

Programming Assignment 1

– Solving TSP by Searching

CMSC 421 Summer 2024
Due Date: 11:59 PM July 31st 2024

Welcome to your first programming assignment. In this assignment you are going to explore searching algorithms on the [Traveling Salesman Problem \(TSP\)](#). The goal of this assignment is for you to explore various mathematical and algorithmic trade offs. So you are encouraged to collaborate, discuss and compare results with your classmates. You may work in groups of three at most. The final code and lab report submitted should include the name of all the team members.

Stuff you will need to do:

1. Familiarize yourself with the [ATSPA](#) and [TSPA](#) datasets. You may use any software you see fit. Feel free to adapt/copy code from the internet. Please cite the sources with proper citation. Hints: [ATSPA](#) and [TSPA](#) are hosted on the [ATSPA](#) website. You may also find them on Piazza.
2. Represent a graph as an adjacency matrix. The matrix should be a square matrix with up to N^2 elements. The matrix should be a square matrix with up to N^2 elements. The matrix should be a square matrix with up to N^2 elements.
3. Read an $N * N$ matrix. The matrix should be a square matrix with up to N^2 elements. The matrix should be a square matrix with up to N^2 elements. The matrix should be a square matrix with up to N^2 elements.
4. Write your output in a file. The file should contain 4 entries: the solution quality, the solution accuracy, the run time, and the solution space. The file should contain 4 entries: the solution quality, the solution accuracy, the run time, and the solution space.
5. Perform basic search algorithms. The search algorithms should be implemented in a way that is efficient and accurate. The search algorithms should be implemented in a way that is efficient and accurate.
6. Use your output to generate a report. The report should be generated using a GUI or other means. The report should be generated using a GUI or other means.
7. Perform experiments and conduct some empirical experimentation to explore various mathematical and algorithmic trade offs.

Grading rubrics:

- Presentation of the results in graphical form. Several experiments are defined for each of which there is a capstone plot or chart showing tradeoffs in solution quality and accuracy and measures of time/space used to reach that solution (submitted via GRADESCOPE). Are these graphs illustrative of the phenomena expected?

```
n      /* integer # of rows/columns , all matrices will be square */
M11, M12, ..., M1n /* each row of the matrix separated by a new line */
M21, M22, ..., M2n /* you can assume these are integers */
...
Mn1, Mn2, ..., Mnn
/* end of file */
```

Figure 1: Format of .txt

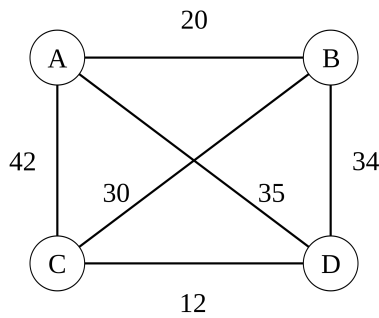


Figure 2: TSP Graph

```

4
0, 20, 42, 35
20, 0, 30, 34
42, 30, 0, 12
35, 34, 12, 0

```

Figure 3: Example of `infile.txt`

- Discussion of the experiments. For each of the experiments a short narrative explaining the phenomena observed will be submitted (via GRADESCOPE). Does this narrative accurately describe the mathematical and algorithmic phenomena behind the observations?

Part 1: Exploring Graph Algorithms

Implement A^* on a random graph.

1. $h(n) = 0$. This is a heuristic that always returns 0. This is a valid heuristic, but it is not very useful.
2. Grab edges randomly. For each node, grab n edges randomly. This is a valid heuristic, but it is not very useful.
3. Cheapest remaining edge. This is a valid heuristic, but it is not very useful.

For the experiment:

1. Randomly create a graph with n nodes, start with n edges.
2. Run the above algorithms.
3. For each family of graphs, record the cost, number of edges, and runtime.
4. Use those data to compare the algorithms. For each family of graphs, record the cost, number of edges, and runtime.

Compare and discuss the results in a short paragraph. Questions you can explore are for example:

- Which algorithm provides solution with the lowest cost? What's the difference of their best solutions, and how that changes when the size of the graph increases?
- Which algorithm has the least runtime and how do their runtimes change with the size of the graph increases?
- Is there difference between GPU and real-world runtime?

What to submit:

Code: Your implementation of three functions: `A.uniformCost()`, `A.randomEdge()` and `A.cheapestEdge()`. Each function should first read in the graph/matrix from `infile.txt`. Then perform the algorithm and finally return the cost of the best solution.

Report: Two graph plots of your experiment results and your discussion.

Part 2: Explore Local Search Algorithm

In this exercise, we explore the use of local search methods to approach TSPs.

1. Implement and test a hill-climbing method to approach TSPs.
2. Implement and test a simulated annealing method to approach TSPs.
3. Implement and test a genetic algorithm method to approach TSPs. (see section 4.3 of [Larranaga et al. 1999](#))

Compare these results with each other and with the optimal solutions obtained from the A^* algorithm with the MST heuristic. Compare not just the quality of the results but the time it takes to obtain them. Give an assessment of the relationship between computation time and solution quality, i.e., if all I wanted was a 75% solution, which method is quickest? How about a 90% solution? Is there a general rule here?

For the experiment:

1. Randomly create a family of 30 TSP graphs/matrices for each size 30 (or you can experiment with the size).
2. Run the A^* algorithm to obtain optimal solutions.
3. Run the above algorithms to obtain near-optimal solutions.
4. Try different algorithms and parameters:
 - number of iterations
 - number of neighbors to consider
 - number of iterations for simulated annealing
 - number of iterations for genetic algorithm
 - length of crossover, and mutation
5. Fix your parameters and run the algorithms on the 30 graphs. Record the cost, number of iterations, and CPU time.
6. Use those data to create a graph plot. The x-axis is the Cost, the y-axis is the number of iterations. For each graph, the cost between the optimal solution and the local searching solution. Use a different shape/color to represent AVE.
7. Maybe repeat the experiment with different parameters.



What to submit:

Code: Your implementation of three functions: `hillClimbing()`, `simuAnnealing()` and `genetic()`.

Report: At least three graph plots of your experiment results and a short paragraph of your discussion.

1 Extra Credit

Use your implementation from Part 2 to solve the Knapsack problem. See the [0-1 knapsack problem definition](#).

2 Submission Instructions

Submit a single zip file to Gradescope by 11:59 PM on July 31st. This should include your implementations of the following functions:

- `A.uniformCost()`
- `A.randomEdge()`

- `A.cheapestEdge()`
- `hillClimbing()`
- `simuAnnealing()`
- `genetic()`

It should also include any additional code necessary for running these functions.

