

Simulated Annealing

Large Scale Optimization for Data Science

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Implementation of the Simulated Annealing Algorithm in solving the (Symmetric) Travelling Salesman Problem.

In the outer loop of our algorithm, we reduce T by setting $T = 0.9 \times T$. While, in the inner loop, we obtain L different solutions at each T . These solutions are generated by randomly swapping two indexes(cities) of tour x .

Then we apply the Metropolis acceptance Probability= $\min \{1, \exp(-(f_y - f_x)/T)\}$ as the probability of accepting a new possible solution. The algorithm then repeats while $T > \epsilon$, with $\epsilon = 10^{-2}$, and stops otherwise.

In [5]:

```
import numpy as np
import math

class SimulatedAnnealing:
    def __init__(self,x,M,T,n,L,E):
        #Step 1 - Initialize values to what is passed in.
        self.initialX = x
        self.M = M
        self.T = T
        self.n = n
        self.L = L
        self.E = E

    # Distance function to calculate length of tour
    def calculateLengthOfTour(self,tour):
        sumDistance = 0
        for ind in range(len(tour)):
            v_1 = tour[ind]

            #If at the end of tour, use home V
            if(ind == (len(tour) -1 )):
                v_2 = tour[0]
            else:
                v_2 = tour[ind+1]

            sumDistance += self.M[v_1][v_2]

        return sumDistance

    # Function to generate a new tour, with 2 cities swapped at random
    def generateNewTour(self,x):
        # Step 3.1 - 3.2 - replace = False ensures no duplicates indexes
        # Choosing two non-repeating random numbers for indexes
        rangeForRandomNumbers = np.arange(len(x))
        uniqueRandomIndexes = np.random.choice(rangeForRandomNumbers,2,replace=False)
        n1 = uniqueRandomIndexes[0]
        n2 = uniqueRandomIndexes[1]

        # Step 3.3 - 3.4 - create y
        y = np.copy(x)
        temp = y[n1]
        y[n1] = y[n2]
        y[n2] = temp

        return y
```

```

def run(self):
    x = self.initialX
    fx = self.calculateLengthOfTour(x)
    T = self.T
    E = self.E
    # Step 2 - looping while T > E
    while T > E:
        # Step 3 - run for length L
        # L - The total number of solutions generated at a particular temperature.
        for i in range(L):
            # Step 3.1 - 3.3 - create new tour y
            y = self.generateNewTour(x)
            # Step 3.4 - evaluate y
            fy = self.calculateLengthOfTour(y)

            # Step 4
            if(fy < fx):
                x = y
                fx = fy
            else:
                # Generate random number between 0 and 1
                randomNum = np.random.rand()
                # Metropolis acceptance Probability
                #  $\exp(-(fy-fx)/T)$  function is same as  $e^{-(fy-fx)/T}$ 
                if(randomNum < math.exp(-(fy-fx)/T)):
                    x = y
                    fx = fy

        # End forloop
        # Step 5 - Decreasing T
        T = 0.9*T

    #End while loop

    #Return best tour x and best fx
    return x,fx

```

In [6]:

```

#Step 1 - Initialize Values
T = 15
n = 7
L = 20
E = 10 ** -2
#Equivalent to ([1,7,2,4,6,5,3]), starting at index 0
x = np.array([0,6,1,3,5,4,2])
M = np.array(
    [ 0, 1, 3, 5, 2, 1, 1,
      1, 0, 1, 6, 9, 4, 3,
      3, 1, 0, 1, 5, 3, 2,
      5, 6, 1, 0, 1, 2, 5,
      2, 9, 5, 1, 0, 1, 6,
      1, 4, 3, 2, 1, 0, 1,
      1, 3, 2, 5, 6, 1, 0
    ])
M = M.reshape(n,n)
SA = SimulatedAnnealing(x,M,T,n,L,E)

# Run simulated annealing algorithm
bestx,bestfx = SA.run()
print("Best tour: ",bestx,"Best fx: ",bestfx)

```

Best tour: [2 1 0 6 5 4 3] Best fx: 7