

```

1 from bokeh.plotting import figure, show
2 import matplotlib.pyplot as plt
3 import pandas as pd
4 import networkx as nx
5 import seaborn as sns

```

Define the problem parameters such as n = number of nodes , adj = adjacency matrix of the graph ,
 $Graph_edge$ = array of graphs - this will simplify the graph visualization process - And lastly
 $default_n_community$ = number of communities

```

class Problem:
    def __init__(self, n, adj, graph_edge, default_n_community):
        self.n = n
        self.adj = adj
        self.m = int(sum([len(x) for x in self.adj])/2)
        self.default_n_community = default_n_community
        self.graph_edge = graph_edge

```

We want to create a random first population and improve on this. To this we first shuffle the array list then we will iterate on it. In each iteration we will choose members of a community randomly and assign their community to them. Done by this section of the code :

```

selected = random.choices(vertices, k=int(self.n / self.default_n_community))
vertices = [e for e in vertices if e not in selected]
for i in selected:
    individual_map[i] = counter
counter += 1

```

Keep in mind in the last iteration there is only one possible community to assign the remaining nodes

In the end we will sort the randomly generated populations based on the fitness function. We can see the formula and code below :

$$Q = \frac{1}{2m} \sum_{i,j} \left[A_{ij} - \frac{k_i k_j}{2m} \right] \delta(c_i, c_j)$$

```

self.population = sorted(self.population, key=lambda agent: self.fitness(
    agent), reverse=False)

```

subsequently the function which is calculating the fitness can be seen below

```
def fitness(self, individual):
    # Calculate the fitness of a chromosome based on the formula provided
    Q = 0
    for i in range(self.n):
        for j in range(self.n):
            Q += (int(j in self.adj[i]) - ((len(self.adj[i]) * len(self.adj[j])) / (2 * ε
    Q /= (2 * self.m)

    individual[1] = Q
    return Q
```

Now for the algorithm : In each step we will select a random number of candidates for parents(tournament_size) then choose the best "parent_size" based on fitness function.

This part is done by the selection function :

```
def selection(self):
    random.shuffle(self.population)
    self.parents = self.population[:self.tournament_size]
    self.parents = sorted(self.parents, key=lambda agent: self.fitness(
        agent), reverse=True)[:self.parents_size]
```

Of the selected parents candidates, we will randomly link pairs of them and then we will create two "children" of the pairs selected and create new offspring with a uniform crossover method.

```
def breed(self, parent1, parent2):
    # We will breed new children based on single point cossrossover
    child1 = [None, None]
    child2 = [None, None]
    temp1 = []
    temp2 = []
    for i in range(len(parent1[0])):
        choose = random.randint(1, 2)
        if choose == 1:
            temp1.append(parent1[0][i])
            temp2.append(parent2[0][i])
        else :
            temp1.append(parent2[0][i])
            temp2.append(parent1[0][i])
    child1[0] = temp1
    child2[0] = temp2
    return child1, child2
```

```

def breed_offsprings(self):
    self.children = []
    # Select two parent for breeding two children
    for _ in range(self.breed_rate):
        random.shuffle(self.parents)
        for i in range(int(len(self.parents)/2)):
            child1, child2 = self.breed(self.parents[i], self.parents[len(self.parents)-i-1])
            self.children.append(child1)
            self.children.append(child2)

```

We will mutate some of the children based on our mutation rate so we could avoid getting stuck at local minimas

```

def mutate(self, individual):
    # we will randomly mutate some genes based on our mutation rate
    individual_copy = [individual[0][:], None]
    # Choosing whether to mutate or not
    if random.random() < self.mutation_rate:
        # Changing a gene to another possible answer
        gene = random.choice(range(self.n))
        cluster = random.choice(list(set(individual[0])))
        # We need to make sure we don't randomly select the first one :)
        while cluster == individual[0][gene]:
            cluster = random.choice(list(set(individual[0])))
        individual_copy[0][gene] = cluster
    self.fitness(individual_copy)
    return individual_copy

