Figure 8: Best weighted-sum-criterion solution to TSPD (66,263)

6 Source code

The source code for all the experiments and this report is hosted on GitHub: https://github.com/RoyalDonkey/put-ec-tasks

7 Conclusions

After updating the first report, greedy cycle became the best method, and results obtained in this report seem to reinforce that – While 2-regret is pretty good, it gives worse solutions and weighing it with greedy cycle always leads to improvements.

In hindsight, I don't think implementing Restricted Candidate List was at all necessary for this assignment, but I did it anyway (initially I thought I needed it, and then it was too late to turn back).