



Genetic Programming in the Wild: Evolving Unrestricted Bytecode

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GP in the wild

Evolving
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GP: Programs or Representations?



“While it is common to describe GP as evolving **programs**, GP is not typically used to evolve programs in the familiar Turing-complete languages humans normally use for software development.”

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A Field Guide to Genetic Programming
[Poli, Langdon, and McPhee, 2008]

GP: Programs or Representations?



“While it is common to describe GP as evolving **programs**, GP is not typically used to evolve programs in the familiar Turing-complete languages humans normally use for software development.”

“It is instead more common to evolve programs
(or expressions or formulae)
in a **more constrained** and often **domain-specific language**.”

A Field Guide to Genetic Programming
[Poli, Langdon, and McPhee, 2008]

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Our Goals



From programs. . .

Evolve actual programs
written in Java

. . . to software!

Improve (existing) software
written in unrestricted Java

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Our Goals



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Evolve actual programs
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Extending prior work

Existing work uses **restricted subsets** of Java bytecode as
representation language for GP individuals

We evolve **unrestricted bytecode**

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Let's Evolve Java Source Code



- Rely on the building blocks in the initial population
- Defining **genetic operators** is problematic
- How to define **good** source code crossover?

Factorial (*recursive*)

```
class F {  
    int fact(int n) {  
        int ans = 1;  
  
        if (n > 0)  
            ans = n *  
                fact(n-1);  
  
        return ans;  
    }  
}
```



Factorial (*iterative*)

```
class F {  
    int fact(int n) {  
        int ans = 1;  
  
        for (; n > 0; n--)  
            ans = ans * n;  
  
        return ans;  
    }  
}
```

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“Stupid” Example



- **Source-level crossover** typically produces garbage

Factorial (*recursive* \times *iterative*)

```
class F {  
    int fact(int n) {  
        int ans = 1;  
  
        if (n >= 1;  
            for (; n > 0; n--)  
                ans = ans * n; n-1);  
  
        return ans;  
    }  
}
```

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- Maybe we can design **better** genetic operators?

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- Maybe we can design **better** genetic operators?
- Maybe... but too much harsh **syntax**
Possibly use **parse tree**?

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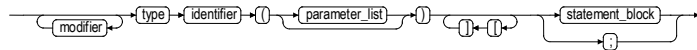


- Maybe we can design **better** genetic operators?
- Maybe... but too much harsh **syntax**
Possibly use **parse tree**?

Just one BNF rule (*of many*)

```
method_declaration ::=  $\Rightarrow$   
  modifier* type identifier  
  "(" parameter_list? ")" "[" "]"*  
  < statement_block | ";" >
```

method_declaration



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Better than parse trees:
Let's use **bytecode!**

Java Virtual Machine (JVM)

- Source code is compiled to **platform-neutral bytecode**
- Bytecode is executed with **fast** just-in-time compiler
- High-order, **simple** yet powerful architecture
- **Stack-based**, supports hierarchical object **types**
- Not limited to Java!
(*Scala, Groovy, Jython, Kawa, Clojure, ...*)

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Bytecode (cont'd)

Some basic bytecode instructions



Stack ↔ Local variables

- iconst** 1 pushes **int** 1 onto operand stack
- aload** 5 pushes **object** in local variable 5 onto stack
(*object type is deduced when class is loaded*)
- dstore** 6 pops two-word **double** to local variables 6–7

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Some basic bytecode instructions



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iconst 1 pushes **int 1** onto operand stack

aload 5 pushes **object** in local variable **5** onto stack
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dstore 6 pops two-word **double** to local variables **6–7**

Arithmetic instructions (*affect operand stack*)

imul pops two **ints** from stack, pushes multiplication result

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Stack ↔ Local variables

- iconst 1** pushes **int 1** onto operand stack
- aload 5** pushes **object** in local variable **5** onto stack
(*object type is deduced when class is loaded*)
- dstore 6** pops two-word **double** to local variables **6–7**

Arithmetic instructions (*affect operand stack*)

- imul** pops two **ints** from stack, pushes multiplication result

Control flow (*uses operand stack*)

- ifle +13** pops **int**, jumps **+13** bytes if value ≤ 0
- lreturn** pops two-word **long**, returns to caller's stack

