# Applying Genetic Programming to Bytecode and Assembly

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The big picture

Overview

## The big picture

Background

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



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



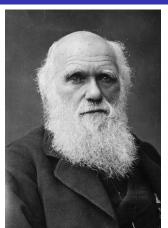
#### **Outline**

- 1 Evolutionary Computation
  - What is it?
  - How does it work?
  - Genetic Programming
- 2 Why Byetocde and Assembly?
- 3 Java Bytecode and the JVM
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- 5 Using Instruction-level Code to Automate Ruf® Refair きゃ まっつく

Background

## What is Evolutionary Computation?

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



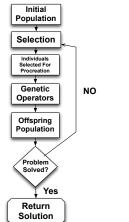
Charles Darwin
http://tinyuml.com/lqwj3wt



**Evolutionary Computation** 

## How does it work?

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

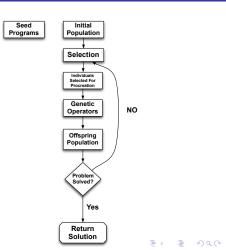




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## Genetic Programming

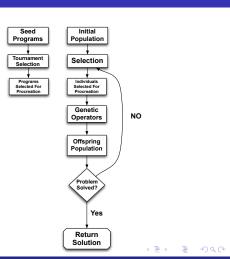
- Genetic programming (GP) uses the EC process to evolve programs
- This done by using an Evolutionary Algorithm (EA)



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## Genetic Programming

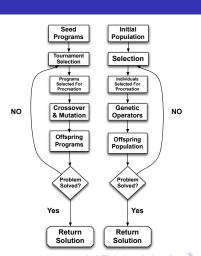
- Tournament Selection
  - Randomly select a specified number of programs
  - Pick the program with the highest fitness
  - 3 That program then is selected for reproduction



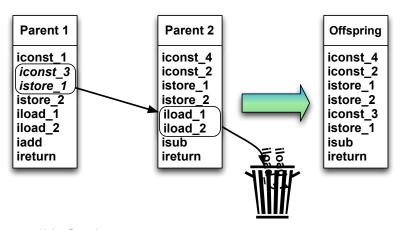
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## Genetic Programming

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



#### Crossover



Crossover with Java Bytecode



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#### Mutation

iconst\_1
iconst\_3
istore\_1
istore\_2
iload\_1
iload\_2
iadd
ireturn



Offsrping

iconst\_1 iconst\_2 iconst\_4 istore\_2 iload\_1 iload\_2 iadd ireturn

Crossover with Java Bytecode



## Outline

- 1 Evolutionary Computation
- Why Byetocde and Assembly?
  Difficulties in Source Code
- 3 Java Bytecode and the JVM
- 4 FINCH: Evolving Java Bytecode
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## Source Code Semantic Constraints

- It is difficult to apply evolution to an entire program in source code
  - Source code is made to simplify reading and writing programs
  - Source code does not represent the semantic constraints of the program.



## Syntax vs Semantics

- Syntax represents structure
- Semantics represent meaning

Semantically Wrong: The sun rises in the West. Semantically Correct: The sun rises in the East.



Background

## 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;
        x = y;
        System.out.println(z);
System.out.println(z);
```

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;
  }
...
}
```



#### Instruction-Level Code Constraints

- Consists of a smaller alphabets
- Simpler syntactically
- Less semantic constraints to violate



## Instruction-level Code Benefits

- Do not need to understand the structure of the program being evolved
- Can evolve a lot from a little
- If there is a compiler for it we can evolve it



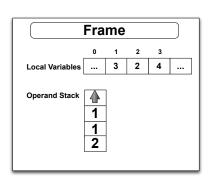
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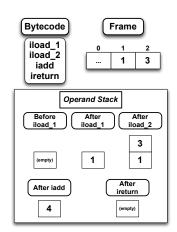
#### Java Virtual Machine

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



## Java Bytcode and Frames

- Opcodes
- The prefix indicates type





## Outline

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



#### What is FINCH?

- FINCH is an EA developed by M. Orlov and M. Sipper
- It evolves Java bytecode
- It deals with semantic constraints



## **Dealing With Semantic Constraints**

The semantic constraints that are checked for are

- Stack and Frame Depth
- Variable Types
- Control Flow



## **Dealing With Semantic Constraints**

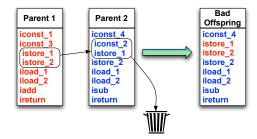
- Apply crossover to two parents
- Check if they comply to semantic constraints
- If the program passes the constraint test then it proceeds to offspring generation
- If it fails the constrain check then another attempt is made with the same parents



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How it works

#### **Bad Crossover**

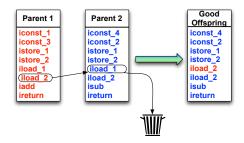




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How it works

## **Good Crossover**





The Array Sum Problem

## **Array Sum**

- The array sum problem
  - Started with a zero fitness seed program
  - Counted function calls to check for a non-halting state

The Array Sum Problem

## Array Sum

#### **Decompiled Solution**

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- 5 Using Instruction-level Code to Automate Bug Repair
  - How it Works
  - Results



## Automating Bug Repair

- Schulte et al., automated bug repair by evolving Java bytecode and x86 assembly
- Does not check for semantic constraints



## Tests and Fitness

- Fitness was determined by tests
- Test consisted of one negative test and multiple positive tests
- The negative test was used to check if the bug was fixed



Background

- Programs at times consist of thousands of lines of code
- Uses a weighted path due to size of programs
- The weighted path was determined by what tests execute that instruction



## Instruction Weight

- Only executed by failing test: weight = 1.0
- Executed by negative test and one positive: weight = 0.1
- Not executed by negative test case: weight = 0



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## Conclusions



[1] [2]

Conclusions

#### References



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.

