

Chapter 10: Genetic Programming

Computational Intelligence: Second Edition

Contents

- Introduction
- Tree-Based Representation

Compact Overview

- Developed by Koza in the late 1980s
- A specialization of GAs, where tree-based representations are used
- Originally developed to evolve computer programs
- As part of the fitness function, each program is executed to determine its outcomes

Implications of Tree Representations

- Each chromosome represents a solution as a tree
- Implications:
 - Adaptive individuals
 - Size of individuals not fixed
 - Size depends on depth of tree and branching factor
 - Domain-specific grammar
 - A grammar needs to be defined that accurately reflects the problem to be solved
 - It should be possible to represent any possible solution using the defined grammar

The Grammar

- Terminal set specifies all the variables and constants
- Function set contains all the functions that can be applied to the elements of the terminal set
 - mathematical
 - arithmetic
 - Boolean
 - decision structures such as *if-then-else* and loops
- Semantic rules which characterize valid solutions

Example: Boolean Expressions

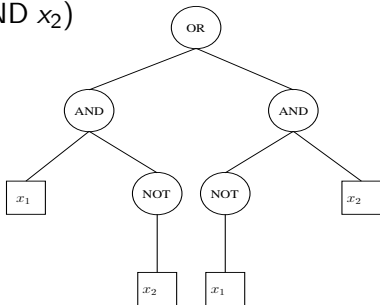
- Evolve Boolean expression

$$(x_1 \text{ AND NOT } x_2) \text{ OR } (\text{NOT } x_1 \text{ AND } x_2)$$

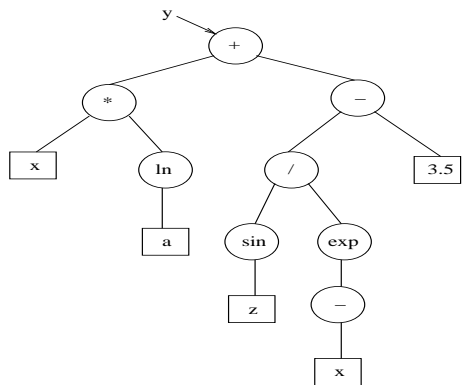
given data set

x_1	x_2	Target Output
0	0	0
0	1	1
1	0	1
1	1	0

- Function set: $\{AND, OR, NOT\}$
- Terminal set: $\{x_1, x_2\}$ where $x_1, x_2 \in \{0, 1\}$



Example: Boolean Expressions



- Evolve mathematical expression

$$y = x * \ln(a) + \sin(z) / \exp(-x) - 3.4$$

- Terminal set: $\{a, x, z, 3.4\}$
with $a, x, z \in \mathbb{R}$
- Function set:
 $\{-, +, *, /, \sin, \exp, \ln\}$

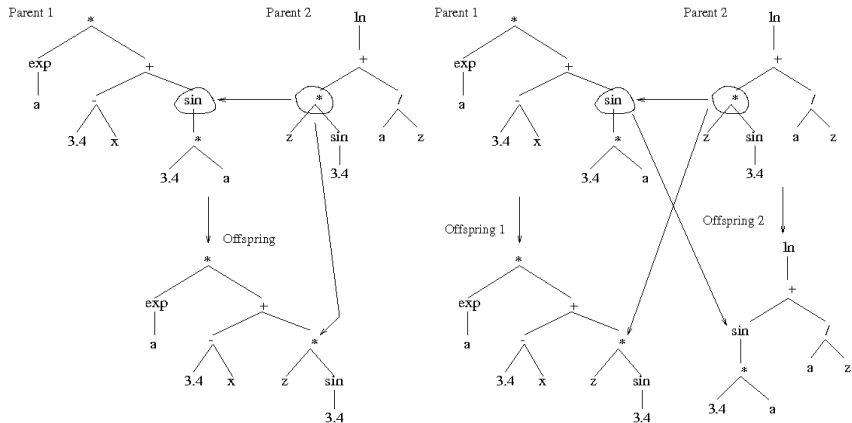
Initial Population

- Randomly generated within the restrictions of a maximum depth and semantics as expressed by the given grammar
- For each individual, a root is randomly selected from the set of function elements
- Non-terminal nodes are randomly selected from the function set
- Terminal nodes selected from the terminal set
- Start with small initial trees

