

EXPERIMENT NO . 5

Genetic Algorithm

i) Without Using Inbuilt Library

Code:-

```
from random import randint
def selection(li):
    dec = list(map(lambda x : int(x, 2), li))
    fit = list(map(lambda x : x*x*x, dec))
    s = sum(fit)
    prob = list(map(lambda x : round(x/s, 3), fit))
    avg = s/n
    exe = list(map(lambda x : round(x/avg, 3), fit))
    ac = list(map(lambda x : round(x), exe))
    return dec, fit, prob, exe, ac
def pp(li, ac, n):
    co = []
    temp = []
    index = []
    for i in range(n):
        if ac[i] == 1:
            co.append(li[i])
        elif ac[i] >= 2:
            for j in range(ac[i] - 1):
                temp.append(li[i])
            co.append(li[i])
        elif ac[i] == 0 and len(temp) != 0:
            co.append(temp[0])
            temp.pop(0)
        elif ac[i] == 0 and len(temp) == 0:
            index.append(i)
    if len(index) != 0 and len(temp) != 0:
        for i in index:
            co.insert(i, temp[0])
            temp.pop(0)
    elif len(index) != 0 and len(temp) == 0:
        co.insert(i, li[i])
    return co
def cr(x):
    s = 0
    for i in x:
        if i == '1':
            s = s + 1
    return s
def crossing(li, n):
    crossed = []
    for i in range(0, n, 2):
        temp1 = li[i]
```

```

        j = i + 1
        temp2 = li[j]
        crosspoint = cr(temp1)
        print("The crosspoint for pair " + str(i) + " is " + str(crosspoint))
        temp3 = temp1[crosspoint:]
        temp4 = temp2[crosspoint:]
        temp1 = temp1[0 : crosspoint] + temp4
        temp2 = temp2[0 : crosspoint] + temp3
        crossed.append(temp1)
        crossed.append(temp2)
    return crossed
def mutation(li, n):
    mut = []
    for i in li:
        j = randint(0, n - 1)
        print("For pair " + str(i) + ", the bit that will be changed is "
+str(j))
        if i[j] == '1':
            i = i[0 : j] + '0' + i[j + 1 : ]
            mut.append(i)
        elif i[j] == '0':
            i = i[0 : j] + '1' + i[j + 1 : ]
            mut.append(i)
    return mut
n = int(input("Enter number of samples: "))
sam = []
for i in range(n):
    sam.append(input("Enter gene: "))
m = int(input("Enter number of generations to be computed: "))
crossed = sam.copy()
for i in range(m):
    dec, fit, prob, exe, ac = selection(crossed)
    s = sum(ac)
    if s < n:
        maxi = max(ac)
        k = ac.index(maxi - 1)
        ac[k] += 1
    if s > n:
        maxi = max(ac)
        k = ac.index(maxi)
        ac[k] -= 1
    print("\n----- GENERATION ", i, "-----")
    print("Initial Population\tX Value\tFitness Value\tProbability\tExpected\nCount\tActual Count")
    for j in range(n):
        print(crossed[j], "\t\t", dec[j], "\t\t", fit[j], "\t",
prob[j], "\t\t", exe[j], "\t\t\t", ac[j])

```

```

co = pp(crossed, ac, n)
print("\nSelected Genes for Crossover - \n", co)
crossed = crossing(co, n)
print("\nCrossover - \n", crossed)
crossed = mutation(crossed, n)
print("\nMutated - \n", crossed)
print("\nGENERATION ", (m + 1), " - ", crossed)

