## GENETIC OPTIMIZATION ALGORITHM:

```
import numpy as np
def objective_function(x):
return x ** 2
population_size = 100
num_generations = 50
mutation_rate = 0.1
crossover_rate = 0.7
value_range = (-10, 10)
def initialize_population(size, value_range):
return np.random.uniform(value_range[0], value_range[1], size)
def evaluate_fitness(population):
return np.array([objective_function(x) for x in population])
def selection(population, fitness):
   probabilities = fitness / fitness.sum()
return population[np.random.choice(len(population), size=2, p=probabilities)]
def crossover(parent1, parent2):
   if np.random.rand() < crossover_rate:</pre>
        return (parent1 + parent2) / 2
  return parent1 if np.random.rand() < 0.5 else parent2
def mutate(individual, mutation_rate, value_range):
   if np.random.rand() < mutation_rate:</pre>
        return np.random.uniform(value_range[0], value_range[1])
   return individual
def genetic_algorithm():
    population = initialize_population(population_size, value_range)
    best_solution = None
   best_fitness = -np.inf
   for generation in range(num_generations):
       fitness = evaluate_fitness(population)
        current_best_index = np.argmax(fitness)
        if fitness[current_best_index] > best_fitness:
            best_fitness = fitness[current_best_index]
            best_solution = population[current_best_index]
     new_population = []
```

```
for _ in range(population_size):
    parent1, parent2 = selection(population, fitness)
    offspring = crossover(parent1, parent2)
    offspring = mutate(offspring, mutation_rate, value_range)
    new_population.append(offspring)

population = np.array(new_population)

return best_solution, best_fitness

best_solution, best_fitness = genetic_algorithm()

print(f"Best solution found: x = {best_solution:.2f}")
print(f"Maximum value of f(x) = x^2: f(x) = {best_fitness:.2f}")
```

## OUTPUT:

```
Best solution found: x = -10.00
Maximum value of f(x) = x^2: f(x) = 99.92
```