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Evolutionary operators



- Java bytecode is **less fragile** than source code

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Bytecode (cont'd)

Evolutionary operators



- Java bytecode is **less fragile** than source code
- But, bytecode must be **correct** in order to run **correctly**

Correct bytecode requirements

Stack use is **type-consistent**

(e.g., can't multiply an *int* by an **Object**)

Local variables use is **type-consistent**

(e.g., can't read an *int* after storing an **Object**)

No stack underflow

No reading from uninitialized variables

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- So, genetic operators are still delicate

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- So, genetic operators are still delicate
- Need **good** genetic operators to produce **correct** offspring

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Local variables use is **type-consistent**

(e.g., can't read an *int* after storing an **Object**)

No stack underflow

No reading from uninitialized variables

- So, genetic operators are still delicate
- Need **good** genetic operators to produce **correct** offspring
- Conclusion: Avoid **bad** crossover and mutation

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Unidirectional bytecode crossover:

Bytecode A

...

...

...

Section α

...

...

...

Bytecode B

...

...

...

Section β

...

...

...

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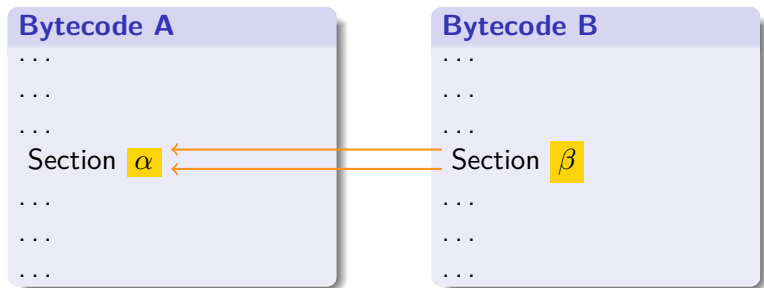
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Unidirectional bytecode crossover:



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Good and bad crossovers



Parent **A**:

Factorial (recursive)

```
class F
{
    int fact(int n)
    {
        int ans = 1;

        if (n > 0)
            ans = n * fact(n-1);

        return ans;
    }
}
```

Compiled bytecode

```
0  iconst_1
1  istore_2
2  iload_1
3  ifle 16
6  iload_1
7  aload_0
8  iload_1
9  iconst_1
10 isub
11 invokevirtual #2
14 imul
15 istore_2
16 iload_2
17 ireturn
```

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Good and bad crossovers



Parent **B**:

Factorial (*iterative*)

```
class F
{
  int fact(int n)
  {
    int ans = 1;
    for (; n > 0; n--)
      ans = ans * n;
    return ans;
  }
}
```

Compiled bytecode

```
0 iconst_1
1 istore_2
2 iload_1
3 ifle 16
6 iload_2
7 iload_1
8 imul
9 istore_2
10 iinc 1, -1
13 goto 2
16 iload_2
17 ireturn
```

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Good and bad crossovers



Replace a section in **A** with section from **B**

Bytecode A

```
0  iconst_1
1  istore_2
2  iload_1
3  ifle 16
6  iload_1
7  aload_0
8  iload_1
9  iconst_1
10 isub
11 invokevirtual #2
14 imul
15 istore_2
16 iload_2
17 ireturn
```



Bytecode B

```
0  iconst_1
1  istore_2
2  iload_1
3  ifle 16
6  iload_2
7  iload_1
8  imul
9  istore_2
10 iinc 1, -1
13 goto 2
16 iload_2
17 ireturn
```

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Good crossover example



Stack use is depth- and type-consistent, variables are initialized.

Bytecode A

```
0 iconst_1
1 istore_2
2 iload_1
3 ifle 16
6 iload_1
7 aload_0
8 iload_1
9 iconst_1
10 isub
11 invokevirtual #2
14 imul
15 istore_2
16 iload_2
17 ireturn
```



Bytecode B

```
0 iconst_1
1 istore_2
2 iload_1
3 ifle 16
6 iload_2
7 iload_1
8 imul
9 istore_2
10 iinc 1, -1
13 goto 2
16 iload_2
17 ireturn
```

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Good crossover example



Stack use is depth- and type-consistent, variables are initialized.

