ECE449

Lab 5

November 1st, 2023

- Genetic Algorithm: A heuristic search algorithm for optimizing an objective function
- Genes: building blocks of our solutions, string these genes together, we get a chromosome, which represents a full solution to our problem.
- Population: a diverse group of solutions competing to solve our problem.
- Fitness Function: objective function which is minimized or maximized

- Crossover: mixing the best parts of two solutions, hoping that their offspring inherits the best attributes of each parent
- Mutation: adding a pinch of randomness to our new solutions, ensuring that we explore a wide range of possibilities and avoid getting stuck

- Same Tipping Problem
- Use GA to optimize membership functions so that the solution are as close as possible to a dataset
- Clean the given dataset: clip rows with values smaller than 0 and greater than 1
- Note that your tip is between 0-30 in this lab. Other values in the dataset are between 0-1, so multiply by your max universe set value if needed.
- !pip install scikit-fuzzy
- !pip install EasyGA

Don't use .automf (3) for membership functions instead

```
These are the solutions we are trying to find

temperature['poor'] = fuzz.trimf(temperature.universe, [0, 0, 5])

temperature['average'] = fuzz.trimf(temperature.universe, [0, 5, 10])

temperature['good'] = fuzz.trimf(temperature.universe, [5, 5, 10])
```

- Or Any other membership function you want
- You should be able to access membership parameters this way.

- Define enough number of rules so that at least one rule is activated for every combination of memberships.
- Define your fitness function:
- Should return the error between the tip found by your fuzzy system and the tip in the dataset.
- \bullet error = $\sum abs$ (actual_tip predicted_tip)

For Instance:

```
def fitness(chromosome):
    food sim, service sim, tip sim = setup fuzzy_system(chromosome)
    total error = 0
    for index, row in train data.iterrows():
        inputs = {
            'temperature': row['temperature'],
            'flavor': row['flavor'],
            'portion size': row['portion size'],
            'attentiveness': row['attentiveness'],
            'friendliness': row['friendliness'],
            'speed': row['speed']
        actual tip = row['tip']
        predicted tip = execute fuzzy inference(food sim, service sim, tip sim, inputs)
        error = abs (actual tip - predicted tip)
        total error += error
   return total error
```

Change based on your needs and implementation.

- Define your GA:
- ga = EasyGA.GA()
- Define how the chromosomes should be generated.
- You can assert any constraints on the membership function parameters here.
- Remember that your membership functions should cover the whole universe set they are defined on, and should be within boundaries set by universe set.

```
ga.gene_impl = lambda: generate_chromosome()
```

```
ga.chromosome length =
ga.population size = 200 (whatever you like)
ga.target fitness type = 'min'
ga.generation goal = Number of iteration to run
ga.fitness function impl = fitness (your fitness function)
ga.evolve()
ga.print best chromosome()
```

Final model you get should outperform unoptimized model.