

Applying Genetic Programming to Bytecode and Assembly

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The big picture

- Evolving whole programs is hard to do with source code.
- Evolving whole programs with bytecode and assembly is not as hard.

Outline

- 1 Evolutionary Computation
- 2 Why Bytecode and Assembly?
- 3 Java Bytecode and the JVM
- 4 FINCH:Evolving Java Bytecode
- 5 Using Instruction-level Code to Automate Bug Repair
- 6 Conclusions

Outline

1 Evolutionary Computation

- What is it?
- How does it work?
- Genetic Programming

2 Why Bytecode and Assembly?

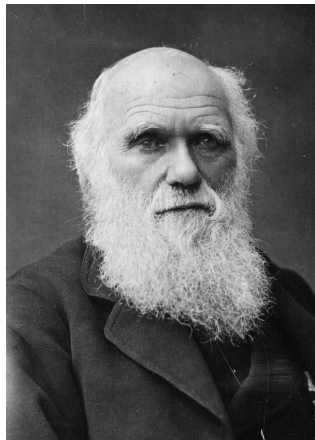
3 Java Bytecode and the JVM

4 FINCH: Evolving Java Bytecode

5 Using Instruction-level Code to Automate Bug Repair

What is Evolutionary Computation?

- Evolutionary Computation (EC) is a technique that is used to automate computer problem solving.
- Loosely emulates evolutionary biology.



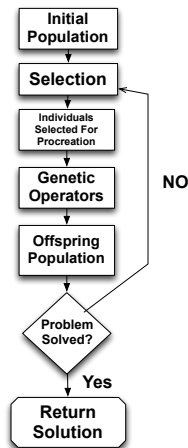
Charles Darwin

<http://tinyurl.com/lqwj3wt>



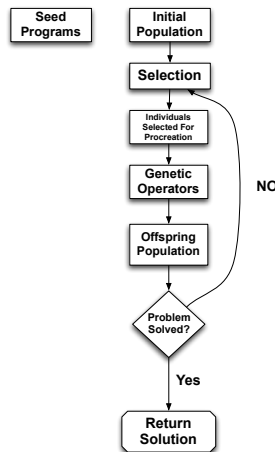
How does it work?

- Continuous optimization
- Selection is driven by the *fitness* of individuals
- Genetic operators mimic sexual reproduction and mutation



Genetic Programming

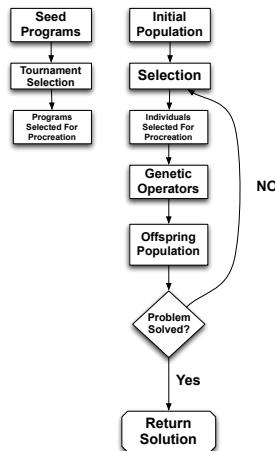
- Genetic programming (GP) uses the EC process to evolve **programs**
- This done by using an Evolutionary Algorithm (EA)



Genetic Programming

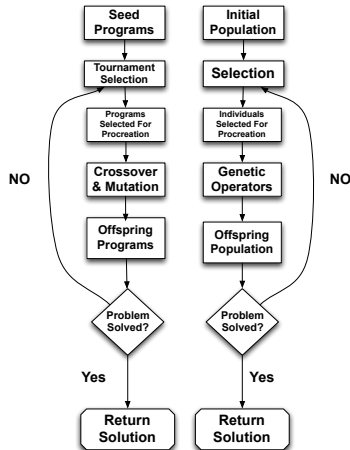
■ Tournament Selection

- 1 Randomly select a specified number of programs
- 2 Pick the program with the highest fitness
- 3 That program then is selected for reproduction