Bytecode (A \times B)

```
0 iconst_1
1 istore_2
2 iload_1
3 ifle 12
6 iload_1
7 iload_2
8 iload_1
9 imul
10 imul
11 istore_2
12 iload_2
13 ireturn
```

Decompiled source

```
class F
{
    int fact(int n)
    {
        int ans = 1;

        if (n > 0)
            ans = n * (ans * n);

        return ans;
    }
}
```

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Bad crossover example



Stack use is depth- and type-inconsistent.

Bytecode A

```
0  iconst_1
1  istore_2
2  iload_1
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6  iload_1
7  aload_0
8  iload_1
9  iconst_1
10 isub
11 invokevirtual #2
14 imul
15 istore_2
16 iload_2
17 ireturn
```



Bytecode B

```
0  iconst_1
1  istore_2
2  iload_1
3  ifle 16
6  iload_2
7  iload_1
8  imul
9  istore_2
10 iinc 1, -1
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16 iload_2
17 ireturn
```

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Stack use is depth- and type-inconsistent.

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```
0 iconst_1
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3 ifle 13
6 iload_2
7 iload_1
8 invokevirtual #2
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12 istore_2
13 iload_2
14 ireturn
```

“Decompiled” source

```
class F {
    int fact(int n)
    {
        int ans = 1;

        if (n > 0)
            ans = ans.fact(n) * ?;

        return ans;
    }
}
```

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Compatible Crossover

Constraints of unidirectional crossover $A \overset{\leftarrow}{\times} B$



Good crossover is achieved by respecting bytecode constraints:

(α is target section in **A**, β is source section in **B**)

Operand stack

e.g., β doesn't pop values with types incompatible to those popped by α

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(α is target section in **A**, β is source section in **B**)

Operand stack

e.g., β doesn't pop values with types incompatible to those popped by α

Local variables

e.g., variables read by β in **B** must be written before α in **A** with compatible types

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(α is target section in **A**, β is source section in **B**)

Operand stack

e.g., β doesn't pop values with types incompatible to those popped by α

Local variables

e.g., variables read by β in **B** must be written before α in **A** with compatible types

Control flow

e.g., branch instructions in β have no "outside" destinations

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Formal Definition

(Example of operand stack requirement)

α and β have compatible stack frames up to stack depth of β :
pops of α have identical or narrower types as pops of β ;
pushes of β have identical or narrower types as pushes of α

Good crossover

	α	β
pre-stack	**AB	**AA
post-stack	**B	**C
depth	3	2

Stack pops “AB”
(2 stop tack frames) are narrower than “AA”,
whereas stack push “C” is narrower than “B”

Types hierarchy: $C \rightarrow B \rightarrow A$

(see [Orlov and Sipper, 2009] for full formal definitions)

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Formal Definition

(Example of operand stack requirement)

α and β have compatible stack frames up to stack depth of β :
pops of α have identical or narrower types as pops of β ;
pushes of β have identical or narrower types as pushes of α

Bad crossover

	α	β
pre-stack	**AB	**Af
post-stack	**B	**A
depth	3	2

Stack pops "AB" are not narrower than "Af"
(*B and f are incompatible*);
stack push "A" is not narrower than "B"

Types hierarchy: $B \rightarrow A$; f is a **float**

(see [Orlov and Sipper, 2009] for full formal definitions)

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Symbolic Regression

As an evolutionary example...



Parameters

- Objective: symbolic regression, $x^4 + x^3 + x^2 + x$
- Fitness: sum of errors on 20 random data points in $[-1, 1]$
- Input: **Number** num (*a Java type*)

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Symbolic Regression

As an evolutionary example...



Parameters

- Objective: symbolic regression, $x^4 + x^3 + x^2 + x$
- Fitness: sum of errors on 20 random data points in $[-1, 1]$
- Input: **Number** num (a Java type)

Seeding

- Population initialized using seeding
[Langdon and Nordin, 2000]
- Seed population with clones of Koza's original worst-of-generation-0
[Koza, 1992]

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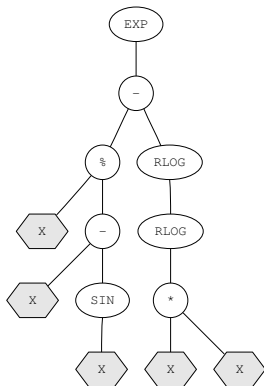
Symbolic Regression

Seeding with Koza's worst-of-generation-0



Original **Lisp** individual and its **tree** representation:

