

TSP_Anealing_Personal

February 26, 2023

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[ ]: '''
Resources:
https://towardsdatascience.com/
    ↪how-to-solve-the-traveling-salesman-problem-a-comparative-analysis-39056a916c9f
https://github.com/syakoo/simulated-annealing
https://towardsdatascience.com/
    ↪how-to-solve-travelling-salesman-problem-with-simulated-annealing-c248447a8bcd
https://towardsdatascience.com/simulated-annealing-with-restart-a19a53d914c8
https://github.com/jedrazb/python-tsp-simulated-annealing
https://en.wikipedia.org/wiki/2-opt

'''

import math
import random
import numpy as np
import pandas as pd
import plotly.express as px
from plotly.subplots import make_subplots
import plotly.graph_objects as go
import os
import matplotlib.pyplot as plt
class tsp_utils:
    #Convert coordant information into distance matrix commonly used in TSP
    ↪Problems
    def vectorToDistMatrix(coords):
        '''
        Create the distance matrix
        '''
        return np.sqrt((np.square(coords[:, np.newaxis] - coords).sum(axis=2)))

    def nearestNeighbourSolution(dist_matrix):
        '''
        Computes the initial solution (nearest neighbour strategy)
        '''
        #grabs random node in the size of the matrix
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node = random.randrange(len(dist_matrix))
#stores node in result list
result = [node]
#sets nodes to visit equal to the size of the distance matrix
nodes_to_visit = list(range(len(dist_matrix)))
#removes first node
nodes_to_visit.remove(node)
#goes through and tries to find the nearest neighbour for every node
while nodes_to_visit:
    nearest_node = min([(dist_matrix[node][j], j) for j in ↪
nodes_to_visit], key=lambda x: x[0])
    node = nearest_node[1]
    nodes_to_visit.remove(node)
    result.append(node)

return result

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[ ]: #Class to store the data generated by the annealing algorithm
class solution:
    def __init__(self, coords):
        #converts data form coordants into distmatrix
        self.dist_matrix = tsp_utils.vectorToDistMatrix(coords)
        #uses distance matrix to find solution
        self.saved_solution = tsp_utils.nearestNeighbourSolution(self.
↪dist_matrix)

    def __iter__(self):
        return self

    def get_matrix(self):
        return self.dist_matrix
    def get_solution(self):
        return self.saved_solution

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[ ]: class MSA:
    def __init__(self, coords, iterations, temp, gamma, stopping_temp):

        #intial coordant points
        self.coords = coords
        #size of array
        self.sample_size = len(coords)
        #initialize iteration counter
        self.iteration = 1
        #check point to stop iterations
        self.stopping_iter = iterations

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        #tempature
        self.temp = temp
        #tempature to stop program
        self.stopping_temp = stopping_temp
        #self.df = df
        #gamma value to multiply tempature by
        self.gamma = gamma
        #tmp variable to store the data in the solution object
        self.tmp_obj = solution(coords)
        #current solution based on the return value fo the object
        self.current = self.tmp_obj.get_solution()
        #gets the distance matrix from the object
        self.matrix = self.tmp_obj.get_matrix()
        #saves all the solutions in a list
        self.solution_history = [self.current]
        #saves the current solution as the best solution
        self.best = self.current
        #saves the weight as the current weight
        self.cweight = self.weight(self.current)
        #saves the current weight as the immediate weight weight of current_
↪solution
        self.iweight = self.cweight
        #saves the current weight as the minimum weight
        self.mweight = self.cweight
        #saves the weights into a list
        self.weight_list = [self.cweight]
        #prints the immediate weight
        print('Intial weight: ', self.iweight)

#unused function to calcuate distance
@staticmethod
def total_distance(df):

    def euclidean_distance(x1, x2, y1, y2):
        return np.sqrt((x1-x2)**2+(y1-y2)**2)

    distance = 0
    for idx in range(0, len(df)):
        if idx + 1 >= len(df):
            break
        distance += euclidean_distance(df['x'].loc[idx], df['x'].loc[idx+1],
                                       df['y'].loc[idx], df['y'].loc[idx+1])
    return distance

def weight(self, sol):

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        #Calculate weight

        return sum([self.matrix[i, j] for i, j in zip(sol, sol[1:] + [sol[0]])])

    def probability(self, candidate_weight):
        """
        Acceptance probability as described in:
        https://stackoverflow.com/questions/19757551/
        ↪basics-of-simulated-annealing-in-python
        """
        return math.exp(-abs(candidate_weight - self.cweight) / self.temp)

    def accept(self, candidate):
        """
        Accept with probability 1 if candidate solution is better than
        current solution, else accept with probability equal to the
        acceptance_probability()
        """
        #sets the candidate_weight as the weight of the current solution being
        ↪tested
        candidate_weight = self.weight(candidate)
        #if the candidate weight is better than the current weight then the
        ↪current solution is updated with the candidate information as well as the
        ↪miniumm weight is checked and updated based on the candidate information
        if candidate_weight < self.cweight:
            self.cweight = candidate_weight
            self.current = candidate
            if candidate_weight < self.mweight:
                self.mweight = candidate_weight
                self.best_solution = candidate
        #if the candidate_weight is not better than than the current weight it
        ↪is run through the probabiltiy function and if that probability is less than
        ↪the random that candidate is saved as the current weight
        else:
            if random.random() < self.probability(candidate_weight):
                self.cweight = candidate_weight
                self.current = candidate

    def anneal(self):
        """
        Annealing process with 2-opt
        described here: https://en.wikipedia.org/wiki/2-opt
        """
        #if the tempt threshold or the iteration threshold is not reached then
        ↪it continues through the while loop

