

EE2703 week7

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```
[3]: # Set up imports
%matplotlib ipynb
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.animation import FuncAnimation
import math
import random
```

1 Part 1

```
[17]: # Function with many minima
def yfunc(x):
    return x**2 + np.sin(8*x)
xbase = np.linspace(-2, 2, 100)
ybase = yfunc(xbase)
```

```
[18]: T=3
decayrate=0.95
bestcost=10000
bestx=-2

rangemin, rangemax = -2, 2
fig, ax = plt.subplots()
ax.plot(xbase, ybase)
xall, yall = [], []
lnall, = ax.plot([], [], 'ro')
lngood, = ax.plot([], [], 'go', markersize=10)

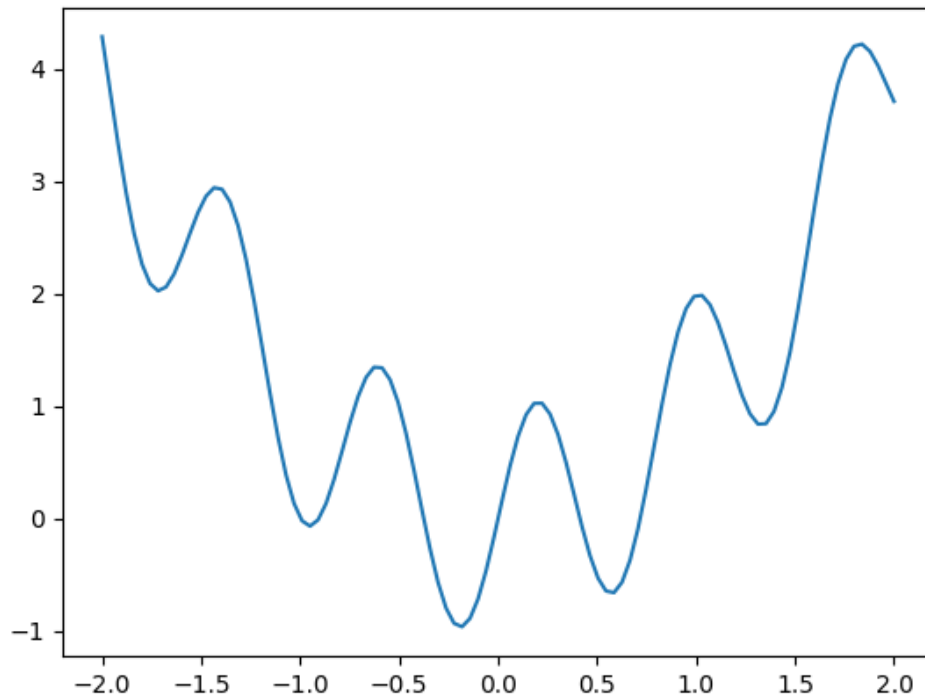
def f1(f,T,decayrate,bestx,bestcost):
    dx = (np.random.random_sample() - 0.5) * T
    x = bestx + dx
    # print(f"Old x = {x}, delta = {dx}")
    y = f(x)
    if y < bestcost:
        # print(f"Improved from {bestcost} at {bestx} to {y} at {x}")
        bestcost = y
        bestx = x
```

```

else:
    toss = np.random.random_sample()
    if toss < np.exp(-(y-bestcost)/T):
        bestcost = y
        bestx = x
        # print(f"New cost {y} worse than best so far: {bestcost}")
    T = T * decayrate
    return bestx,bestcost,T,x,y
def onestep(frame):
    global T,bestx,bestcost
    bestx,bestcost,T,x,y=f1(yfunc,T,decayrate,bestx,bestcost)
    xall.append(x)
    yall.append(y)
    lnall.set_data(xall, yall)
    lngood.set_data(bestx, bestcost)

ani= FuncAnimation(fig, onestep, frames=range(100), interval=100, repeat=False)
plt.show()

```



For part 1, I created a function that takes as inputs the function to be minimised, T, decayrate, bestx, bestcost and returns bestx, bestcost, T, x, y. And in onestep function I declared T,

bestx,bestcost as global variables so that they can be changed in function f1. I updated the values of T, bestx, bestcost for every frame and plotted them.

2 Part 2

2.1 Reading the inputs

