

# Optimization via Gene Expression Algorithms:

```
import random
```

```
def objective_function(x):
```

```
    return x ** 2
```

```
POP_SIZE = 20
```

```
GENS = 20
```

```
GENE_LENGTH = 10
```

```
CROSSOVER_RATE = 0.8
```

```
MUTATION_RATE = 0.1
```

```
BOUNDS = [-10, 10]
```

```
def create_population():
```

```
    return [[random.uniform(BOUNDS[0], BOUNDS[1]) for _ in range(GENE_LENGTH)] for _ in range(POP_SIZE)]
```

```
def gene_expression(gene):
```

```
    return sum(gene) / len(gene)
```

```
def evaluate(population):
```

```
    expressed = [gene_expression(g) for g in population]
```

```
    return [objective_function(x) for x in expressed], expressed
```

```
def select(population, fitness):
```

```
    i, j = random.sample(range(len(population)), 2)
```

```
    return population[i] if fitness[i] > fitness[j] else population[j]
```

```
def crossover(parent1, parent2):
```

```
    if random.random() < CROSSOVER_RATE:
```

```

    point = random.randint(1, GENE_LENGTH - 1)

    return parent1[:point] + parent2[point:]

return parent1[:]

```

```
def mutate(gene):
```

```
    return [g + random.uniform(-1, 1) if random.random() < MUTATION_RATE else g for g in gene]
```

```
def gene_expression_algorithm():
```

```
    population = create_population()
```

```
    for gen in range(GENS):
```

```
        fitness, expressed = evaluate(population)
```

```
        new_population = []
```

```
        for _ in range(POP_SIZE):
```

```
            parent1 = select(population, fitness)
```

```
            parent2 = select(population, fitness)
```

```
            child = crossover(parent1, parent2)
```

```
            child = mutate(child)
```

```
            new_population.append(child)
```

```
        population = new_population
```

```
        best_idx = fitness.index(max(fitness))
```

```
        best_x = expressed[best_idx]
```

```
        best_fit = fitness[best_idx]
```

```
        print(f"Gen {gen+1}: Best x = {best_x:.4f}, f(x) = {best_fit:.4f}")
```

```
    return best_x, best_fit
```

```
best_x, best_val = gene_expression_algorithm()
```

```
print("\nBest solution found:")
```

```
print(f"x = {best_x:.4f}, f(x) = {best_val:.4f}")
```

## OUTPUT:

➡ Gen 1: Best  $x = -4.3850$ ,  $f(x) = 19.2279$   
Gen 2: Best  $x = 3.4352$ ,  $f(x) = 11.8008$   
Gen 3: Best  $x = 3.0554$ ,  $f(x) = 9.3358$   
Gen 4: Best  $x = 3.6876$ ,  $f(x) = 13.5987$   
Gen 5: Best  $x = 5.4107$ ,  $f(x) = 29.2759$   
Gen 6: Best  $x = 3.9887$ ,  $f(x) = 15.9100$   
Gen 7: Best  $x = 5.4586$ ,  $f(x) = 29.7961$   
Gen 8: Best  $x = 5.5366$ ,  $f(x) = 30.6537$   
Gen 9: Best  $x = 5.9909$ ,  $f(x) = 35.8907$   
Gen 10: Best  $x = 6.0167$ ,  $f(x) = 36.2010$   
Gen 11: Best  $x = 6.0452$ ,  $f(x) = 36.5444$   
Gen 12: Best  $x = 6.0391$ ,  $f(x) = 36.4712$   
Gen 13: Best  $x = 6.1264$ ,  $f(x) = 37.5330$   
Gen 14: Best  $x = 6.1264$ ,  $f(x) = 37.5330$   
Gen 15: Best  $x = 6.2923$ ,  $f(x) = 39.5931$   
Gen 16: Best  $x = 6.4059$ ,  $f(x) = 41.0352$   
Gen 17: Best  $x = 6.5065$ ,  $f(x) = 42.3346$   
Gen 18: Best  $x = 6.5738$ ,  $f(x) = 43.2153$   
Gen 19: Best  $x = 6.6635$ ,  $f(x) = 44.4028$   
Gen 20: Best  $x = 6.6529$ ,  $f(x) = 44.2616$

Best solution found:

$x = 6.6529$ ,  $f(x) = 44.2616$