Evolutionary Computing Genetic Programming

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Automatic Programming

- Automatic generation of computer programs
- Saying WHAT is wanted but not HOW to do it
- The programs can be of the most GENERAL form:
 - Subroutines (e.g. sorting, search)
 - Planning (e.g. sequence of moves of robotic arm)
 - Strategy in games (e.g. Pacman, Quake, Civilization ...)
 - Classification (e.g. character recognition)
 - Prediction (e.g. consumption of electricity)
 - Control (e.g. vehicle steering ...)
 - Design of electronic circuits (e.g. chip layout)

Genetic Programming (GP)

- GP is a branch of EC with a goal to generate a computer program (or just function/expression) that best fits user's requirements.
- The computer program acts as a candidate solution (like a binary string in GA).
- GP is computationally expensive and its application area was limited in 1990s.
- Recently, GP has started delivering human-competitive machine intelligence thanks to exponential growth in CPU power.

Representation of programs

- Programming languages: C, Java, Prolog, machine language, LISP
- Special languages: robots, Turing machines
- Typically, data and programs are treated the same way (lists or S-expressions):
 - (dotimes i 3 (setq v (* i i)))
 - (3 4 5 (to b c))
 - Language = functions + terminals

Example: Even parity

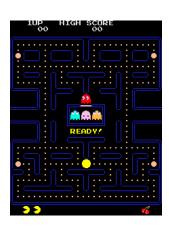
- Generate a computer program that takes 10 bits and returns whether the number of 1's is even:
 - Even-Parity $(1,0,0,0,1,0,0,1,1,0) \Rightarrow TRUE$
- In terms of:
 - Functions: AND, OR, NAND, NOR, NOT
 - Terminals: B0, B1, B2..., B9
- Large number of input/output test cases (2¹⁰ = 1024 cases)

Test cases for 10-bit even parity (1024 cases)

B0	B1	B2	B3	B4	B5	B6	B7	B8	B9	OUT
0	0	0	0	0	0	0	0	0	0	TRUE
0	0	0	0	0	0	0	0	0	1	FALSE
0	0	0	0	0	0	0	0	1	1	TRUE
1	1	1	1	1	1	1	1	1	1	TRUE

Example: Strategy for Pacman game

- Functions:
 - if-obstacle,
 if-pill,
 if-power-pill,
 if-ghost (they are of the
 form if-then-else)
 - sequence2, 3, 4 ...
- Terminals: advance, turn-left, -right
 - advance, turn-left, -right
- GOAL: to eat all pills within a time limit



Example of program in Pacman game

GP Terminology

- Parse tree
 - A tree structure representing a program for GP
- Function set
 - Function set is the set of operators used for the program.
 - Functions take the internal points of the parse tree.
 - An example of function set is $\{+, -, *, \%, >, IF\}$.
- Terminal set
 - The set of terminal nodes in the parse tree.
 - A terminal might be a variable, a constant, or a function taking no argument.
 - An example of terminal set is X, Y, random-constants

Preparation Steps of GP

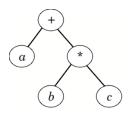
Determination of

- terminal set
- function set
- fitness measure
- parameters for GP run
- termination criterion

Program Representation

The program is generally represented by a parse tree of the function set and terminal set rather than lines of code.

E.g., a simple expression $a + b \cdot c$ would be represented as:



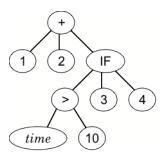
Program Representation

As another example, consider the following simple function in C language:

```
int foo (int time)
{
    int a, b;
    if(time > 10)
        a = 3;
    else
        a = 4;
    b = a + 1 + 2;
    return b;
}
```

Program Representation

The function can be represented by the following tree:



GP Algorithm

- Formation of an initial random population computer of programs, using functions and terminals
- Execution (!) of all programs and their evaluation (fitness function)
- Selection of well-performing programs
- Creation of a new population by applying genetic operators to selected programs
- Return to 2 until finding a "good" program

GP as search

- Genetic Programming performs a heuristic search in the space of computer programs (a type of "best-first" search with an opened list of limited size – the population)
- The heuristic function is the fitness function
- The search operators are the genetic operators (mutation and crossover)

Generation of an initial population

- Choose a function for the root of the tree
- Check the arity of the function
- For each argument of the function, generate:
 - a terminal; or
 - a subtree
- Practically, trees deeper that certain constant should not be generated

Methods to generate individuals

- Full: same depth for all branches of a tree
- Grow: variable depth (up to a set limit)
- Ramped half and half:
 - Mixture of full and grow
 - 50% will be full and 50% grow
- Objective: to maximize diversity

Functions and Terminals

- Functions: functions or macros that take arguments
 - e.g. (+34)
- Terminals:
 - Functions that do not have arguments; e.g. (advance)
 - Constant: 3, a, . . .
 - (Ephemeral) random constant \mathcal{R} : for numerical problems and symbolic regression
 - Input variables: D0, D1, ...