```
[4]: filename="tsp_100.txt"
f=open(filename,"r")
p=f.readlines()
l=len(p)-1
c=[]
x=[]
y=[]
z=0
for line in p:
    if z==0:
        z=z+1
        continue
    p=line.split()
    d1=[]
    x.append(float(p[0]))
    y.append(float(p[1]))
    d1.append(float(p[0]))
    d1.append(float(p[1]))
    c.append(d1)

coords=np.array(c)
x=np.array(x)
y=np.array(y)
```

2.2 Solving

```
[5]: def tsp(coords, T, dr, ST,N):
    def path_dist(coords, path):
        distance = 0
        for i in range(len(path)-1):
            distance += math.dist(coords[path[i]], coords[path[i+1]])
        distance += math.dist(coords[path[-1]], coords[path[0]])
        return distance

    def AP(cost, new_cost, temp):
        if new_cost < cost:
            return 1.0
        else:
            return math.exp((cost - new_cost) / temp)
```

```

n = coords.shape[0]
# Create a random initial path
path = np.arange(n)
np.random.shuffle(path)
# Initialize the current best path and its length
best_path = path.copy()
best_dist = path_dist(coords, path)
# Initialize the current path and its length
current_path = path.copy()
current_dist = best_dist
# Initialize the iteration counter
i = 0
# Main loop
while T>=ST and i < N:
    # Generate a new path by swapping two cities
    new_path = current_path.copy()
    i1 = np.random.randint(0, n-1)
    i2 = np.random.randint(0, n-1)
    while i2 == i1:
        i2 = np.random.randint(0, n-1)
    new_path[i1], new_path[i2] = new_path[i2], new_path[i1]
    # Calculate the cost of the new path
    new_dist = path_dist(coords, new_path)
    # Decide whether to accept the new path
    ap = AP(current_dist, new_dist, T)
    if ap > random.random():
        current_path = new_path.copy()
        current_dist = new_dist
    # Update the current best path if necessary
    if current_dist < best_dist:
        best_path = current_path.copy()
        best_dist = current_dist
    # Decrease the temperature
    T = T * dr
    i = i + 1
return best_path, best_dist

```

I created a function that takes in as inputs coordinates of the cities, T , decayrate, Stopping Temperature (so that it won't go below that temperature) and number of iterations. First I calculated the distance travelled for a random path then I interchanged positions of two random cities and checked for the distance. Iterated the above process for N times and it returns the best path and distance among those calculated.

```

[11]: T = 2
      dr = 0.99
      ST = 1e-128
      N = 1000000

```

```
best_path, best_dist = tsp(coords, T, dr, ST, N)
print(f"Shortest path found: {best_path}")
print(f"Length of shortest path: {best_dist:.3f}")
```

```
Shortest path found: [64 77 35 58 75 82 46 98 62  1 81 44 32 33 23  8 26 67 69
83 66  9 11 37
 10 85 78  5 18 80 22  3 55 31 97 90 71 94 74 40 13 43 79 70 20 91 28 41
 17 88 34  0 95 50 39 52 65 27 14 93 15 42 12 57 38 63  6 76 16  4 89  2
 48 49 24 60 61 19 30 25 86 84 72 87 45 56 47 51 68 92 99  7 36 21 54 53
 59 29 96 73]
Length of shortest path: 130.170
```

3 Plot

```
[13]: xplot = x[best_path]
      yplot = y[best_path]
      xplot = np.append(xplot, xplot[0])
      yplot = np.append(yplot, yplot[0])
      plt.plot(xplot, yplot, 'o-')
      plt.show()
```

