

## Lab Experiment – 7

### GEA [Gene Expression Algorithm]

Lab - 7 GEA.					
Gene expression Algorithm: 6 Main Phases.					
	Initialisation	Fitness Assignment	Selection	Cross over	Gene expression
1)	select encoding Technique: 0 - 31				
	use chromosome of fixed length with terminals (variables, constants and functions (+, -, *, /)).				
2)	Initial Population:				
S.NO	Initial chromosome	Phenotype	Value	Fitness	P
1.	$\alpha \alpha$	$\alpha^2$	12	144	0.1247
2.	$+\alpha \alpha$	$2\alpha$	25	625	0.541
3.	$\alpha$	$\alpha$	5	25	0.0216
4.	$-\alpha$	$\alpha - 2$	19	361	0.8125
$\Sigma P(\alpha) = 1155$			Actual cost	Exp. cost	
$Avg = 288.75$			1	0.5	
			2	2.1	
			0	0.08	
			1	1.25	
(x) result					
IPR					
DRP					
PRP					
ODP					

### 3) Selection of Mating Pool.

S.No	Selected chromosome	crossover point	Offspring	Phenotype
1	*xx	2	*x+	x*(x+...)
2	+xx	1	+x*	2x
3	+x*	3	+x-	x+(x...)
4	-x2	1	+x2	x+2

$\alpha$ values	Fitness
13	169
24	576
24	729
17	289

### 4) Crossover:

Perform crossover randomly chosen gene positions (not raw bits).

Max fitness after crossover = 729.

### 5) Mutation:

S.No	Offspring before mutation	Mutation Applied	Offspring after mutation	Phenotype
1	*x+	+ → -	+x-	x*(x...)
2	+xx	None	+x*	2x
3	+x-	- → *	+x*	x+x*
4	+x2	None	+x2	x+2

$\alpha$ value	Fitness $f(x)$
29	841
24	576
27	729
20	400

by gene expression and evaluation:  
Decode each genotype  $\rightarrow$  phenotype.

$$\Sigma f(x) = 841 + 576 + 729 + 400 = 2546$$

$$\text{Avg} = 636.5$$

$$\text{Max} = 841$$

1) Iterate until convergence:

Repeat - stop 3-6 until fitness improvement is negligible or generation limit has reached.

Pseudocode / Algorithm:

- \* Start
- \* Define Fitness function.
- \* Define parameters
- \* Select mating pool.
- \* Mutation after mating.
- \* Gene expression and evaluation.
- \* Iterate.
- \* Output best value.

Output:

Genes: [29.53, 29.82, 29.34, 28.57, 15.09,  
21.83, 23.43, 30.81, 22.51, 26.22].

$$x : 26.37$$

$$f(x) : 695.45$$

## Output:

```
. Enter 4 chromosomes (each of 5 bits, e.g., '10101'):
Chromosome 1: 01101
Chromosome 2: 11000
Chromosome 3: 11011
Chromosome 4: 10001
Generation 1: Best Fitness = 729, Best x = 27
Generation 2: Best Fitness = 729, Best x = 27
Generation 3: Best Fitness = 729, Best x = 27
Generation 4: Best Fitness = 729, Best x = 27
Generation 5: Best Fitness = 729, Best x = 27
Generation 6: Best Fitness = 729, Best x = 27
Generation 7: Best Fitness = 729, Best x = 27
Generation 8: Best Fitness = 729, Best x = 27
Generation 9: Best Fitness = 729, Best x = 27
Generation 10: Best Fitness = 729, Best x = 27
Generation 11: Best Fitness = 729, Best x = 27
Generation 12: Best Fitness = 729, Best x = 27
Generation 13: Best Fitness = 729, Best x = 27
Generation 14: Best Fitness = 729, Best x = 27
Generation 15: Best Fitness = 729, Best x = 27
Generation 16: Best Fitness = 729, Best x = 27
Generation 17: Best Fitness = 729, Best x = 27
Generation 18: Best Fitness = 729, Best x = 27
Generation 19: Best Fitness = 729, Best x = 27
Generation 20: Best Fitness = 729, Best x = 27
...
Best solution found:
Chromosome: 11011
x = 27
f(x) = 729
```

## Code:

```
import random

# Define the function to maximize

def fitness_function(x):

    return x ** 2

# Convert binary string to integer

def decode(chromosome):

    return int(chromosome, 2)

# Evaluate fitness for the entire population

def evaluate_population(population):

    return [fitness_function(decode(individual)) for individual in population]

# Selection: Roulette Wheel
```

```

def select(population, fitnesses):
    total_fitness = sum(fitnesses)
    if total_fitness == 0:
        return random.choice(population)
    pick = random.uniform(0, total_fitness)
    current = 0
    for individual, fitness in zip(population, fitnesses):
        current += fitness
        if current > pick:
            return individual

# Crossover: Single-point crossover
def crossover(parent1, parent2):
    if random.random() < CROSSOVER_RATE:
        point = random.randint(1, CHROMOSOME_LENGTH - 1)
        return (parent1[:point] + parent2[point:], parent2[:point] + parent1[point:])
    return parent1, parent2

# Mutation: Flip random bit
def mutate(chromosome):
    new_chromosome = ""
    for bit in chromosome:
        if random.random() < MUTATION_RATE:
            new_chromosome += '0' if bit == '1' else '1'
        else:
            new_chromosome += bit
    return new_chromosome

```

```

# Get user input for initial population

def get_initial_population(size, length):

    population = []

    print(f"Enter {size} chromosomes (each of {length} bits, e.g., '10101'):")

    while len(population) < size:

        chrom = input(f"Chromosome {len(population)+1}: ").strip()

        if len(chrom) == length and all(bit in '01' for bit in chrom):

            population.append(chrom)

        else:

            print(f"Invalid input. Please enter a {length}-bit binary string.")

    return population


# Run the Genetic Algorithm

def genetic_algorithm():

    population = get_initial_population(POPULATION_SIZE, CHROMOSOME_LENGTH)

    best_solution = None

    best_fitness = float('-inf')

    for generation in range(GENERATIONS):

        fitnesses = evaluate_population(population)

        # Update best solution

        for i, individual in enumerate(population):

            if fitnesses[i] > best_fitness:

                best_fitness = fitnesses[i]

                best_solution = individual

        print(f"Generation {generation + 1}: Best Fitness = {best_fitness}, Best x = {decode(best_solution)}")

        new_population = []

        while len(new_population) < POPULATION_SIZE:

```

```
parent1 = select(population, fitnesses)

parent2 = select(population, fitnesses)

offspring1, offspring2 = crossover(parent1, parent2)

offspring1 = mutate(offspring1)

offspring2 = mutate(offspring2)

new_population.extend([offspring1, offspring2])

population = new_population[:POPULATION_SIZE]

print("\nBest solution found:")

print(f"Chromosome: {best_solution}")

print(f"x = {decode(best_solution)}")

print(f"f(x) = {fitness_function(decode(best_solution))}")

# Parameters

POPULATION_SIZE = 4

CHROMOSOME_LENGTH = 5

MUTATION_RATE = 0.01

CROSSOVER_RATE = 0.8

GENERATIONS = 20

# Run it

if __name__ == "__main__":
    genetic_algorithm()
```