

# Applying Genetic Programming to Bytecode and Assembly

Eric Collom

Division of Science and Mathematics  
University of Minnesota, Morris  
Morris, Minnesota, USA

29 April '14,  
UMM Senior Seminar

# 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 Evolve Instruction-Level Code
- 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

### ■ EC

## 2 Why Evolve Instruction-Level Code

## 3 Java Bytecode and the JVM

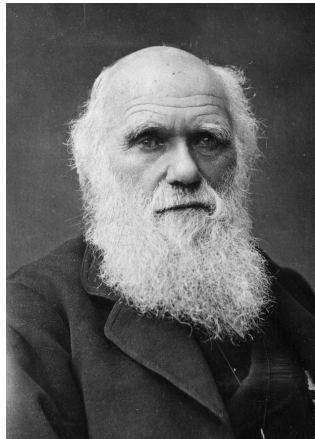
## 4 FINCH:Evolving Java Bytecode

## 5 Using Instruction-level Code to Automate Bug Repair

## 6 Conclusions

# What is Evolutionary Computation?

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



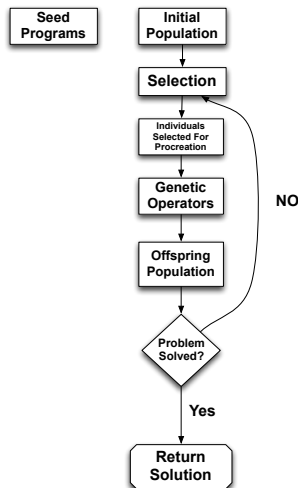
Charles Darwin

<http://tinyurl.com/lqwj3wt>



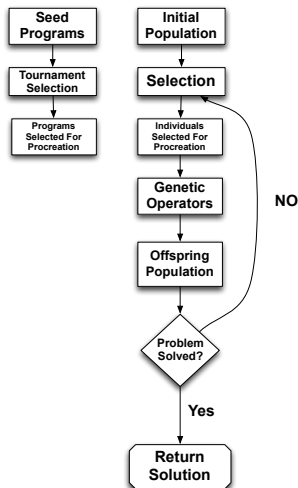
# Genetic Programming

- Uses the EC technique to evolve programs
- The population is programs



# Genetic Programming

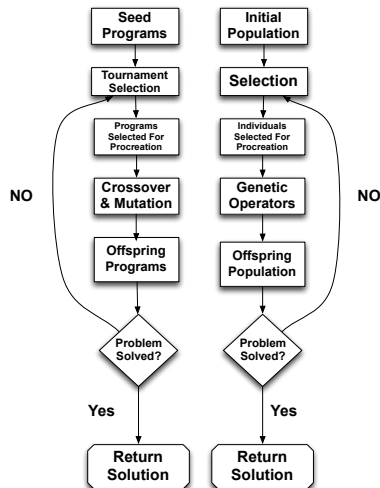
## ■ Tournament Selection



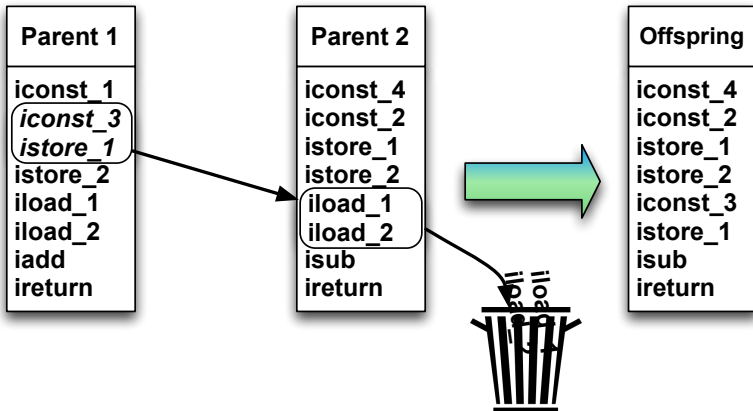


# Genetic Programming

- Crossover
- Mutation

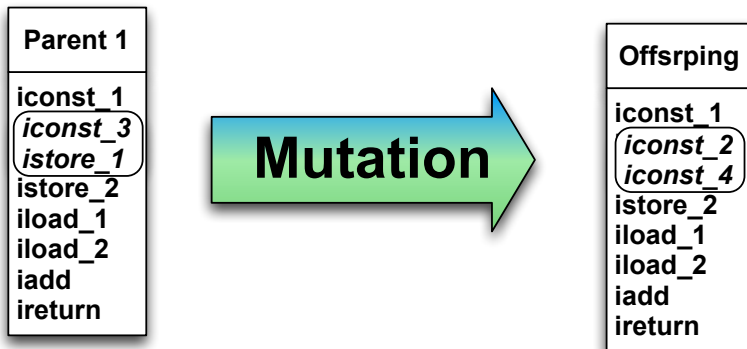


# Crossover



Crossover with Java Bytecode

# Mutation



Crossover with Java Bytecode

# Outline

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

# Source Code Constraints

- While it would be useful, 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.

# Source Code Constraints

```
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.

# Source Code Constraints

- EAs 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

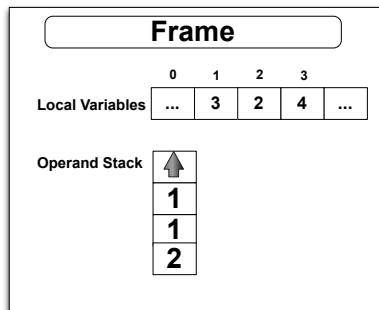
- Do not need to understand the structure of the program being evolved
- Can evolve a lot from a little
- If there is a compiler for it we can evolve it

# Outline

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

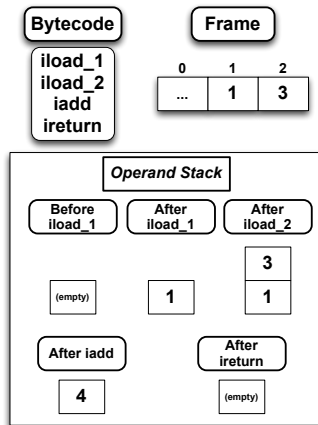
