```
import numpy as np
population size = 100
generations = 50
mutation rate = 0.1
crossover rate = 0.7
range min = -10
range max = 10
def fitness(x):
   return x**2
def initialize population(size, range min, range max):
    return np.random.uniform(range min, range max, size)
def select parents(population):
    fitness values = fitness(population)
    total fitness = np.sum(fitness values)
    selection probs = fitness values / total fitness
    selected indices = np.random.choice(len(population), size=2,
p=selection probs)
    return population[selected indices]
def crossover(parent1, parent2):
   if np.random.rand() < crossover rate:</pre>
        return (parent1 + parent2) / 2 # Simple average crossover
    return parent1 # No crossover occurs
def mutate(offspring, range min, range max):
    if np.random.rand() < mutation rate:</pre>
        return np.random.uniform(range min, range max)
    return offspring
def genetic algorithm():
```

```
population = initialize_population(population_size, range_min,
range_max)

for generation in range(generations):
    new_population = []

# Step 2: Create new population
for _ in range(population_size):
    parent1, parent2 = select_parents(population)
        child = crossover(parent1, parent2)
        child = mutate(child, range_min, range_max)
        new_population.append(child)

population = np.array(new_population)

# Return the best solution found
best_solution = population[np.argmax(fitness(population))]
return best_solution, fitness(best_solution)

# Running the genetic algorithm
best_x, best_fitness = genetic_algorithm()
print(f"Best x: {best_x}, Maximum f(x): {best_fitness}")
```

Output:

 \rightarrow Best x: 9.677365496658311, Maximum f(x): 93.65140295591276