```

Output:-

```

PS D:\AI pracs> python genetic.py
Enter number of samples: 4
Enter gene: 01010
Enter gene: 11010
Enter gene: 11111
Enter gene: 01010
Enter number of generations to be computed: 2

----- GENERATION 0 -----
Initial Population   X Value   Fitness Value   Probability   Expected Count   Actual Count
01010               10         1000           0.02         0.081           0
11010               26        17576          0.356        1.424           2
11111               31        29791          0.603        2.414           2
01010               10         1000           0.02         0.081           0

Selected Genes for Crossover -
['11111', '11010', '11111', '11010']
The crosspoint for pair 0 is 5
The crosspoint for pair 2 is 5

Crossover -
['11111', '11010', '11111', '11010']
For pair 11111, the bit that will be changed is 3
For pair 11010, the bit that will be changed is 2
For pair 11111, the bit that will be changed is 1
For pair 11010, the bit that will be changed is 1

Mutated -
['11101', '11110', '10111', '10010']

----- GENERATION 1 -----
Initial Population   X Value   Fitness Value   Probability   Expected Count   Actual Count
11101               29        24389          0.351        1.406           1
11110               30        27000          0.389        1.556           2
10111               23        12167          0.175        0.701           1
10010               18         5832          0.084        0.336           0

Selected Genes for Crossover -
['11101', '11110', '10111', '11110']
The crosspoint for pair 0 is 4
The crosspoint for pair 2 is 4

Crossover -
['11100', '11111', '10110', '11111']
For pair 11100, the bit that will be changed is 0
For pair 11111, the bit that will be changed is 1
For pair 10110, the bit that will be changed is 0
For pair 11111, the bit that will be changed is 2

Mutated -
['01100', '10111', '00110', '11011']

GENERATION 3 - ['01100', '10111', '00110', '11011']

```

ii) Using Inbuilt Library

Code:-

```
import numpy
class ga:
    def cal_pop_fitness(equation_inputs, pop):
        fitness = numpy.sum(pop*equation_inputs, axis=1)
        return fitness
    def select_mating_pool(pop, fitness, num_parents):
        parents = numpy.empty((num_parents, pop.shape[1]))
        for parent_num in range(num_parents):
            max_fitness_idx = numpy.where(fitness == numpy.max(fitness))
            max_fitness_idx = max_fitness_idx[0][0]
            parents[parent_num, :] = pop[max_fitness_idx, :]
            fitness[max_fitness_idx] = -99999999999
        return parents
    def crossover(parents, offspring_size):
        offspring = numpy.empty(offspring_size)
        crossover_point = numpy.uint8(offspring_size[1]/2)
        for k in range(offspring_size[0]):
            parent1_idx = k%parents.shape[0]
            parent2_idx = (k+1)%parents.shape[0]
            offspring[k, 0:crossover_point] =
parents[parent1_idx,0:crossover_point]
            offspring[k, crossover_point:] =
parents[parent2_idx,crossover_point:]
        return offspring
    def mutation(offspring_crossover):
        for idx in range(offspring_crossover.shape[0]):
            random_value = numpy.random.uniform(-1.0, 1.0, 1)
            offspring_crossover[idx, 4] = offspring_crossover[idx, 4] +
random_value
        return offspring_crossover
equation_inputs = [4,-2,3.5,5,-11,-4.7]
num_weights = 6
sol_per_pop = 8
num_parents_mating = 4
pop_size = (sol_per_pop,num_weights)
new_population = numpy.random.uniform(low=-4.0, high=4.0, size=pop_size)
print(new_population)
num_generations = 5
for generation in range(num_generations):
    print("Generation : ", generation)
    fitness = ga.cal_pop_fitness(equation_inputs, new_population)
    print("fitness : ", fitness)
    parents = ga.select_mating_pool(new_population,
fitness,num_parents_mating)
    print("parents : ", parents)
```

```

    offspring_crossover = ga.crossover(parents,offspring_size=(pop_size[0]-
parents.shape[0],num_weights))
    offspring_mutation = ga.mutation(offspring_crossover)
    print("mutation : ", offspring_mutation)
    new_population[0:parents.shape[0], :] = parents
    new_population[parents.shape[0]:, :] = offspring_mutation
    print("new population : ", new_population)
    print("Best result : ",
    numpy.max(numpy.sum(new_population*equation_inputs, axis=1)))
    fitness = ga.cal_pop_fitness(equation_inputs, new_population)
    best_match_idx = numpy.where(fitness == numpy.max(fitness))
print("Best solution : ", new_population[best_match_idx, :])
print("Best solution fitness : ", fitness[best_match_idx])