# Java Virtual Machine

- Frames
- Array of local variables
- Operand Stack



# Java Bytecode and Frames

- Opcodes
- Prefix indicates type



# Outline

- 1 Evolutionary Computation
- 2 Why Evolve Instruction-Level Code
- 3 Java Bytecode and the JVM
- 4 FINCH:Evolving Java Bytecode**
  - How it Works
  - Results
- 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 using crossover
- It is a good example of dealing with semantic constraints

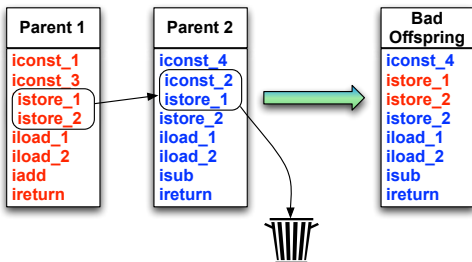
# Selecting Offspring

- There is still a chance to produce non-compilable code
- Solution: check if offspring follows semantic constraints
  - Stack and Frame Depth
  - Variable Types
  - Control Flow

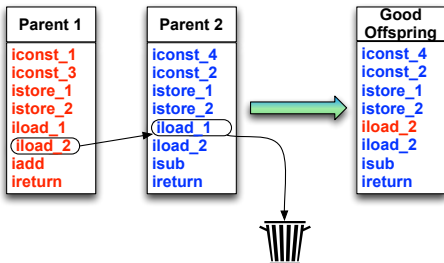
- 1 Apply crossover to two parents
- 2 Check if they comply to semantic constraints
- 3 If the program passes the constraint test then it proceeds to offspring generation
- 4 If it fails the constrain check then another attempt is made with the same parents



# Bad Crossover



# Good Crossover



# Array Sum

- 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;
}
```

# Array Sum

```

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;
}

```

# Outline

- 1 Evolutionary Computation
- 2 Why Evolve Instruction-Level Code
- 3 Java Bytecode and the JVM
- 4 FINCH:Evolving Java Bytecode
- 5 Using Instruction-level Code to Automate Bug Repair**
  - How it Works
  - Results



# Selecting Offspring

- Each instruction is given a weight
- The weight is determined by what tests execute that instruction
- Each experiment started with one negative test case and multiple positive test cases

# 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



## Results

- ## Buffer overflow

- ## Infinite loops

# Outline

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

# Conclusions

# References