
TSP using Ant Colony Optimization

Group 2

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Project Goal

Travelling salesman problem (TSP) is a widely known NP-hard problem. It asks the following question: "Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city and returns to the origin city?". In this project, we try to provide a solution to TSP through a meta-heuristic technique called Ant Colony Optimization (ACO).

Overview

- Travelling Salesman Problem
- Ant Colony Optimization
- Flow Diagram
- Code Snippets (commented)
- User Interface
- Types of errors
- Output
- Future Work
- Acknowledgements
- Reference/Sources

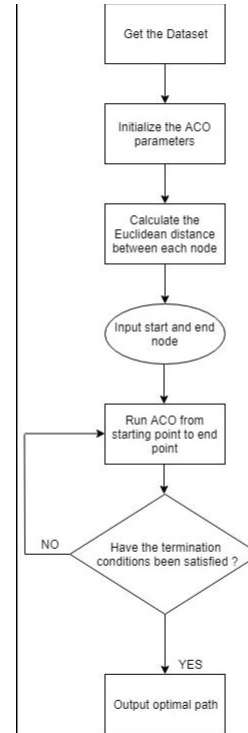
Travelling Salesman Problem

- Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city?
 - NP-hard problem
 - No solution which can solve it in polynomial time
-

Ant Colony Optimization

- Probabilistic technique for solving problems which can be reduced to finding good paths through graphs.
 - Optimization technique
 - Inspired by behavior of real ants
-

Flow Diagram



Code Snippets

```
11 # Parameters used -
12 # index - stores the city number
13 # x - x coordinate for the city i
14 # y - y coordinate for the city i
15 class City:
16     def __init__(self,i=0, x_cord=0, y_cord=0):
17         self.index = i
18         self.x = x_cord
19         self.y = y_cord
20
21 # Parameters used -
22 # num_cities - number of cities for travel
23 # initial_pheromone - initial value of pheromone deposited
24 # alpha - weight parameter for pheromones
25 # beta - weight parameter for desirability (attractiveness), which is inverse of distance
26 # pheromone_deposit - amount of pheromones which can be deposited
27 # evaporation_constant - amount of pheromone which will evaporate after a cycle
28 class ACO:
29     def __init__(self, num_cities, initial_pheromone=1, alpha=1, beta=3,
30                 pheromone_deposit=1, evaporation_constant=0.6):
31         self.cities = []
32         self.shortest_paths = []
33         self.shortest_paths_lens = []
34         self.shortest_path_len = -1
35         self.evaporationConst = evaporation_constant
36         self.pheromone_deposit = pheromone_deposit
37         self.pheromone = numpy.full((num_cities,num_cities),initial_pheromone)
38         self.alpha = alpha
39         self.beta = beta
40         self.attractiveness = numpy.zeros((num_cities,num_cities))
41         self.routing_table = numpy.full((num_cities,num_cities),(1.00/(num_cities-1)))
```