```

Output:-

```

PS D:\AI project\python\genetic.py
[[ -0.3309084  -2.13315511 -2.3283651 -2.3096295  -0.8189706  -2.21131976
 -0.94028765  3.54951771  2.0341086  -2.0266401  2.4380706  -0.7302862
 -1.1348758  -2.82296254 -2.29191793 -2.0001649  1.8094854  1.2585088
 -0.0021106  -0.87892777 -0.3959830  0.3039982  1.5332146  1.04229325
 -1.71256754  0.86148456 -3.81787088  2.66211143 -3.81853085 -2.48312669
 0.0859412  -1.87953286  1.13250895 -1.5039302  6.8137796 -0.53206609
 -2.11030529  1.8154508  1.13892322 -3.5598302 -2.7108864 -2.49766236
 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]
new_population : 0
Fitness : [ 20.40720732 45.67896291 13.12018769  2.41115469 44.80384625
 3.58698902 62.95099675]
parents : [[ -1.71256754  0.86148456 -3.81787088  2.66211143 -3.81853085 -2.48312669]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]
 [-2.11030529  1.8154508  1.13892322 -3.5598302 -2.7108864 -2.49766236]
 [-1.1348758  -2.82296254 -2.29191793 -2.0001649  1.8094854  1.2585088]
 [ 1.5405758  2.81990254 -2.29191793 -2.0001649  1.8094854  1.2585088]]
DEPRECATED: practice2.py:26: DeprecationWarning: Conversion of an array with ndim > 0 to a scalar is deprecated, and will error in future. Ensure you extract a single element from your array before performing this operation. (Deprecated NumPy 1.20.)
offspring_crossover[idx, :] = offspring_crossover[idx, :] + random.value
new_population : [[ -1.71256754  0.86148456 -3.81787088  2.66211143 -3.81853085 -2.48312669]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]
 [-2.11030529  1.8154508  1.13892322 -3.5598302 -2.7108864 -2.49766236]
 [-1.1348758  -2.82296254 -2.29191793 -2.0001649  1.8094854  1.2585088]
 [-1.71256754  0.86148456 -3.81787088  2.66211143 -3.81853085 -2.48312669]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]
 [-2.11030529  1.8154508  1.13892322 -3.5598302 -2.7108864 -2.49766236]
 [-1.1348758  -2.82296254 -2.29191793 -2.0001649  1.8094854  1.2585088]]
Best result : 62.950996750444

```

```

Generation : 1
Fitness : [44.96130455 22.55202293 15.37389268 13.12018769 -2.41115469 27.93473036
 3.58698902 62.95099675]
parents : [[ -1.13485758 -2.02190254 -2.29191793  2.66211143 -4.22609838 -2.48312669]
 [ -1.71256754  0.86148456 -3.81787088  2.66211143 -3.81853085 -2.48312669]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]]
mutation : [[ -1.13485758 -2.02190254 -2.29191793  2.66211143 -3.45557446 -2.48312669]
 [-1.71256754  0.86148456 -3.81787088 -3.5598302 -2.41689029 -2.49766236]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.18818886  1.48217088]
 [ 3.57545542  2.68210442 -0.6054428  2.66211143 -4.02457721 -2.48312669]]
new_population : [[ -1.13485758 -2.02190254 -2.29191793  2.66211143 -4.22609838 -2.48312669]
 [-1.71256754  0.86148456 -3.81787088  2.66211143 -3.81853085 -2.48312669]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.34129309  1.48217088]
 [-1.13485758 -2.02190254 -2.29191793  2.66211143 -3.45557446 -2.48312669]
 [-1.71256754  0.86148456 -3.81787088 -3.5598302 -2.41689029 -2.49766236]
 [ 3.57545542  2.68210442 -0.6054428  1.58908781 -1.18818886  1.48217088]
 [ 3.57545542  2.68210442 -0.6054428  2.66211143 -4.02457721 -2.48312669]]
Best result : 76.070164988664

```



```

Generation : 2
fitness : [62.95099675 44.96130455 27.93473036 22.55202293 54.47523371 -1.41013206
20.86787643 76.07016498]
parents : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.02457721 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -4.22609838 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.45557446 -2.48312669]
[-1.71256754 0.86148456 -3.81787088 2.66211143 -3.81053085 -2.48312669]]
mutation : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.15548491 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.39007318 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.55270573 -2.48312669]
[-1.71256754 0.86148456 -3.81787088 2.66211143 -4.59928874 -2.48312669]]
new population : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.02457721 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -4.22609838 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.45557446 -2.48312669]
[-1.71256754 0.86148456 -3.81787088 2.66211143 -3.81053085 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.15548491 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.39007318 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.55270573 -2.48312669]
[-1.71256754 0.86148456 -3.81787088 2.66211143 -4.59928874 -2.48312669]]
Best result : 77.51014961696165

