ICS674 Mini Project

Lambert Leong

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For my evolutionary computation mini project I implemented a genetic algorithm that matches an input string. Code implementations are written in Python 2.7.

1 Search Space

The search space consists of all alphabet characters, upper and lower case. This can be seen in Listing 1 where we randomly fill the search strings with letters by calling string.letters from Pythons string module.

```
Listing 1: Search space is all letter characters, upper and lower case
```

```
self.string = ''.join(random.choice(string.letters) for _ in xrange(length))
```

2 Variation Operator

Two variation operator are implemented which include crossover and mutation.

2.1 Crossover

To perform crossover we we randomly select two parent strings from the previous generation, seen in lines 4&5 of Listing 2. I randomly select an integer which corrisponds to the array index at which the parent string is split for each child to inherit, seen i line 8. Child 1 gets the char from the first half from parent 1 and the second half from parent 2. Child 2 getst the first half from parent 2 and second half from parent 1.

Listing 2: Crossover Function

```
def crossover(individuals):
    offspring = []

for _ in xrange((population - len(individuals))/2):
    parent1 = random.choice(individuals)
    parent2 = random.choice(individuals)
    child1 = Individual(in_str_len)
    child2 = Individual(in_str_len)
    split = random.randint(0, in_str_len)
```

2.2 Mutation

Mutation occurs in the for of switching out a character in a search string with a random letter. All strings in the population are suseptible to mutation and multiple mutation can occur in a individual string. The mutation rate is 5% as indicated in line 4 of Listing 3.

Listing 3: Mutation Function

3 Selection Operator

```
def selection(individuals):
    individuals = sorted(individuals, key=lambda individual:
        individual.fitness, reverse=True)

max_fit.append(max(individuals, key=lambda individual:
        individual.fitness).fitness)

min_fit.append(min(individuals, key=lambda individual:
        individual.fitness).fitness)

avg_fit.append(float(sum(i.fitness for i in
        individuals)//len(individuals)))

individuals = individuals[:int(0.2*len(individuals))]

return individuals
```

4 Termination Criterion

```
for generation in xrange(generations):
generation_list.append(generation)
```

```
individuals = fitness(individuals)
individuals = selection(individuals)
individuals = crossover(individuals)
individuals = mutation(individuals)
if any(individual.fitness >= 100 for individual in individuals):
found = True
break
```

5 Objective Fuction

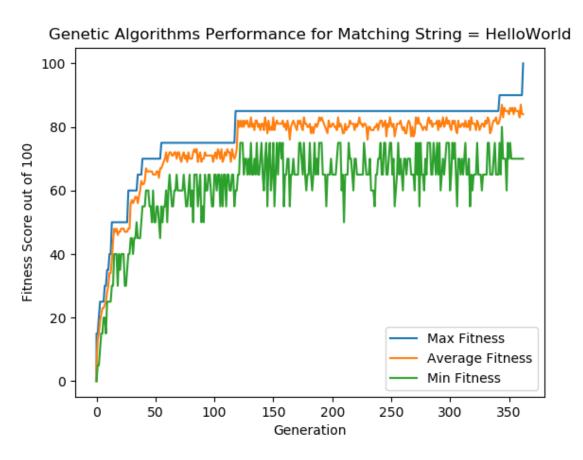


Figure 1: Max, average, and min fitness values of each indvidual string for each generation

Listing 4: Fitness Function

```
def fitness(individuals):
          for individual in individuals:
                 total = len(in_str)*2
3
                 score = 0
                 for i, letter in enumerate(individual.string):
                                if in_str[i] == letter:
                                        score += 1
                 compare_str = in_str
                 for a_char in individual.string:
                         for i, in_char in enumerate(compare_str):
10
                                 if a_char == in_char:
11
                                        score += 1
12
                                        compare_str =
13
                                            compare_str[:i]+compare_str[i+1:]
14
                  individual.fitness = int((float(score)/float(total))*100)
15
          return individuals
```