

EE2703 - Week 7

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1 Importing Libraries

```
[1]: %matplotlib ipynb
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.animation import FuncAnimation
from matplotlib import cm
from functools import partial
```

2 General annealing function

The below function is a general annealing function.

```
[2]: def optimize(cost, random_jump, init_guess, T, decay,
                k, iterations, save_list=False):
    """
    Optimize a function through annealing, and return
    a list of intermediate guesses.

    -----

    `cost` - calculate cost of guess
    `random_jump` - take a guess and a distance, and move
    randomly
    `init_guess` - initial (random) guess
    `T` - starting temperature
    `decay` - decay rate for temperature
    `k` - boltzmann constant
    `iterations` - number of iterations to make
    `save_list` - if `True`, this returns the of intermediate
    guesses made -- otherwise, returns just the final guess
    """

    p_arr = None
```

```

if save_list:
    p_arr = [init_guess]

x = init_guess
c = cost(x)

for i in range(iterations):
    print(f"Progress: {100 * i / iterations:<8.4}%", end="\r")
    # create a new guess
    new_x = random_jump(
        x,
        (np.random.random_sample() - 0.5) * T
    )

    new_c = cost(new_x)

    if new_c < c: # if it's better, jump
        x = new_x
        c = new_c
        if save_list:
            p_arr.append(x)
    else:
        # otherwise, jump with some probability
        toss = np.random.random_sample()
        if toss < np.exp(-(new_c - c) / (k * T)):
            x = new_x
            c = new_c
            if save_list:
                p_arr.append(x)

    T *= decay

if save_list:
    return p_arr
else:
    return x

```

3 Travelling Salesman

3.1 Useful functions

These functions are used in optimizing the travelling salesman problem using the general optimizer algorithm.

```
[3]: def tsp_cost(cities, guess):
    """
    Gives the cost of travelling between cities.

    -----

    `cities` is the list of co-ordinates of the cities
    `guess` is the order in which you visit the cities
    """
    sum_dist = 0

    for i in range(-1, len(guess) - 1):
        sum_dist += np.linalg.norm(cities[guess[i]] - cities[guess[i + 1]])

    return sum_dist

def tsp_jump(guess, distance):
    """
    Swap up to `distance` + 1 values in `guess` to make a
    new guess
    """
    swaps = np.random.randint(int(np.abs(distance)) + 1) + 1
    # print(swaps)
    new_guess = guess.copy()

    for _ in range(swaps):
        swap0, swap1 = np.random.randint(len(guess), size=2)
        tmp = new_guess[swap0]
        new_guess[swap0] = new_guess[swap1]
        new_guess[swap1] = tmp

    return new_guess
```

Run the run.py file for animations.

3.2 Choosing input file

3.2.1 10 cities, 500 iterations

```
[4]: input_filename = "tsp_10.txt"
    start_temp_scale = 1/5
    iterations = 500
```

3.2.2 100 cities, 5,000 iterations

```
[4]: input_filename = "tsp_100.txt"
     start_temp_scale = 1/10
     iterations = 5000
```

3.2.3 100 cities, 30,000 iterations

```
[ ]: input_filename = "tsp_100.txt"
     start_temp_scale = 1/5
     iterations = 30000
```

3.3 Read the list of cities

```
[5]: lines = open(input_filename, "r").readlines()

     # list of co-ordinates
     cities = []

     for line in lines[1:]:
         s = line.split()
         cities.append(np.array([float(s[0]), float(s[1])]))
```

3.4 Prepare the initial guess

We can start with a random guess, or start with a more optimal one based on some heuristic. One of the 2 cells below should be run.

3.4.1 Random initial guess

```
[7]: init_guess = list(range(len(cities)))
```

3.4.2 Optimized initial guess

We greedily choose the nearest unvisited neighbour to each point to create an initial guess.

```
[9]: def tsp_greedy_solve(cities):
     """
     Get an initial guess for the TSP greedily
     """
     guess = [0]
     unused = list(range(1, len(cities)))
     prev = cities[0]

     for i in range(1, len(cities)):
         min_city = unused[0]
         min_dist = np.linalg.norm(cities[i] - prev)
```

```

    for j in range(1, len(unused)):
        dist = np.linalg.norm(cities[unused[j]] - prev)
        if dist < min_dist:
            min_city = unused[j]
            min_dist = dist

    unused.remove(min_city)
    prev = cities[min_city]
    guess.append(min_city)

print(guess)
return guess

