## **Experiment 3: Ant Colony Optimisation**

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# This program is developed to find shortest path (minimum cost)between

some cities.

There are 4 parts in this program: 1.Main program of ACO (myaco.m) 2.Function to generate solution (ant\_tour.m) 3.Function to calculate the cost (distance) (calculate\_cost.m) 4.Function to update the traces (update\_the\_trace.m)

```
% Theory: Ant Colony Optimization (ACO) are a set of probabilistic
% metaheuristics and an intelligent optimization algorithms, inspired
by
% social behavior of ants. ACO algorithms are also categorized as
Swarm
% Intelligence methods, because of implementation of this paradigm,
via
% simulation of ants behavior in the structure of these algorithms.
```

#### Code:

```
function myaco()
% inputs
miter=10;
m=10;
n=10;
% parameters
e=.15;
                  % evaporation coefficient.
alpha=1;
                  % effect of ants' sight.
                  % trace's effect.
t=0.0001*ones(n); % primary tracing.
el=.97;
                  % common cost elimination.
% Generate coordinates of cities and plot
for i=1:n
    x(i)=rand*20;
    y(i) = rand*20;
end
subplot(3,1,1);
```

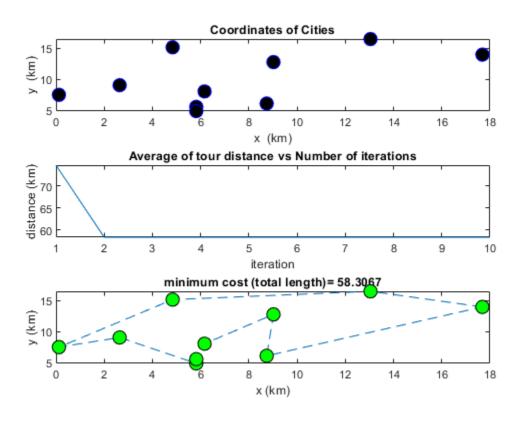
### Experiment 3: Ant Colony Optimisation

```
plot(x,y,'o','MarkerFaceColor','k','MarkerEdgeColor','b','MarkerSize',10);
title('Coordinates of Cities');
xlabel('x (km)');
ylabel('y (km)');
% generating distace between cities matrix.
for i=1:n
    for j=1:n
        d(i,j) = sqrt((x(i)-x(j))^2+(y(i)-y(j))^2);
    end
end
% generating sight matrix.
for i=1:n
    for j=1:n
        if d(i,j) == 0
            h(i,j)=0;
        else
            h(i,j)=1/d(i,j);
        end
    end
end
h=h
             Main Algorithm: ACO Meta heuristic procedure
% a. Probabilistic solution construction biased by
    pheromone trails, without forward pheromone
응
    updating
% b. Deterministic backward path with loop elimination
     and with pheromone updating--> update_the_trace
% c. Evaluation of the quality of the solutions
     generated and use of the solution quality in
      determining the quantity of pheromone to deposit--
>calculate cost
for i=1:miter
% Step 1: Forward ants and solution construction
% Generate places for each ant
for j=1:m
    start_places(j,1)=fix(1+rand*(n-1));
end
% Step 2:probabilistic solution contruction
    [tour] = ant tour(start places, m, n, h, t, alpha, beta);
    tour=horzcat(tour,tour(:,1));
% Step 3: Calculate the cost --> total distace
    [cost,f]=calculate_cost(m,n,d,tour,el);
    [t]=update_the_trace(m,n,t,tour,f,e);
    average_cost(i)=mean(cost);
% Step 4: Determine the best route
```

```
[min_cost(i),best_index]=min(cost);
   besttour(i,:)=tour(best index,:);
   iteration(i)=i;
end
% Plot Average of tour distance vs Number of Iterations
subplot(3,1,2);plot(iteration,average_cost);
title('Average of tour distance vs Number of iterations');
xlabel('iteration');
ylabel('distance (km)');
% Plot the best route
[k,l]=min(min cost);
for i=1:n+1
   X(i)=x(besttour(l,i));
   Y(i)=y(besttour(1,i));
subplot(3,1,3); plot(X,Y,'--o',...
              'MarkerEdgeColor','k',...
              'MarkerFaceColor','g',...
              'MarkerSize',10)
xlabel('x (km)');ylabel('y (km)');
title(['minimum cost (total length)= ',num2str(k)]);
end
h =
 Columns 1 through 7
            0.1352
                     0.1820
                              0.1803
                                       0.1143
                                                0.2075
                                                         0.1261
        0
   0.1352
                     0.0780 0.2730
                                       0.0631
                                               0.1533
                                                         0.2113
            0
                              0.0915
   0.1820
            0.0780
                      0
                                       0.1892
                                                0.1203
                                                        0.0760
   0.1803
           0.2730 0.0915
                                       0.0771 0.1375 0.3965
                               0
   0.1143
           0.0631 0.1892 0.0771
                                       0
                                                0.0775 0.0686
   0.2075
           0.1533 0.1203 0.1375
                                       0.0775
                                                0
                                                         0.1031
   0.1261
           0.2113 0.0760
                            0.3965
                                      0.0686
                                               0.1031
                                                             0
   0.0965
           0.3374 0.0634 0.1649
                                      0.0534
                                                         0.1661
                                               0.1107
   0.1488
           0.1474 0.0884 0.3088
                                      0.0837
                                               0.1009
                                                         0.3368
   0.1167
           0.1897
                    0.0728 0.3110
                                      0.0667
                                               0.0962
                                                        1.4317
 Columns 8 through 10
   0.0965
            0.1488
                     0.1167
   0.3374
           0.1474
                     0.1897
   0.0634
           0.0884
                     0.0728
   0.1649
            0.3088
                    0.3110
   0.0534
            0.0837
                    0.0667
   0.1107
                   0.0962
           0.1009
   0.1661
           0.3368
                    1.4317
            0.1145 0.1593
        0
   0.1145
            0
                    0.3153
```

0.1593 0.3153

0



## **Conclusion:**

This code implementation of Ant Colony Optimization (ACO) to solve traveling #salesman problem (TSP). Given a list of cities and their pairwise distances, the task is to find a shortest #possible tour that visits each city exactly once. #And hence through the output graphs plotted we can observe the desired output.

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