```
(EXP (- (% X (- X (SIN X))) (RLOG (RLOG (* X X)))))
```



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Seeding with Koza's worst-of-generation-0



Translation to **unrestricted Java**

```
class Gecco {
    Number simpleRegression(Number num) {
        double x      = num.doubleValue();
        double llsq   = Math.log(Math.log(x*x));
        double dv     = x / (x - Math.sin(x));
        double worst  = Math.exp(dv - llsq);
        return Double.valueOf(worst + Math.cos(1));
    }

    /* Rest of class omitted */
}
```

We added a couple of building blocks in the last line

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Symbolic Regression

Setup and Statistics



Setup (similar to Koza's)

- Population: 500 individuals
- Generations: 51 (or less)
- Probabilities: $p_{\text{cross}} = 0.9$
(α and β segments are uniform over segment sizes)
- Selection: binary tournament

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- Population: 500 individuals
- Generations: 51 (or less)
- Probabilities: $p_{\text{cross}} = 0.9$
(α and β segments are uniform over segment sizes)
- Selection: binary tournament

Statistics

- Yield: 99% of runs successful (out of 100)
- Runtime: 30–60 s on dual-core 2.6 GHz Opteron
- Memory limits: insignificant w.r.t. runtime

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Symbolic Regression

Evolved perfect individuals



A perfect solution easily evolves:

(beware of decompiler quirks!)

```
class Gecco_0_7199 {
    Number simpleRegression(Number num) {
        double d = num.doubleValue();
        d = num.doubleValue();
        double d1 = d; d = Double.valueOf(d + d * d *
            num.doubleValue()).doubleValue();
        return Double.valueOf(d +
            (d = num.doubleValue()) * num.doubleValue());
    }

    /* Rest of class unchanged */
}
```

Computes $(x + x \cdot x \cdot x) + (x + x \cdot x \cdot x) \cdot x = x(1 + x)(1 + x^2)$

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Evolved perfect individuals



Another solution:

```
class Gecco_0_2720 {
    Number simpleRegression(Number num) {
        double d = num.doubleValue();
        d = d; d = d;
        double d1 = Math.exp(d - d);
        return Double.valueOf(num.doubleValue() *
            (num.doubleValue() * (d * d + d) + d) + d);
    }

    /* Rest of class unchanged */
}
```

Computes $x \cdot (x \cdot (x \cdot x + x) + x) + x = x(1 + x(1 + x(1 + x)))$

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Completely **unrestricted** Java programs can be **evolved**
(*via bytecode*)

Extant (bad) Java programs can be **improved**
(*e.g., initial regression seed*)

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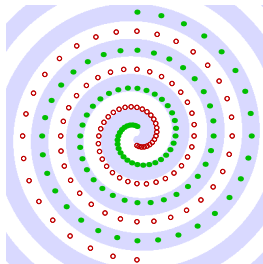
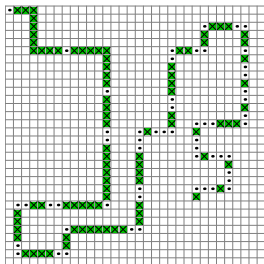
Future Work

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Exhibit **viability** on other problems

*We currently have results for:
complex regression, artificial ant, intertwined spirals, ...*



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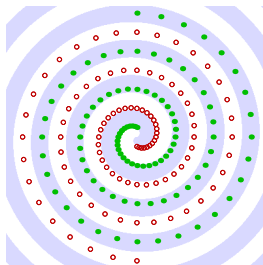
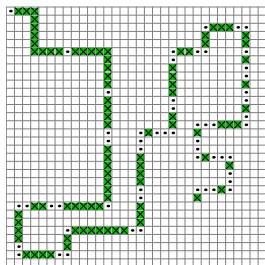
Future Work

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Exhibit **viability** on other problems

*We currently have results for:
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Loops and recursion are not a problem!

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