```

```
init_guess = tsp_greedy_solve(cities)
```

```

[0, 34, 95, 63, 38, 57, 12, 72, 92, 87, 45, 47, 68, 51, 49, 48, 75, 58, 46, 82,
98, 62, 33, 32, 79, 70, 81, 44, 37, 11, 9, 66, 20, 83, 91, 28, 64, 96, 73, 90,
71, 59, 29, 53, 97, 31, 55, 3, 54, 21, 22, 36, 80, 18, 99, 7, 27, 14, 93, 65,
52, 39, 15, 50, 42, 84, 25, 86, 30, 19, 6, 76, 16, 88, 17, 41, 4, 89, 2, 5, 26,
8, 23, 13, 85, 1, 43, 69, 35, 77, 67, 78, 94, 24, 10, 74, 60, 61, 40, 56]

```

3.5 Get result

```

[10]: optimum_path = optimize(
    partial(tsp_cost, cities),
    tsp_jump,
    init_guess,
    len(cities) * start_temp_scale,
    0.99,
    0.1,
    iterations,
    save_list=False
)

print()
print(optimum_path)
print(tsp_cost(cities, optimum_path))

```

```

Progress: 99.98 %
[34, 0, 95, 63, 38, 12, 57, 72, 92, 45, 56, 47, 68, 51, 49, 48, 75, 58, 82, 46,
98, 62, 33, 32, 79, 1, 44, 81, 37, 11, 9, 66, 20, 83, 91, 28, 64, 96, 73, 90,
71, 59, 29, 97, 53, 31, 55, 3, 54, 21, 22, 36, 80, 18, 99, 7, 27, 14, 93, 65,
52, 39, 15, 50, 42, 84, 86, 25, 30, 19, 6, 76, 16, 88, 17, 41, 4, 89, 2, 8, 26,
85, 23, 13, 10, 43, 70, 69, 77, 35, 67, 78, 5, 61, 94, 24, 60, 74, 40, 87]
89.48988006724814

```

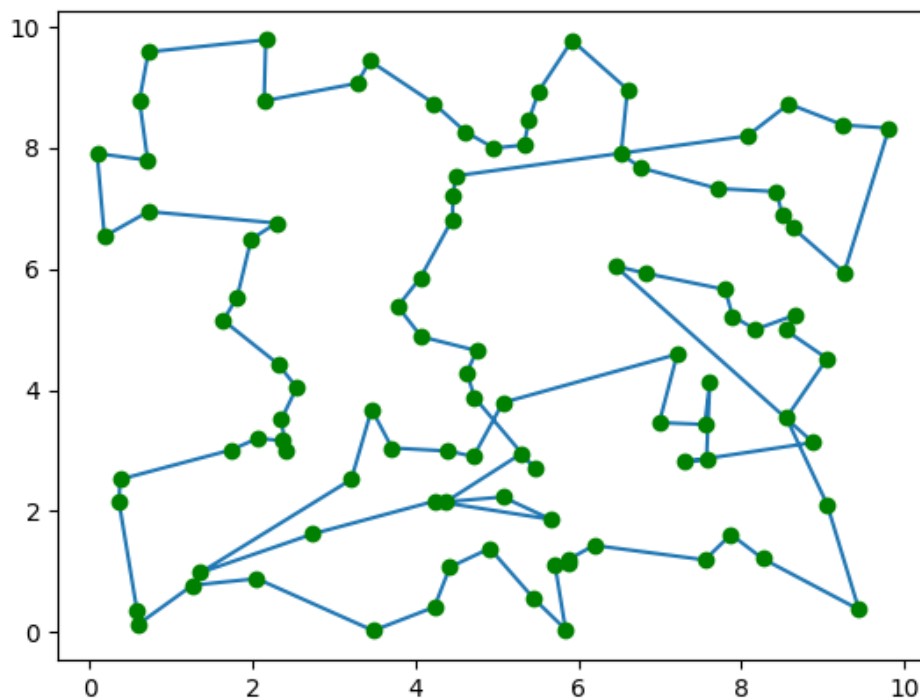
3.6 Plot result

```
[12]: fig, ax = plt.subplots()

ln, = ax.plot([], [])
lnpoints, = ax.plot(
    [cities[i][0] for i in range(len(cities))],
    [cities[i][1] for i in range(len(cities))],
    'go'
)

coords = []
for c in optimum_path:
    coords.append(cities[c])
coords.append(cities[optimum_path[0]])
coords = np.array(coords)
ln.set_data(coords[:, 0], coords[:, 1])

plt.show()
```



3.7 Animation

Please run `run.py` for animations.

4 Conclusions

If we ran the annealing function for 100 cities with a random initial guess, we get a cost of around 120, while with an optimized initial guess we can lower it to around 80.

Therefore, annealing on its own is not a good enough technique to minimize cost functions, we should also carefully select an initial guess.