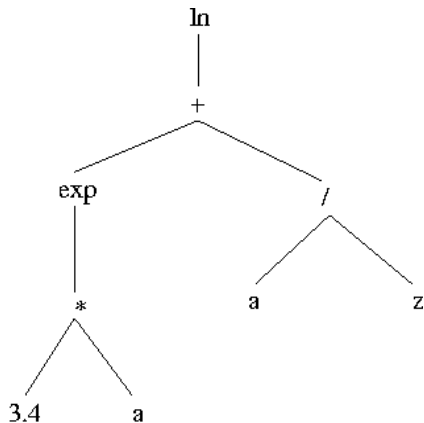
Fitness Function

- Fitness function is usually problem-dependent
- Fitness calculation requires the program to be evaluated against a number of test cases
- For the Boolean expression:
 - Fitness is the number of correctly predicted target outputs
- For the mathematical expression:
 - Fitness is the mean-squared error with respect to supplied target outputs
- Fitness function may include a penalty to penalize individuals with undesirable structural properties:
 - Penalize invalid solutions
 - Penalize tree size

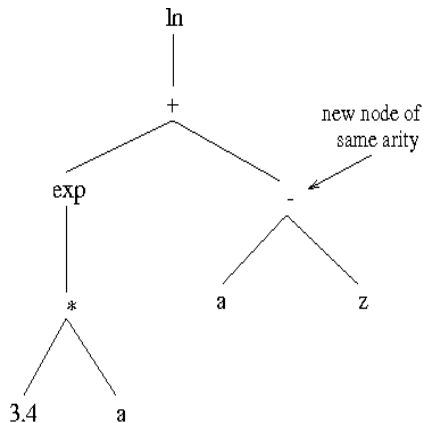
Crossover: Figure 10.3



Mutation: Figure 10.4

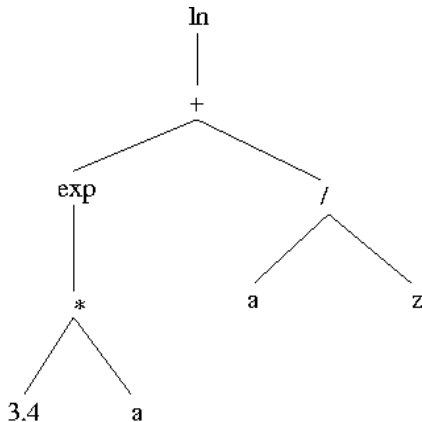


(a) Original Tree

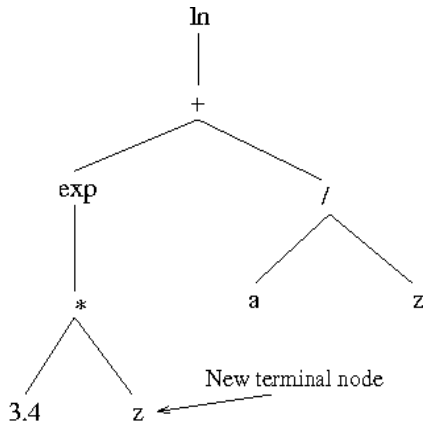


(b) Function Node Mutation

Mutation: Figure 10.4 (cont)

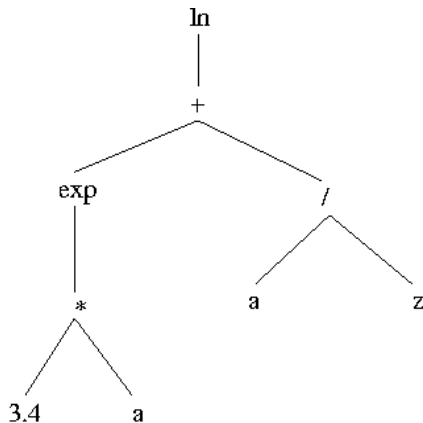


(c) Original Tree

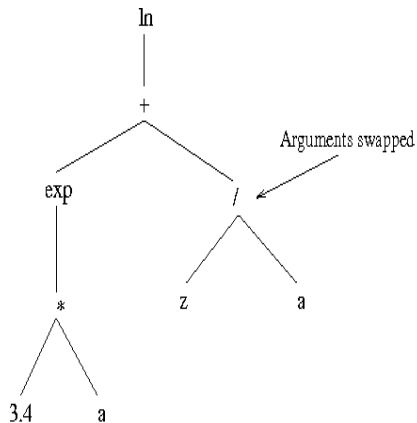


(d) Terminal Node Mutation

Mutation: Figure 10.4 (cont)