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        while self.temp >= self.stopping_temp and self.iteration < self.
↳stopping_iter:
            #candidate is set to the current solution value
            candidate = list(self.current)
            #l and I are randomly generated based on the sample size defined
↳above
            l = random.randint(2, self.sample_size - 1)
            i = random.randint(0, self.sample_size - 1)
            #goes through each element of the candiate based on the randomly
↳generated values and reverses them
            candidate[i: (i + 1)] = reversed(candidate[i: (i + 1)])
            #the reversed candiate is passed through the acceptance check if it
↳passes the current information will be updated for the next look
            self.accept(candidate)
            #the tempiture is decreased based on the predefined gamma value
            self.temp *= self.gamma
            #an itteration is added
            self.iteration += 1
            #the weight list is appened wit the current weight
            self.weight_list.append(self.cweight)
            #the solution history is appended with the current solution
            self.solution_history.append(self.current)

        #once the loop finishes the minimum weight is printed
        print('Minimum weight: ', self.mweight)
        #improvement calculation based on the starting weight and the minimum
↳weight
        print('Improvement: ',
              round((self.iweight - self.mweight) / (self.iweight), 4) * 100,
↳'%')

        #plot the output of the solutions
        def plotLearning(self):
            #plot the weight graph
            plt.plot([i for i in range(len(self.weight_list))], self.weight_list)
            line_init = plt.axhline(y=self.iweight, color='r', linestyle='--')
            line_min = plt.axhline(y=self.mweight, color='g', linestyle='--')
            plt.legend([line_init, line_min], ['Initial weight', 'Optimized
↳weight'])
            plt.ylabel('Weight')
            plt.xlabel('Iteration')
            #plt.show()

        def plotResults(self, name):
            #set up figure
            fig, ax = plt.subplots()

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#set line weight
line, = plt.plot([], [], lw=2)
''' initialize node dots on graph '''
x = [self.coords[i][0] for i in self.solution_history[0]]
y = [self.coords[i][1] for i in self.solution_history[0]]
plt.plot(x, y, 'co')

''' draw axes slightly bigger '''
extra_x = (max(x) - min(x)) * 0.05
extra_y = (max(y) - min(y)) * 0.05
ax.set_xlim(min(x) - extra_x, max(x) + extra_x)
ax.set_ylim(min(y) - extra_y, max(y) + extra_y)
#draw last solution on the graph using the starting solutions points
↳and the last solutions lines
x = [self.coords[i, 0] for i in self.solution_history[len(self.
↳solution_history)-1] + [self.solution_history[len(self.
↳solution_history)-1][0]]
y = [self.coords[i, 1] for i in self.solution_history[len(self.
↳solution_history)-1] + [self.solution_history[len(self.
↳solution_history)-1][0]]
#set lines
line.set_data(x, y)
#add lable to print the final weight
plt.xlabel('Weight:'+str(self.mweight))
#create path to images
if not os.path.exists("./images"):
    os.mkdir("./images")
#save the plot for the graph based on the input name with the default
↳being city_final
plt.savefig("./images/"+name)
#plt.show
def get_mweight(self):
    #get minimum weight
    return self.mweight

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[ ]: #run funtion with all the data needed
def run(nodes,stopping_iter, temp, gamma, stopping_temp,name="cities_final.
↳png", weight= np.Inf ):
    '''run simulated annealing algorithm with 2-opt'''
    #run simulated annealing
    sa = MSA(nodes,stopping_iter, temp, gamma, stopping_temp)

    sa.anneal()
    #set base weight first itteration is infinity
    current_weight = weight
    # get min_weight from the simulated annealing instance

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min_wieght = sa.get_mweight()
'''animate'''

