Lab 2 Gene Expression

```
import random
import math
# Target function to approximate
def target function(x):
   return x^{**}2 + x + 1
# Terminal and function sets
TERMINALS = ['x']
FUNCTIONS = ['+', '-', '*', '/']
# Parameters
POP SIZE = 100
MUTATION_RATE = 0.1
TOURNAMENT_SIZE = 3
GENERATIONS = 100
# -----
# Utility functions
# -----
def safe_div(a, b):
   """Safe division to avoid divide-by-zero."""
       return a / b if b != 0 else 1
   except:
       return 1
def random_gene(max_depth=3):
    """Randomly create an expression (gene) of limited depth."""
    if max_depth == 0 or (max_depth > 1 and random.random() < 0.3):</pre>
       return random.choice(TERMINALS)
   else:
       op = random.choice(FUNCTIONS)
       left = random_gene(max_depth - 1)
       right = random_gene(max_depth - 1)
       return f"({left}{op}{right})"
def evaluate_expression(expr, x):
   """Safely evaluate the evolved expression."""
   try:
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   else:
       op = random.choice(FUNCTIONS)
        left = random_gene(max_depth - 1)
        right = random_gene(max_depth - 1)
        return f"({left}{op}{right})"
def evaluate expression(expr, x):
   """Safely evaluate the evolved expression."""
   try:
        # Define the environment so eval can use math and safe_div properly
       env = {"x": x, "math": math, "safe_div": safe_div}
       # Replace '/' with safe div(a,b)
       expr_safe = expr.replace("/", "safe_div")
       result = eval(expr_safe, env)
       if callable(result): # in case a function is accidentally returned
            return 0
       return float(result)
   except Exception:
        return 0
def fitness(expr):
   """Mean squared error against target function."""
   xs = [i for i in range(-10, 11)]
   errors = []
   for x in xs:
       y true = target_function(x)
       y_pred = evaluate_expression(expr, x)
       errors.append((y_true - y_pred)**2)
   mse = sum(errors) / len(errors)
   return mse
# Genetic operators
# -----
def mutate(expr):
   """Randomly mutate part of the expression."""
   expr_list = list(expr)
   for i in range(len(expr_list)):
        if random.random() < MUTATION_RATE:</pre>
            expr_list[i] = random.choice(['x', '+', '-', '*', '/'])
   return ''.join(expr_list)
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            expr_list[i] = random.choice(['x', '+', '-', '*', '/'])
    return ''.join(expr_list)
def crossover(parent1, parent2):
    """Single-point crossover."""
    if len(parent1) < 2 or len(parent2) < 2:</pre>
        return parent1, parent2
    cut1 = random.randint(1, len(parent1) - 1)
    cut2 = random.randint(1, len(parent2) - 1)
    child1 = parent1[:cut1] + parent2[cut2:]
    child2 = parent2[:cut2] + parent1[cut1:]
    return child1, child2
def tournament selection(pop, fits):
    """Select the best of k random individuals."""
    chosen = random.sample(list(zip(pop, fits)), TOURNAMENT_SIZE)
    return min(chosen, key=lambda x: x[1])[0]
# Main GEP loop
def gene expression optimization():
    population = [random_gene() for _ in range(POP_SIZE)]
    for gen in range(GENERATIONS):
        fitnesses = [fitness(expr) for expr in population]
        best_idx = min(range(len(fitnesses)), key=lambda i: fitnesses[i])
        best_expr = population[best_idx]
        best_fit = fitnesses[best_idx]
        if gen % 10 == 0 or gen == GENERATIONS - 1:
            print(f"Generation {gen:3d} | Best Expr: {best_expr} | Fitness = {best_fit:.6f}")
        new_pop = [best_expr] # elitism
        while len(new pop) < POP SIZE:
            p1 = tournament_selection(population, fitnesses)
            p2 = tournament_selection(population, fitnesses)
            c1, c2 = crossover(p1, p2)
            new_pop.extend([mutate(c1), mutate(c2)])
        population = new pop[:POP SIZE]
```

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def gene expression optimization():
   population = [random_gene() for _ in range(POP_SIZE)]
   for gen in range(GENERATIONS):
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      best_expr = population[best_idx]
      best fit = fitnesses[best idx]
      if gen % 10 == 0 or gen == GENERATIONS - 1:
          print(f"Generation \ \{gen: 3d\} \ | \ Best \ Expr: \ \{best\_expr\} \ | \ Fitness = \{best\_fit:.6f\}")
      new_pop = [best_expr] # elitism
      while len(new_pop) < POP_SIZE:</pre>
          p1 = tournament selection(population, fitnesses)
          p2 = tournament_selection(population, fitnesses)
          c1, c2 = crossover(p1, p2)
          new_pop.extend([mutate(c1), mutate(c2)])
      population = new_pop[:POP_SIZE]
   print("\n ✓ Optimization Complete!")
   print(f"Best evolved expression: {best_expr}")
   return best_expr
# Run
if __name__ == "__main__":
   best = gene expression optimization()
   print("\nSample comparison:")
   for x in [-3, 0, 2, 5]:
      print(f"x=\{x:>2\} \ | \ Target=\{target\_function(x):>8.3f\} \ | \ Evolved=\{evaluate\_expression(best, \ x):>8.3f\}")
            Generation
                0 | Best Expr: x | Fitness = 2487.000000
   Generation 10 | Best Expr: x | Fitness = 2487.000000
   Generation 20 | Best Expr: x | Fitness = 2487.000000
   Generation 30 | Best Expr: x | Fitness = 2487.000000
   Generation 40 | Best Expr: x | Fitness = 2487.000000
   Generation 50 | Best Expr: x | Fitness = 2487.000000
   Generation 60 | Best Expr: x | Fitness = 2487.000000
   Generation 70 | Best Expr: x | Fitness = 2487.000000
   Generation 80 | Best Expr: x | Fitness = 2487.000000
   Generation 90 | Best Expr: x | Fitness = 2487.000000
   Generation 99 | Best Expr: x | Fitness = 2487.000000
    Optimization Complete!
   Best evolved expression: x
   Sample comparison:
   x=-3 | Target= 7.000 | Evolved= -3.000
   x= 0 | Target= 1.000 | Evolved=
                                            0.000
   x= 2 | Target= 7.000 | Evolved=
                                            2.000
   x= 5 | Target= 31.000 | Evolved=
                                            5.000
```