# Applying Genetic Programming to Bytecode and Assembly

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Background

#### Outline

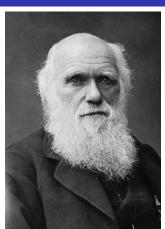
- 1 Evolutionary Computation
- 2 Why Evolve Bytecode and Assembly?
- 3 Java bytecode and the JVM
- 4 FINCH: Evolving Java Bytecode
- 5 Using Instruction-level Code to Automate Bug Repair
- 6 Conclusions

#### Outline

- **Evolutionary Computation** 
  - What is it?
  - How does it work?
  - Genetic Programming

## 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?

Background

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



## Genetic Programming

Background

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

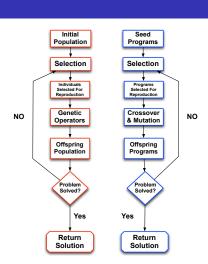


## Genetic Programming

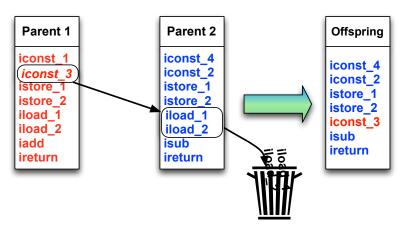
Background

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



#### Crossover



Crossover with Java Bytecode

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#### Mutation







Crossover with Java Bytecode

#### Outline

- 1 Evolutionary Computation
- 2 Why Evolve Bytecode and Assembly?
  - Difficulties With Source Code
  - Instruction-Level Code
- 3 Java bytecode and the JVN
- 4 FINCH: Evolving Java Bytecode
- 5 Using Instruction-level Code to Automate Bug Repair

Background

#### 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.

Difficulties With Source Code

## 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.

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Background

## Syntax vs Semantics

Why

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

```
float x; int y = 7;

if(y>= 0){

    x=y;

}else{

    x= -y;
}

System.out.println(x);

float y; int x = 7;

if(y>= 0){

    y=x;

    x=y;

}

System.out.println(z);
```

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Instruction-Level Code

Background

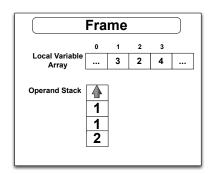
#### Instruction-Level Code Constraints

- Consists of smaller alphabets
- Simpler syntactically
- Fewer semantic constraints to violate

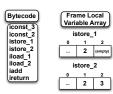
#### Outline

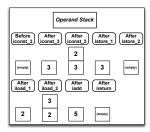
- Java bytecode and the JVM

- A frame stores data and partial results as well as return values for methods
- Each method call has a frame



- Opcodes
- The prefix indicates type





### Outline

- FINCH: Evolving Java Bytecode
  - How it Works
  - The Array Sum Problem

#### What is FINCH?

- FINCH is an EA developed by Orlov and 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

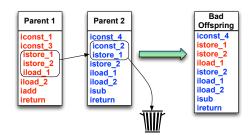
## **Dealing With Semantic Constraints**

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

Background Why Bytecode and Assembly FINCH Evolving Assembly Conclusions References

How it works

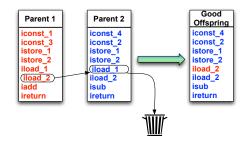
#### **Bad Crossover**



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How it works

#### **Good Crossover**



Background

## **Array Sum**

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

```
int sumlistrec(List list) {
  int sum = 0;
  if(list.isEmpty())
    sum *= sumlistrec(list);
  else
    sum += list.get(0)/2 + sumlistrec(
        list.subList(1, list.size()));
  return sum;
}
```

References

Background

The Array Sum Problem

## 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

## **Automating Bug Repair**

- Schulte et al., automated bug repair by evolving Java bytecode and x86 assembly
- Fixed bugs in real code
- Did not check for semantic constraints

## Weighted Path

- Programs at times consist of thousands of lines of code
- Uses a weighted path due to size of programs
- The weight of a path was determined by the instructions that were executed by tests

## Weighted Path

- Test were provided that consisted of one negative test and multiple positive tests
- The negative test was used to represent the bug and check if it was fixed
- The positive tests were used to retain functionality

## Instruction Weight

- Each instruction executed by only by the failing test was given a weight of 1.0
- An instruction executed by the negative test and atleast one positive was given a weight of 0.1
- If an instruction was not executed by the negative test case a weight of 0 was assigned

Results

## What was debugged?

Schulte et al., were able to debug:

- Infinite loops
- Buffer overflows
- Incorrect type declarations

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### Conclusions

Background

- It is difficult to evolve entire programs in source code due to semantic constraints
- It is easier to deal with semantic constraints with instruction-level code
- It is feasible to not deal with semantic constraints in some situations

References

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## Questions?

References

Background

Background



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