#sa.animateSolutions()
'''show the improvement over time'''
if(min_wieght<current_weight):
    #if the min_weight is less than the current weight then the plot for
    ↳that annealing is generated and the new minweight is returend
    sa.plotLearning()
    sa.plotResults(name)
    return min_wieght
else:
    #else the current weight is returned
    return current_weight

```

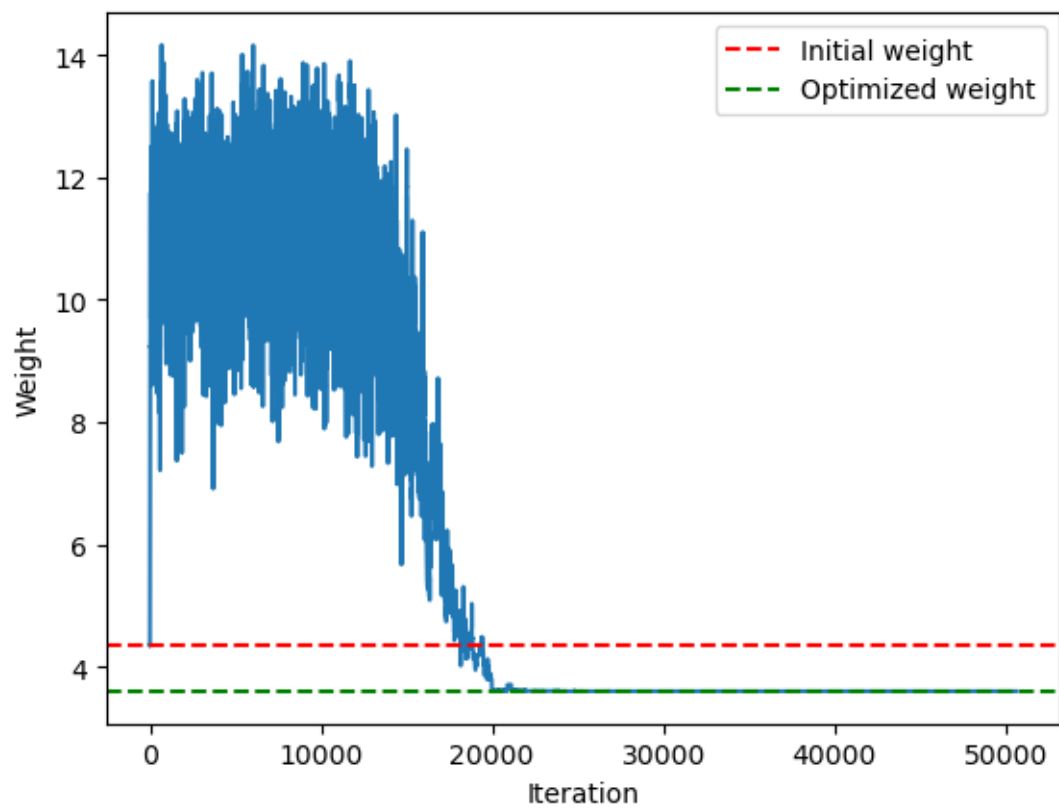
```

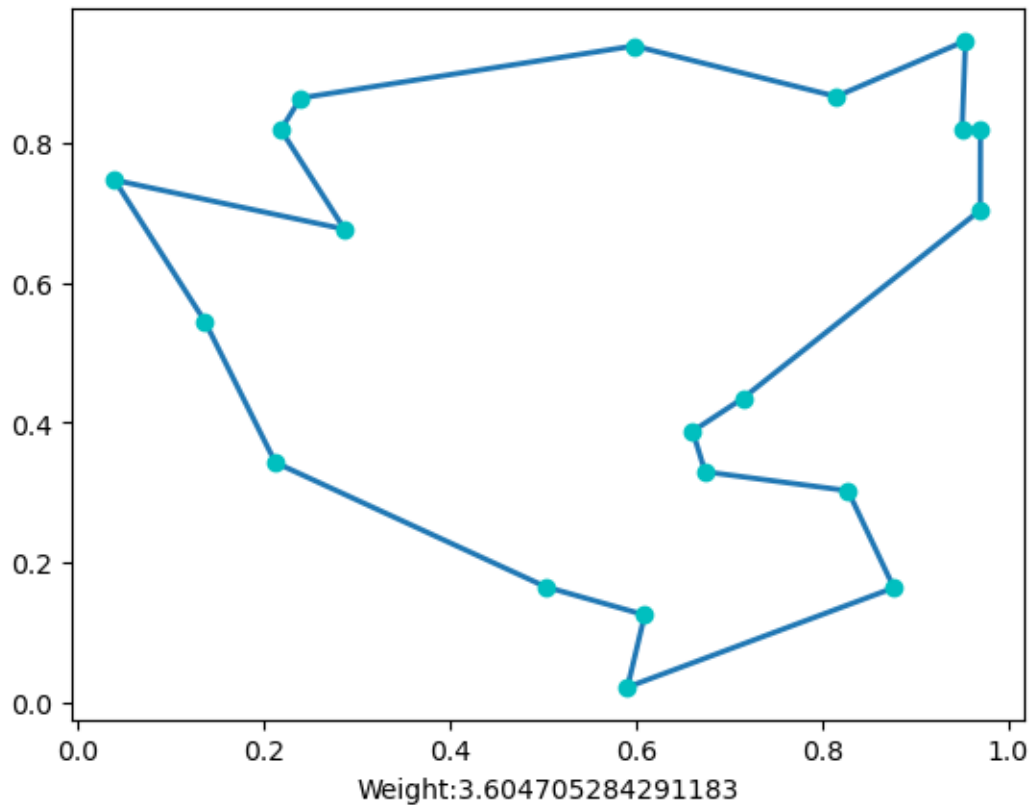
[ ]: '''set the simulated annealing algorithm params'''
    #test run based on wrong points
temp = 1000
stopping_temp = 0.00000001
gamma = 0.9995
stopping_iter = 10000000
xs = [0.6606,0.9695,0.5906, 0.2124, 0.0398, 0.1367,0.9536,0.6091,0.8767,0.
    ↳8148,0.9500,0.6740,0.5029,0.8274,0.9697,0.5979,0.2184,0.7148,0.2395,0.2867]
ys = [0.3876,0.7041,0.0213,0.3429,0.7471,0.5449,0.9464,0.1247,0.1636,0.8668,0.
    ↳8200,0.3296,0.1649,0.3025,0.8192,0.9392,0.8191,0.4351,0.8646,0.6768]
#print(f"X: {d} && Y:{s}",len(xs),len(ys))
nodes = np.column_stack((xs, ys))
run(nodes,stopping_iter, temp, gamma, stopping_temp)