```

Finally we will select the best of the parents, children and mutated children and repeat the process on the chosen ones.

```

def replacement(self):
    # We will consider the best of the children, mutated and parent and then select the k
    self.mutated_children = sorted(
        self.mutated_children, key=lambda agent: agent[1], reverse=True)
    self.population = sorted(
        self.parents, key=lambda agent: agent[1], reverse=True)

    self.population = self.mutated_children[:self.elite_size] + self.parents[:self.elite_size]
    self.population = sorted(
        self.population, key=lambda agent: self.fitness(agent), reverse=True)

```

```

1 import random
2 import copy
3 from itertools import combinations
4 from __future__ import print_function, division
5
6 %matplotlib inline
7 import matplotlib.pyplot as plt
8 import numpy as np
9 import pandas as pd
10
11 ALPHABET = "abcdefghijklmnopqrstuvwxyz"
12
13 class Problem:
14     def __init__(self, n, adj, graph_edge, default_n_community):
15         self.n = n
16         self.adj = adj
17         self.m = int(sum([len(x) for x in self.adj])/2)
18         self.default_n_community = default_n_community
19         self.graph_edge = graph_edge
20
21     def initial_population(self):
22         self.population = []
23         # We want to create a number of populations
24         for _ in range(self.population_size):
25
26             # For each one we will randomly put the nodes on a community
27
28             vertices = list(range(self.n))
29             random.shuffle(vertices)
30             individual_map = [None for _ in range(self.n)]
31
32             counter = 0
33
34             while len(vertices) != 0:
35                 if len(vertices) < int(self.n / self.default_n_community):
36                     selected = vertices[:]
37                     for i in selected:
38                         selected = vertices[:]
39                         for i in selected:
40                             individual_map[i] = counter
41                         counter += 1
42                     break
43                 else:
44                     selected = random.choices(vertices, k=int(self.n / self.default
45                     vertices = [e for e in vertices if e not in selected]
46                     for i in selected:
47                         individual_map[i] = counter
48                     counter += 1
49             self.population.append([individual_map, None])
50

```

```

51         self.population = sorted(self.population, key=lambda agent: self.fitness(a
52
53     def fitness(self, individual):
54         # Calculate the fitness of a chromosome based on the formula provided
55         Q = 0
56         for i in range(self.n):
57             for j in range(self.n):
58                 Q += (int(j in self.adj[i]) - ((len(self.adj[i]) * len(self.adj[j])
59         Q /= (2 * self.m)
60
61         individual[1] = Q
62         return Q
63
64     def selection(self):
65         random.shuffle(self.population)
66         self.parents = self.population[:self.tournament_size]
67         self.parents = sorted(self.parents, key=lambda agent: self.fitness(
68             agent), reverse=True)[:self.parents_size]
69
70     def breed(self, parent1, parent2):
71         # We will breed new children based on single point crossover
72         child1 = [None, None]
73         child2 = [None, None]
74         temp1 = []
75         temp2 = []
76         for i in range(len(parent1[0])):
77             choose = random.randint(1, 2)
78             if choose == 1:
79                 temp1.append(parent1[0][i])
80                 temp2.append(parent2[0][i])
81             else :
82                 temp1.append(parent2[0][i])
83                 temp2.append(parent1[0][i])
84         child1[0] = temp1
85         child2[0] = temp2
86         return child1, child2
87
88     def breed_offsprings(self):
89         self.children = []
90         # Select two parent for breeding two children
91         for _ in range(self.breed_rate):
92             random.shuffle(self.parents)
93             for i in range(int(len(self.parents)/2)):
94                 child1, child2 = self.breed(self.parents[i], self.parents[len(self.
95                 self.children.append(child1)
96                 self.children.append(child2)
97
98
99
100     def mutate(self, individual):
101         # we will randomly mutate some genes based on out mutation rate