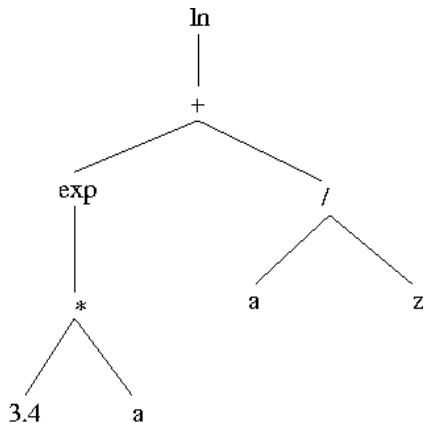


(e) Original Tree

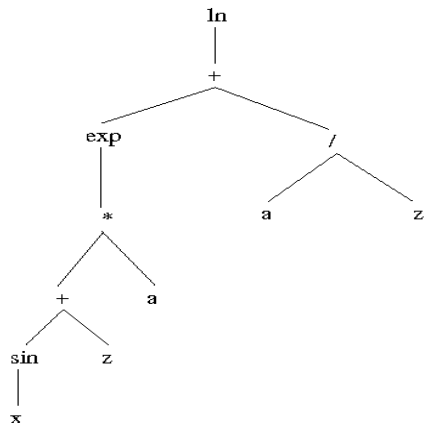


(f) Swapping Mutation

Mutation: Figure 10.4 (cont)

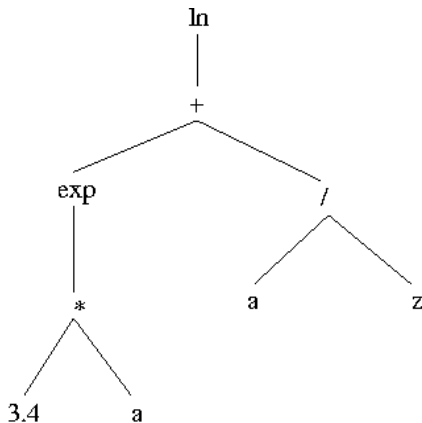


(g) Original Tree

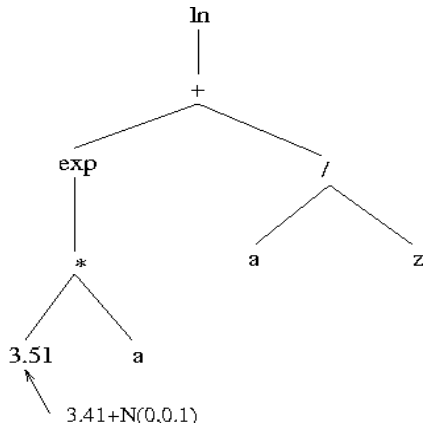


(h) Grow Mutation

Mutation: Figure 10.4 (cont)

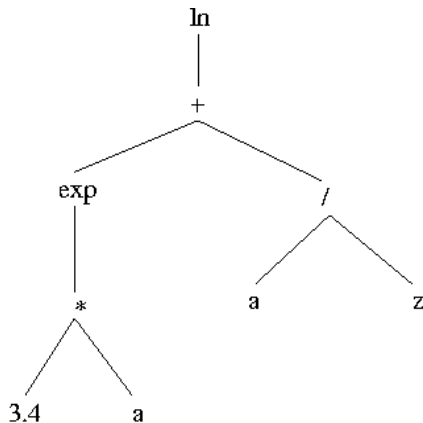


(i) Original Tree

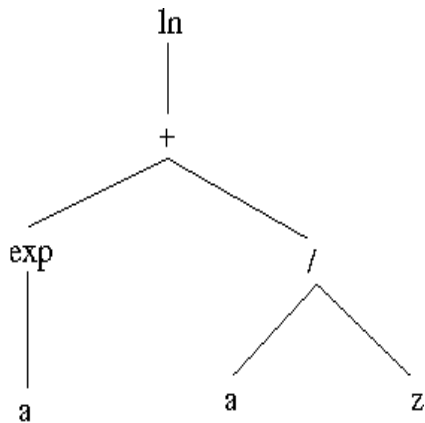


(j) Gaussian Mutation

Mutation: Figure 10.4 (cont)



(k) Original Tree



(l) Trunc Mutation

Building-Block Approach to GP

- Start with the most simple trees, with trees growing during the evolutionary process as needed
- Initial individuals consist of only a root and the immediate children of that node
- Expand trees when
 - the simplicity of the population's individuals can no longer account for the complexity of the problem to be solved
 - no improvement in the fitness of any of the individuals within the population is observed
- Expansion occurs by adding a randomly generated building block (i.e. a new node) to individuals
- Expansion occurs at a specified expansion probability, p_e
- Helps to reduce the computational complexity of the evolution process, and to produce smaller individuals