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Intial weight: 4.3580045910318415
 Minimum weight: 3.604705284291183
 Improvement: 17.29 %

[]: 3.604705284291183

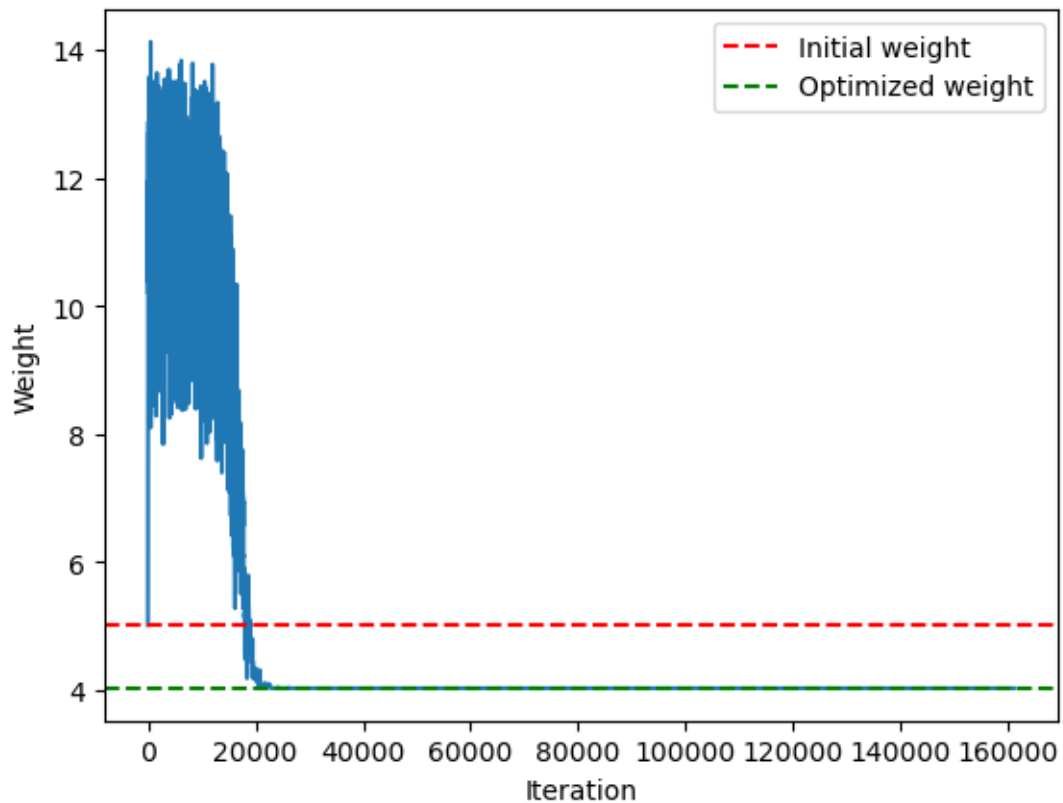


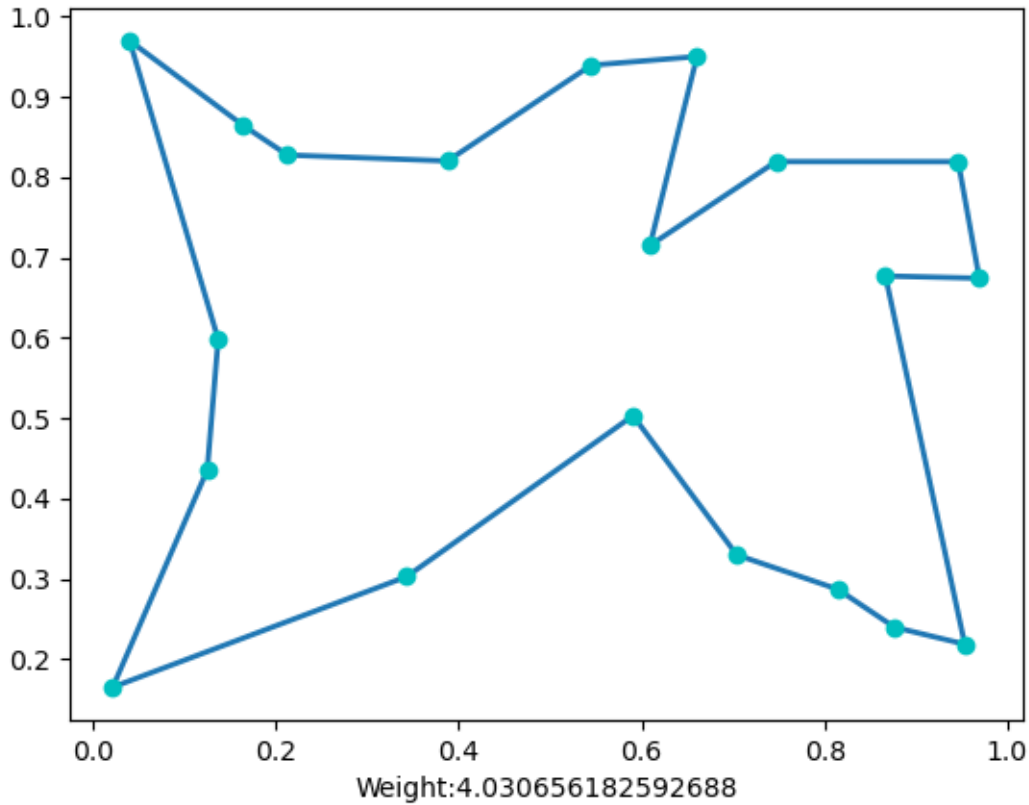
[illegible]