```

```

Generation : 3
fitness : [76.07016498 62.95099675 54.47523371 44.96130455 77.51014961 53.84271956
55.54367767 53.63764137]
parents : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.15548491 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.02457721 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -4.22609838 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.55270573 -2.48312669]]
mutation : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.74575549 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.04797017 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -4.38725664 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.23252107 -2.48312669]]
new population : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.15548491 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.02457721 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -4.22609838 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.55270573 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.74575549 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.04797017 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -4.38725664 -2.48312669]
[-1.13485758 -2.02190254 -2.29191793 2.66211143 -3.23252107 -2.48312669]]
Best result : 87.32748757431156

```

```

Generation : 4
fitness : [77.51014961 76.07016498 62.95099675 55.54367767 84.00312603 87.32748757
64.72373761 52.02164632]
parents : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.04797017 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.74575549 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.15548491 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.02457721 -2.48312669]]
mutation : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.73266178 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -3.59006272 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.96592221 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.22776043 -2.48312669]]
new population : [[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.04797017 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.74575549 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.15548491 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.02457721 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.73266178 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -3.59006272 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -4.96592221 -2.48312669]
[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.22776043 -2.48312669]]
Best result : 94.85909525344171
Best solution : [[[ 3.57545542 2.68210442 -0.6054428 2.66211143 -5.73266178
-2.48312669]]]
Best solution fitness : [94.85909525]

```

HILLCLIMBING

Implementation:-

```
exp4.py > randomSolution
1 import random
2 def randomSolution(tsp):
3     cities = list(range(len(tsp)))
4     solution = []
5     for i in range(len(tsp)):
6         randomCity = cities[random.randint(0, len(cities) - 1)]
7         solution.append(randomCity)
8         cities.remove(randomCity)
9     return solution
10
11 def routeLength(tsp, solution):
12     routeLength = 0
13     for i in range(len(solution)):
14         routeLength += tsp[solution[i - 1]][solution[i]]
15     return routeLength
16
17 def getNeighbours(solution):
18     neighbours = []
19     for i in range(len(solution)):
20         for j in range(i + 1, len(solution)):
21             neighbour = solution.copy()
22             neighbour[i] = solution[j]
23             neighbour[j] = solution[i]
24             neighbours.append(neighbour)
25     return neighbours
26
27 def getBestNeighbour(tsp, neighbours):
28     bestRouteLength = routeLength(tsp, neighbours[0])
29     bestNeighbour = neighbours[0]
30     for neighbour in neighbours:
31         currentRouteLength = routeLength(tsp, neighbour)
32         if currentRouteLength < bestRouteLength:
33             bestRouteLength = currentRouteLength
34             bestNeighbour = neighbour
35     return bestNeighbour, bestRouteLength
36
37 def hillClimbing(tsp):
38     currentSolution = randomSolution(tsp)
39     currentRouteLength = routeLength(tsp, currentSolution)
40
41     neighbours = getNeighbours(currentSolution)
42     bestNeighbour, bestNeighbourRouteLength = getBestNeighbour(tsp, neighbours)
43     while bestNeighbourRouteLength < currentRouteLength:
44         currentSolution = bestNeighbour
45         currentRouteLength = bestNeighbourRouteLength
46         neighbours = getNeighbours(currentSolution)
47         bestNeighbour, bestNeighbourRouteLength = getBestNeighbour(tsp, neighbours)
48     return currentSolution, currentRouteLength
49
50 def main():
51     tsp = [
52         [0, 100, 700, 50],
53         [100, 0, 350, 1200],
54         [700, 350, 0, 400],
55         [50, 1200, 400, 0] ]
56     print(hillClimbing(tsp))
57     if __name__ == "__main__":
58         main()
```

Output:-

[3,2,1,0] 880