```

```

102     individual_copy = [individual[0][:], None]
103     # Choosing wheter to mutate or not
104     if random.random() < self.mutation_rate:
105         # Changing a gene to another possible answer
106         gene = random.choice(range(self.n))
107         cluster = random.choice(list(set(individual[0])))
108         # We need to make sure we don't randomly select the first one :)
109         while cluster == individual[0][gene]:
110             cluster = random.choice(list(set(individual[0])))
111         individual_copy[0][gene] = cluster
112     self.fitness(individual_copy)
113     return individual_copy
114
115     def mutate_offsprings(self):
116         # Return the mutation which were formed
117         self.mutated_children = []
118         for individual in self.children:
119             self.mutated_children.append(self.mutate(individual))
120         return self.mutated_children
121
122     def replacement(self):
123         # We will consider the best of the children, mutated and parent and then se
124         self.mutated_children = sorted(
125             self.mutated_children, key=lambda agent: agent[1], reverse=True)
126         self.population = sorted(
127             self.parents, key=lambda agent: agent[1], reverse=True)
128
129         self.population = self.mutated_children[:-self.elite_size] + self.parents[:
130         self.population = sorted(
131             self.population, key=lambda agent: self.fitness(agent), reverse=
132
133     def evaluate(self):
134         pop_fitness = [agent[1] for agent in self.population]
135
136         return sum(pop_fitness), min(pop_fitness)
137
138     def graph_visulization(self):
139         fig2, ax2 = plt.subplots()
140         ax2.set_title('Communities')
141         G = nx.Graph()
142         G.add_edges_from(self.graph_edge)
143         color_map = [node for node in self.population[-1:][0]]
144         nx.draw_networkx(G, node_color = color_map[0])
145
146
147     def GA(self, population_size, tournament_size, parents_size, mutation_rate, eli
148         self.population_size = population_size
149         self.tournament_size = tournament_size
150         self.parents_size = parents_size
151         self.breed_rate = int(self.population_size/self.parents_size)
152         self.mutation_rate = mutation_rate

```

```

153     self.elite_size = elite_size
154     self.n_generations = n_generations
155
156     # Make the random first population
157     self.initial_population()
158     plotFitness = []
159     plotEpoch = []
160     plotPopFit = []
161     for epoch in range(self.n_generations):
162         # Select a random number of candidates for parents(tournament_size) the
163         self.selection()
164         # Create children from the chosen parents
165         self.breed_offsprings()
166         # Mutate some of them
167         self.mutate_offsprings()
168         # Select the best of them (population_size)
169         self.replacement()
170         eval_ = self.evaluate()
171         # Save the parameters so we can plot them later
172         plotFitness.append(eval_[1])
173         plotEpoch.append(epoch)
174         plotPopFit.append(eval_[0])
175         print("Epoch", epoch, ":\tPopulation total fitness:", eval_[0], "\tBest
176     # Create a graph of the process
177     fig, ax = plt.subplots()
178     ax.scatter(plotEpoch, plotPopFit, color = 'r')
179     ax.set_title('Population Total Fitness')
180     ax.set_xlabel('Generation')
181     ax.set_ylabel('Population Total Fitness')
182     sns.scatterplot(x=plotEpoch, y=plotPopFit)
183
184     fig1, ax1 = plt.subplots()
185     ax1.scatter(plotEpoch, plotFitness, color = 'b')
186     ax1.set_title('Population Best Fitness')
187     ax1.set_xlabel('Generation')
188     ax1.set_ylabel('Population Best Fitness')
189     sns.scatterplot(x=plotEpoch, y=plotFitness)
190     print("Final : ", ":\tPopulation total fitness:", plotPopFit[:-1], "\tBest
191     self.graph_visulization()
192

```

```

1 from google.colab import files
2 uploaded = files.upload()
3 f = open('sample dataset.txt', 'r')
4 lines = f.readlines()
5
6 n = int(lines[0])
7 lines = lines[1:]
8
9 adj = [[] for _ in range(n)]
10

```

```
10 graphn_edge = []
11
12 for edge in lines:
13     edge = edge.split()
14     graphn_edge.append(edge)
15
16     adj[int(edge[0]) - 1].append(int(edge[1]) - 1)
17     adj[int(edge[1]) - 1].append(int(edge[0]) - 1)
18
19 # Define problem parameters : Number of nodes, adjacency matrix, array of edges an
20 problem = Problem(n, adj, graphn_edge, 7)
21
22
23 problem.GA(population_size = 200, tournament_size = 160, parents_size = 120, mutati
24
```



Epoch 190 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 191 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 192 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 193 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 194 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 195 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 196 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 197 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 198 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Epoch 199 :	Population total fitness: 50.37475345167649	Best fitness: 0.0000000000000000
Final : :	Population total fitness: [-2.7707922419460846, -2.105111768573301]	