Intial weight: 5.014274936724524
 Minimum weight: 4.030656182592688
 Improvement: 19.62 %
 Intial weight: 4.502090428504454
 Minimum weight: 4.076046089443103
 Improvement: 9.46 %
 Intial weight: 4.444147300922404
 Minimum weight: 4.076046089443103
 Improvement: 8.28 %
 Intial weight: 4.8067459348638515
 Minimum weight: 4.076046089443103
 Improvement: 15.2 %
 Intial weight: 5.452254618497038
 Minimum weight: 4.054910629557462
 Improvement: 25.629999999999995 %
 Intial weight: 5.22894797811956
 Minimum weight: 4.076046089443103
 Improvement: 22.05 %
 Intial weight: 4.909123268390923
 Minimum weight: 4.030656182592689
 Improvement: 17.89 %
 Intial weight: 4.8322153627319455
 Minimum weight: 4.030656182592688
 Improvement: 16.59 %
 Intial weight: 4.986850665914826
 Minimum weight: 4.030656182592688
 Improvement: 19.17 %
 Intial weight: 4.908185133302875
 Minimum weight: 4.076046089443103
 Improvement: 16.950000000000003 %
 Intial weight: 5.014274936724524
 Minimum weight: 4.076046089443103
 Improvement: 18.709999999999997 %
 Intial weight: 4.887909931685917
 Minimum weight: 4.054910629557462
 Improvement: 17.04 %
 Intial weight: 5.424835518657478
 Minimum weight: 4.076046089443103
 Improvement: 24.86 %
 Intial weight: 4.444147300922404
 Minimum weight: 4.030656182592688
 Improvement: 9.3 %
 Intial weight: 5.050834563973035
 Minimum weight: 4.030656182592689
 Improvement: 20.200000000000003 %
 Intial weight: 4.8322153627319455
 Minimum weight: 4.030656182592688
 Improvement: 16.59 %

Intial weight: 4.502090428504454
Minimum weight: 4.030656182592689
Improvement: 10.47 %
Intial weight: 4.926483896898324
Minimum weight: 4.054910629557462
Improvement: 17.69 %
Intial weight: 4.502090428504454
Minimum weight: 4.030656182592688
Improvement: 10.47 %
Intial weight: 5.307040759194025
Minimum weight: 4.030656182592689
Improvement: 24.05 %

```
[ ]: <function matplotlib.pyplot.show(close=None, block=None)>
```