Functions

- Funcions/terminals sufficient to express the solution (e.g. for boolean functions, AND, OR and NOT are sufficient)
- It may be useful to include powerful functions (so that the system does not have to rediscover them); e.g. sin(x)
- Differentiate between functions (evaluate arguments; e.g. sum) and macros (do not evaluate: if-then-else);
- Functions have to execute with any argument and without producing errors (closure); e.g. protection against division by zero
- In standard GP there are no data types. All functions must be prepared to receive any type and value of argument

Evaluation of programs (fitness)

- Raw:
 - e.g. number of correctly guessed cases or hits (even-parity)
 - e.g. function, such as $[0.7 \times points + 0.3 \times time]$ (Pacman)
- Standard: (to be minimized)
 - Standard Fitness = Maximum raw
- Adjusted: 1/(1 + standard); it exaggerates fitness near 0
- Normalized (or relative):
 - ajusted / (sum of adjusted fitnesses in population)

GP Operations

- Reproduction (selection)
- Mutation
- Crossover
- Architecture-altering operations

Selection

Selection of capable (well-performing) individuals

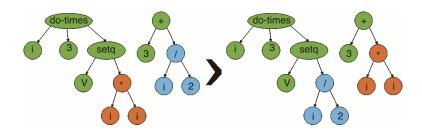
- Fitness proportional (probabilistic): uses normalized fitness
 - Problem: super-individuals
- Ranked
- Tournament
 - Match several individuals to compete among themselves: the best one reproduces
- Greedy Overselection (of the fitter individuals): create a high fitness group, H, and a low fitness group, L
 - 80% of the time select the parent from group H
 - 20% of the time select the parent from group L

Mutation Operation

- Given an individual, randomly pick a (internal or terminal) point in the parse tree.
- Delete subtree at the picked point.
- Grow new subtree at the picked point in the same way as used to generate initial individuals.
- Repair the individual if the associated program is invalid.

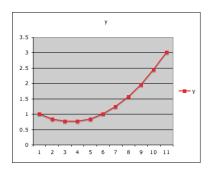
Crossover

- Given two individuals, pick a point in each parse tree independently.
- Swap the subtrees at the picked points.
- Repair the parse tree if the associated program is invalid.



Example I: Nonlinear Function Identification

Find a nonlinear function that satisfies the following relationship:



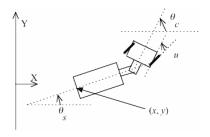
X	У
-1.00	1.00
-0.80	0.84
-0.60	0.76
-0.40	0.76
-0.20	0.84
0.00	1.00
0.20	1.24
0.40	1.56
0.60	1.96
0.80	2.44
1.00	3.00

Example I: Nonlinear Function Identification

Terminal set	$T = \{x, Random\text{-constants}\}$				
Function set	$F = \{+, -, *, \%\}$				
Fitness	Sum of errors between the candidate program's				
	outputs and y's for all values of x.				
Parameter	Population size: 4				
Termination	When the sum of errors is less than, e.g. 0.1				

Example II: Truck Backer Upper Control Law Design

Objective: Develop a control law $u=(x,y,\theta_S,\theta_C)$ that leads the trailer tail to $(x,y,\theta_S)=(0,0,0)$ when the backward linear velocity of the wheels is fixed.



Example II: Truck Backer Upper Control Law Design

There are four input variables of the controller:

- Horizontal position, x
- Vertical position, y
- Trailer angle, θ_S
- Cap angle, θ_C

Thus, *terminals* $T = \{x, y, \theta_S, \theta_C, \text{Random-constants}\}$

Function set: arithmetic and trigonometric functions

To *evaluate* a solution candidate during evolution process, we need to conduct simulation runs over many initial conditions of the trailer for hundreds of time steps.