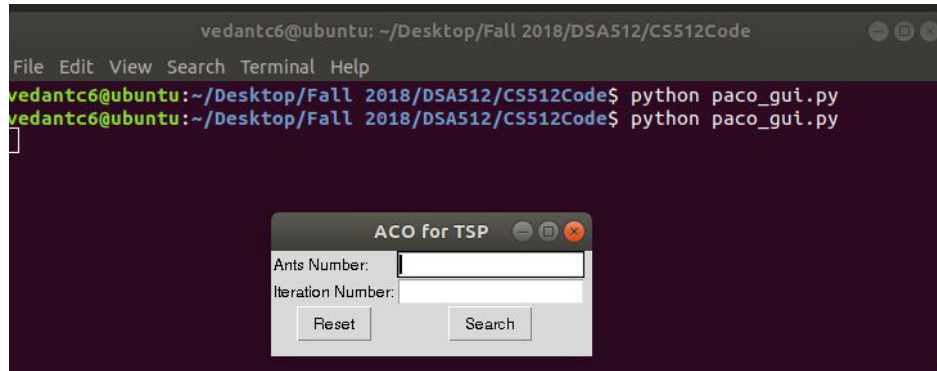
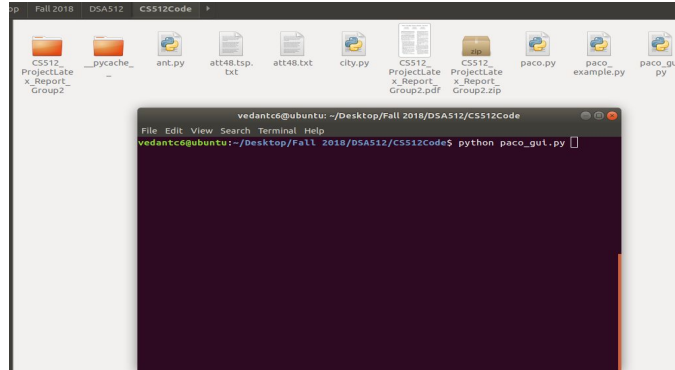
Code Snippets

```
50 # Use euclidean formula to calculate distance between two cities.
51 # Cities class has x and y coordinates
52 def euclidean_distance(self,city1,city2):
53     return (math.sqrt(math.pow((city1.x-city2.x),2)+math.pow((city1.y-city2.y),2)))
54
55 # Attractiveness is reciprocal of distance. This function calculates
56 # attractiveness between cities.
57 def calc_attraction(self):
58     city_list=self.cities
59     for i in range(len(city_list)):
60         for j in range(len(city_list)):
61             distance = self.euclidean_distance(city_list[i], city_list[j])
62             if distance > 0:
63                 self.attractiveness[i][j] = 1/distance
64             else:
65                 self.attractiveness[i][j] = 0.00
66
67 # a - a single ant
68 # pheromone deposit on each path is updated by the ant visiting that path
69 # we have deposited the pheromone uniformly across the path
70 # additionally, evaporation part is removed from the pheromone
71 def update_pheromone(self,a):
72     for i in range(0,len(a.path)-1):
73         try:
74             curr_pher = self.pheromone[a.path[i].index][a.path[i+1].index]
75             self.pheromone[a.path[i].index][a.path[i+1].index] = curr_pher + self.pheromone_deposit/a.path_length
76         except:
77             break
78         self.pheromone = self.pheromone*(1-self.evaporationConst)
79
80 # get the pheromone value for the path between city i and j
81 def get_pheromone(self,i,j):
82     return self.pheromone[i][j]
83
```

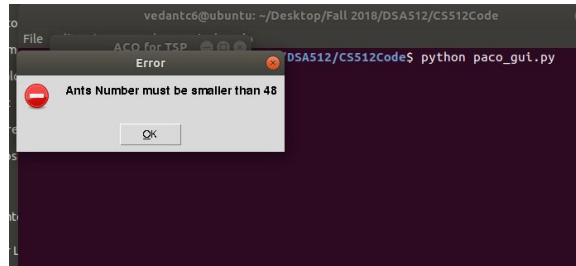
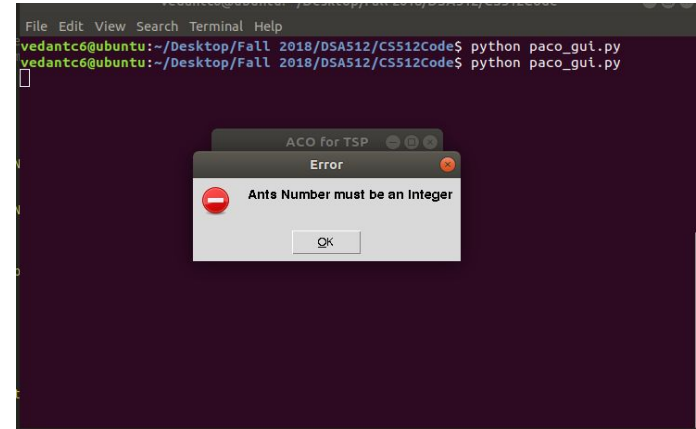
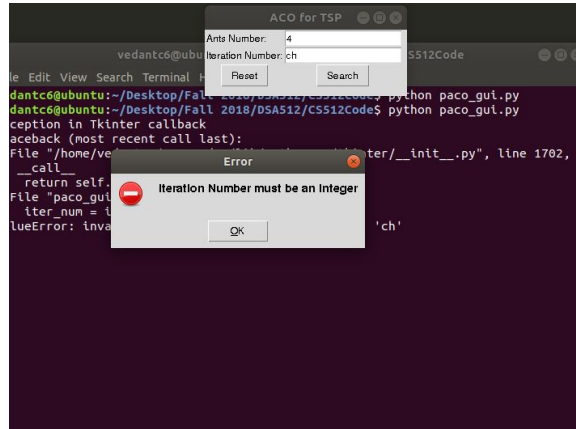
Code Snippets

```
83
84     # Calculation for numerator part of probability distribution for the next city selection
85     def city_sum(self, city_cur, city_next):
86         return ((math.pow(self.pheromone[city_cur.index][city_next.index],
87             *
88             self.alpha))*(math.pow(self.attractiveness[city_cur.index][city_next.index], self.beta)))
89
90     # Calculating probability density for the next city selection,
91     # and storing that value in the routing table
92     def update_routing_table(self, a):
93         denom = 0.0
94         for c in a.path:
95             temp_cities = list(a.path)
96             temp_cities.remove(c)
97             for valid in temp_cities:
98                 denom += self.city_sum(c, valid)
99
100         for valid in temp_cities:
101             numerator = self.city_sum(c, valid)
102             if denom > 0:
103                 self.routing_table[c.index][valid.index] = numerator/denom
104             else:
105                 self.routing_table[c.index][valid.index] = 0
```