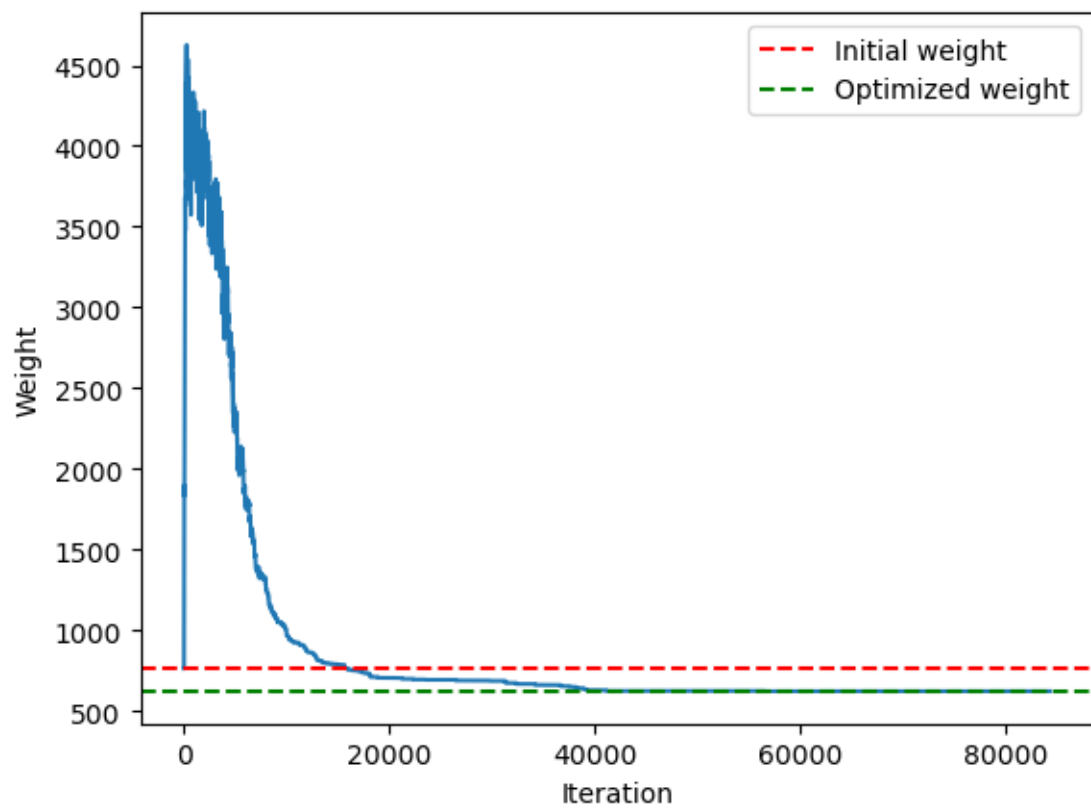
```
[ ]: #data for the 101 cities
data = [1, 41, 49, 2, 35, 17, 3, 55, 45, 4, 55, 20, 5, 15, 30, 6, 25, 30, 7,
↪20, 50, 8, 10, 43, 9, 55, 60, 10, 30, 60, 11, 20, 65, 12, 50, 35, 13, 30,
↪25, 14, 15, 10, 15, 30, 5, 16, 10, 20, 17, 5, 30, 18, 20, 40, 19, 15, 60,
↪20, 45, 65, 21, 45, 20, 22, 45, 10, 23, 55, 5, 24, 65, 35, 25, 65, 20, 26,
↪45, 30, 27, 35, 40, 28, 41, 37, 29, 64, 42, 30, 40, 60, 31, 31, 52, 32, 35,
↪69, 33, 53, 52, 34, 65, 55, 35, 63, 65, 36, 2, 60, 37, 20, 20, 38, 5, 5, 39,
↪60, 12, 40, 40, 25, 41, 42, 7, 42, 24, 12, 43, 23, 3, 44, 11, 14, 45, 6, 38,
↪46, 2, 48, 47, 8, 56, 48, 13, 52, 49, 6, 68, 50, 47, 47, 51, 49, 58, 52, 27,
↪43, 53, 37, 31, 54, 57, 29, 55, 63, 23, 56, 53, 12, 57, 32, 12, 58, 36, 26,
↪59, 21, 24, 60, 17, 34, 61, 12, 24, 62, 24, 58, 63, 27, 69, 64, 15, 77, 65,
↪62, 77, 66, 49, 73, 67, 67, 5, 68, 56, 39, 69, 37, 47, 70, 37, 56, 71, 57,
↪68, 72, 47, 16, 73, 44, 17, 74, 46, 13, 75, 49, 11, 76, 49, 42, 77, 53, 43,
↪78, 61, 52, 79, 57, 48, 80, 56, 37, 81, 55, 54, 82, 15, 47, 83, 14, 37, 84,
↪11, 31, 85, 16, 22, 86, 4, 18, 87, 28, 18, 88, 26, 52, 89, 26, 35, 90, 31,
↪67, 91, 15, 19, 92, 22, 22, 93, 18, 24, 94, 26, 27, 95, 25, 24, 96, 22, 27,
↪97, 25, 21, 98, 19, 21, 99, 20, 26, 100, 18, 18, 101, 35, 35]
```