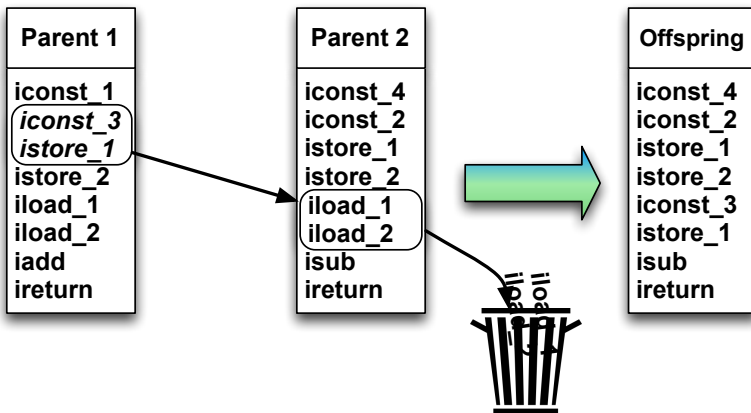


Genetic Programming

Two genetic operators used in GP are *crossover* and *mutation*

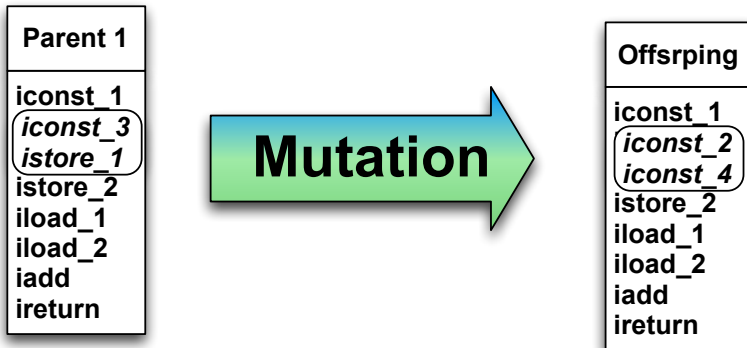


Crossover



Crossover with Java Bytecode

Mutation



Crossover with Java Bytecode

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- 1 Evolutionary Computation
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 - Difficulties in Source Code
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Syntax vs Semantics

```
float x; int y = 7;
if (y >= 0)
    x = y;
else
    x = -y;
System.out.println(x);
```

(a)

```
int x = 7; float y;
if (y >= 0) {
    y = x;
    x = y;
}
System.out.println(z);
```

(b)

Both (a) and (b) are valid syntactically. However (b) is invalid semantically.

EAs and Source Code

- EAs that evolve source code are usually designed to avoid dealing with semantic constraints

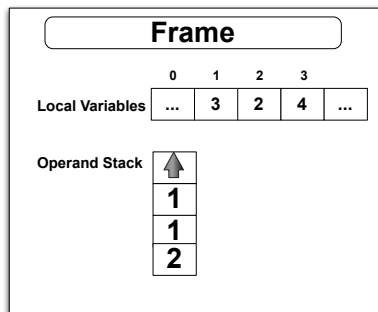
```
class Robot{
...
    double robotSpeed(){
        double evolvedVariable = valueFromEA;
        return (robot.location + evolvedVariable)/2;
    }
...
}
```


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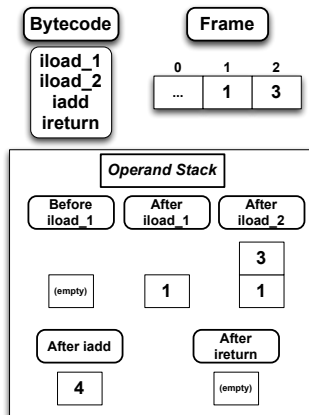
Java Virtual Machine

- A frame stores data and partial results as well as return values for methods



Java Bytecode and Frames

- Opcodes
- The prefix indicates type



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 - How it Works
 - The Array Sum Problem
- 5 Using Instruction-level Code to Automate Bug Repair

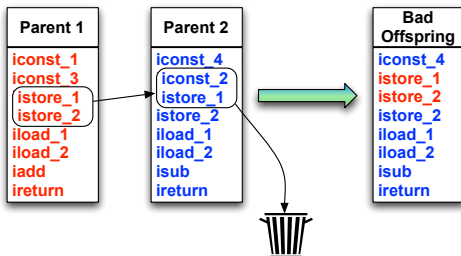
The semantic constraints that are checked for are

- Stack and Frame Depth
- Variable Types
- Control Flow

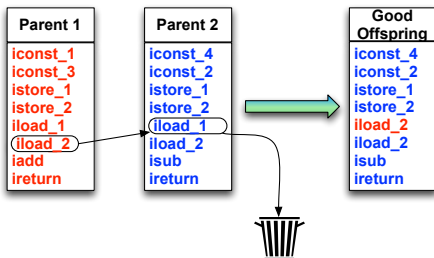
Dealing With Semantic Constraints

- 1 Apply crossover to two parents
- 2 Check if they comply to semantic constraints
- 3 If the program passes the constraint test then it proceeds to offspring generation
- 4 If it fails the constrain check then another attempt is made with the same parents

Bad Crossover



Good Crossover



Array Sum

Decompiled Solution

```
int sumlistrec(List list) {  
    int sum = 0;  
    if (list.isEmpty())  
        sum = sum;  
    else  
        sum += ((Integer)list.get(0)).intValue() +  
               sumlistrec(list.subList(1,  
                                       list.size()));  
    return sum;  
}
```

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 - How it Works
 - Results

Applying Genetic Programming to Bytecode and Assembly

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Conclusions

[1] [2]

References



M. Orlov and M. Sipper.

Flight of the FINCH Through the Java Wilderness.

Evolutionary Computation, IEEE Transactions on,
15(2):166–182, April 2011.



E. Schulte, S. Forrest, and W. Weimer.

Automated Program Repair Through the Evolution of
Assembly Code.

*In Proceedings of the IEEE/ACM International Conference
on Automated Software Engineering, ASE '10*, pages
313–316, New York, NY, USA, 2010. ACM.