User Interface



Types of Errors



The image is a composite of three screenshots from a Linux desktop environment.

Top Left: Terminal Window
 The terminal shows the execution of a Python script. It displays the progress of the Ant Colony Optimization (ACO) algorithm over 10 steps. The best path found is 130536.41872197976, and the shortest route found is 44970.602. The number of ants used is 4.

```

File Edit View Search Terminal
ValueError: invalid literal
Step: 1 of 10
Step best path: 130536.41872197976 Step: 1
Step: 2 of 10
Step best path: 44970.60225166842 Step: 2
Step: 3 of 10
Step best path: 44970.60225166842 Step: 3
Step: 4 of 10
Step best path: 44970.60225166842 Step: 4
Step: 5 of 10
Step best path: 44970.60225166842 Step: 5
Step: 6 of 10
Step best path: 44970.60225166842 Step: 6
Step: 7 of 10
Step best path: 44970.60225166842 Step: 7
Step: 8 of 10
Step best path: 44970.60225166842 Step: 8
Step: 9 of 10
Step best path: 44970.60225166842 Step: 9
Step: 10 of 10
Step best path: 44970.60225166842 Step: 10
Number of ants used: 4
Shortest route found: 44970.602

```

Top Right: ACO for TSP GUI
 A small window titled "ACO for TSP" with a "Ants Number:" field set to 4 and an "Iteration Number:" field set to 10. It has "Reset" and "Search" buttons.

Bottom: Plot of the Solution
 A window titled "Figure 1" showing a plot titled "Visualization of ACO algorithms on TSP (48 cities)". The plot shows the coordinates of 48 cities (X and Y) and the shortest route found by the ACO algorithm, which is a complex path connecting all cities. The plot includes axes for "X - Coordinates of the cities" and "Y - Coordinates of the cities".

Future Work

Try to make UI more attractive by `tkinter.ttk`

Try to ask the user the number of cities the user wants to see, so number of cities also a parameter in UI

Try to allow the user to load tsp instance with the pre-predefined format

Acknowledgements

We would like to thank Professor Antonio Miranda Garcia , Harsha Srimath Tirumala for supporting us and giving useful insights throughout the development of this project.

Reference

- [1] Y. Zhou, “Runtime analysis of an ant colony optimization algorithm for tsp instances,” IEEE Transactions on Evolutionary Computation, vol. 13, pp. 1083–1092, 2009.
- [2] T. Kotzing, F. Neumann, H. Roglin, and C. Witt, “Theoretical analysis of two aco approaches for the traveling salesman problem,” Swarm Intelligence , vol. 6, pp. 1–21, 03 2012.

https://en.wikipedia.org/wiki/Ant_colony_optimization_algorithms

Resource

Python3.6

Tkinter

<http://elib.zib.de/pub/mp-testdata/tsp/tsplib/tsp/>