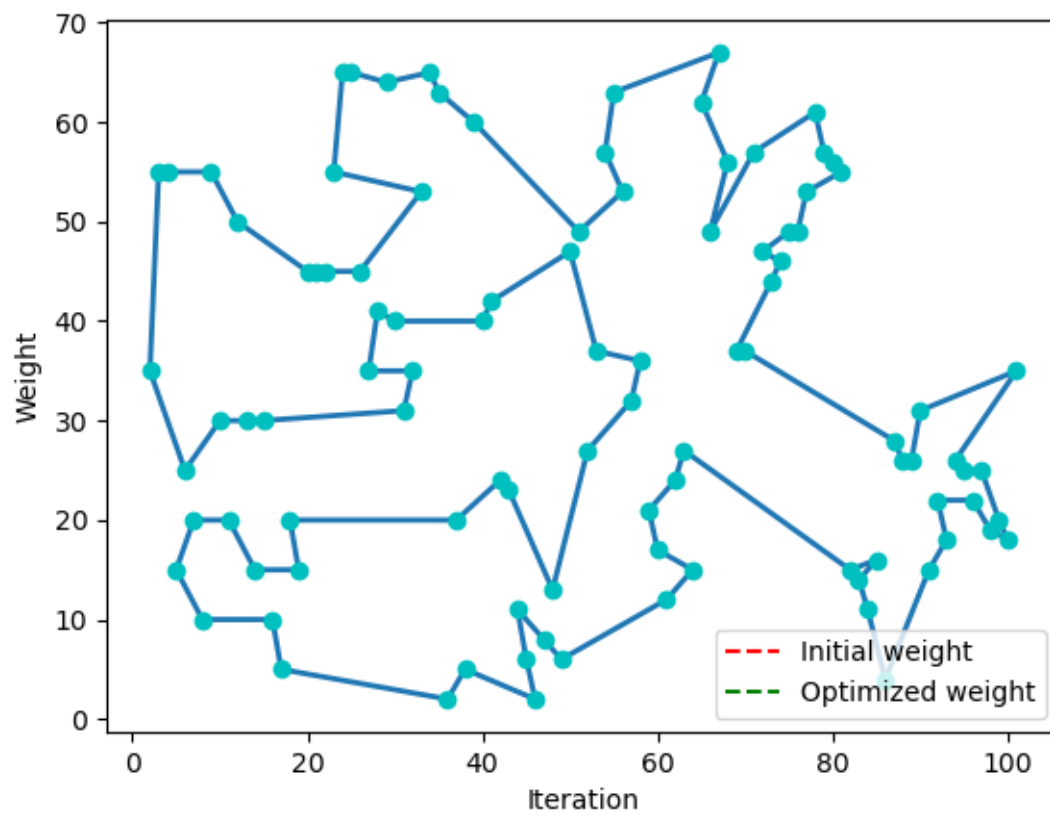
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[ ]:
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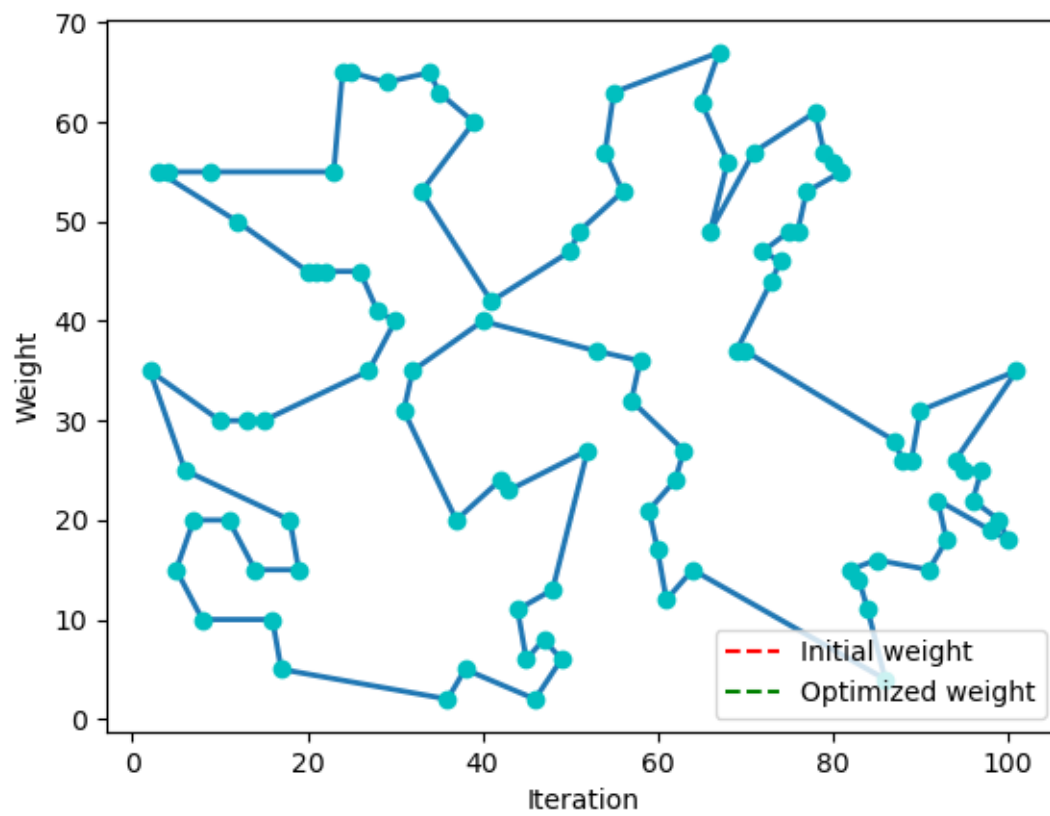

Intial weight: 690.4244499371883
 Minimum weight: 656.5884713316545
 Improvement: 4.9 %
 Intial weight: 739.7171044878103
 Minimum weight: 635.9543103610816
 Improvement: 14.030000000000001 %
 Intial weight: 716.8068069288466
 Minimum weight: 631.4209182484159
 Improvement: 11.91 %
 Intial weight: 757.5822449590481
 Minimum weight: 652.6573300302492
 Improvement: 13.850000000000001 %
 Intial weight: 714.5547584686362
 Minimum weight: 651.4843068627639
 Improvement: 8.83 %
 Intial weight: 718.1690276678853
 Minimum weight: 637.3444681942419
 Improvement: 11.25 %
 Intial weight: 757.5822449590481
 Minimum weight: 631.2214191717109
 Improvement: 16.68 %
 Intial weight: 740.2210042365543
 Minimum weight: 603.3790296087124
 Improvement: 18.490000000000002 %
 Intial weight: 743.2558992691793
 Minimum weight: 603.269397662606
 Improvement: 18.83 %
 Intial weight: 685.9020640792157
 Minimum weight: 621.7365808162575
 Improvement: 9.35 %
 Intial weight: 776.8879194254547
 Minimum weight: 618.1219384703933
 Improvement: 20.44 %
 Intial weight: 721.4447971075517
 Minimum weight: 620.1733156921345
 Improvement: 14.04 %
 Intial weight: 737.8696413893713
 Minimum weight: 634.7078258663577
 Improvement: 13.98 %
 Intial weight: 757.5822449590481
 Minimum weight: 617.0200530403616
 Improvement: 18.55 %
 Intial weight: 739.102229198969
 Minimum weight: 652.8039757362905
 Improvement: 11.68 %
 Intial weight: 785.9984307333415
 Minimum weight: 638.577744063067
 Improvement: 18.759999999999998 %

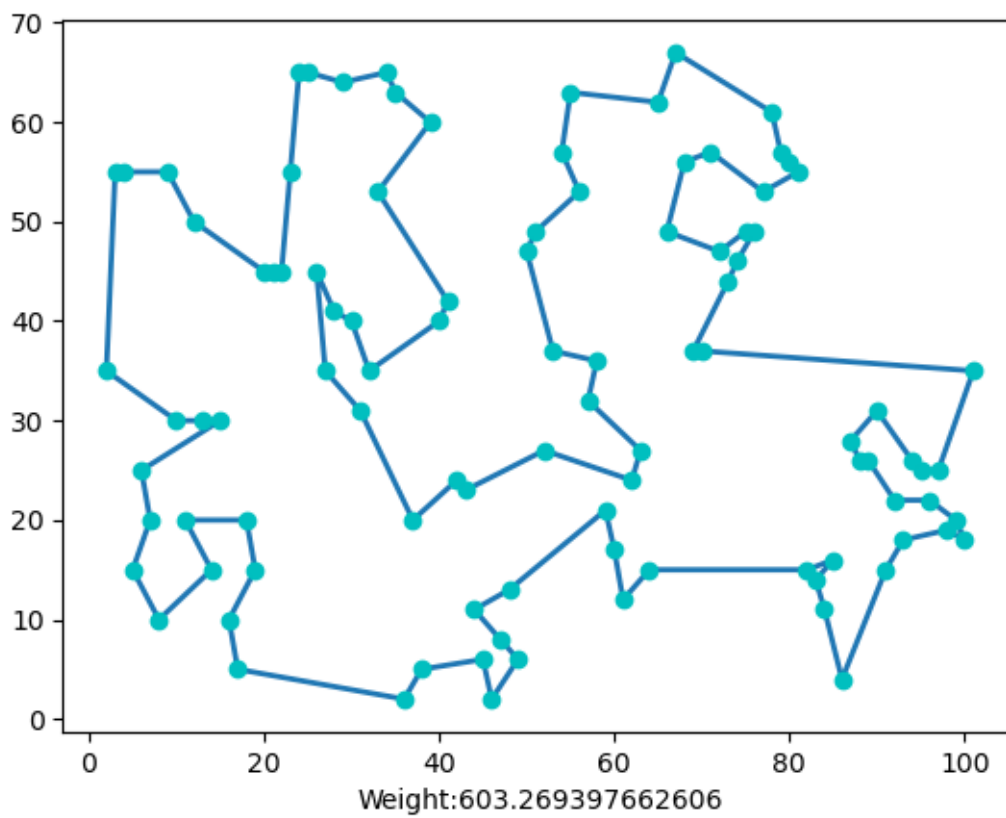
Intial weight: 748.1413216793228
Minimum weight: 630.6103973859985
Improvement: 15.709999999999999 %
Intial weight: 742.1723015979768
Minimum weight: 609.7893925594691
Improvement: 17.84 %
Intial weight: 743.3439569772712
Minimum weight: 633.0655260866142
Improvement: 14.84 %
Intial weight: 774.5778980539324
Minimum weight: 627.7292744033464
Improvement: 18.96 %
Intial weight: 738.4029105009527
Minimum weight: 637.9221040511126
Improvement: 13.61 %
Intial weight: 725.2657342092399
Minimum weight: 622.9892644282279
Improvement: 14.099999999999998 %
Intial weight: 757.5822449590481
Minimum weight: 631.844006783133
Improvement: 16.6 %
Intial weight: 759.3522636136561
Minimum weight: 632.4211769519478
Improvement: 16.72 %
Intial weight: 738.4029105009527
Minimum weight: 634.4192052963676
Improvement: 14.08 %
Intial weight: 720.9008801406616
Minimum weight: 664.3924360680728
Improvement: 7.84 %

```
[ ]: <function matplotlib.pyplot.show(close=None, block=None)>
```









[]: