

Evolving unrestricted Java software with FINCH

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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



From programs...

Evolve actual programs
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... to software!

Improve (existing) software
written in unrestricted Java

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 do we 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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- **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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- **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;  
    }  
}
```




- 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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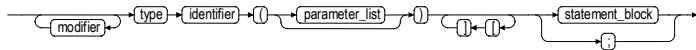


- Maybe we can design **better** genetic operators?
- Maybe... but too much harsh **syntax**
Possibly use **parse tree**?

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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Stack ↔ Local variables

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

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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**

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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Bytecode (cont'd)

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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Evolutionary operators



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No reading from uninitialized variables

- So, genetic operators are still delicate

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Evolutionary operators



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- 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

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

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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

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

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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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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

Local variables

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

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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

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, 2011] 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, 2011] 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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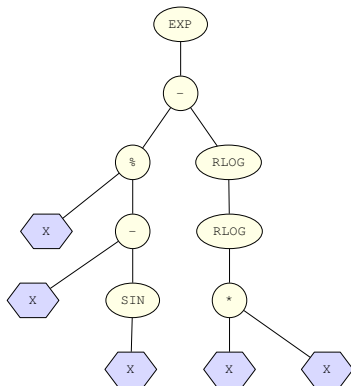
Symbolic Regression

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 SimpleSymbolicRegression {
    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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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

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 SimpleSymbolicRegression_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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Symbolic Regression

Evolved perfect individuals



Another solution:

```
class SimpleSymbolicRegression_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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Java Wilderness

Complex Regression



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Complex Regression

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Tic-Tac-Toe

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Parameters

- Objective: symbolic regression: $x^9 + x^8 + \dots + x^2 + x$
- Fitness: incremental evaluation, $\sum_{i=1}^n x^i$, up to $n = 9$
- Crossover: Gaussian distribution over segment sizes
- Parsimony pressure, growth limit

Initialization

- Worst of generation-0 from simple regression

Java Wilderness

Complex Regression



A perfect solution:

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

Computes

$$x + (x \cdot (x \cdot (x + (x + (((x \cdot x + x) \cdot x + x) \cdot x + x) \cdot x) \cdot x) + x) + x) \cdot x$$

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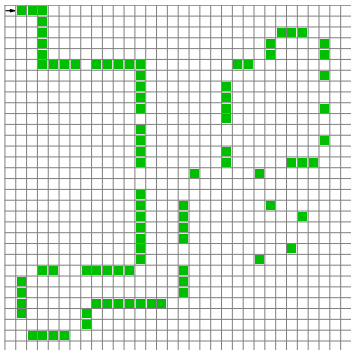
References

Java Wilderness

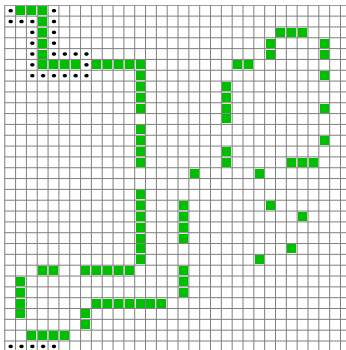
Artificial Ant



Santa Fe Trail:



(a) Initial setup



(b) Avoider

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Intertwined Spirals



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Parameters

- Objective: two-class classification of intertwined spirals
- Fitness: number of correctly classified points

Initialization

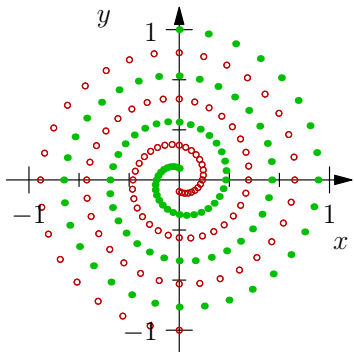
- Arbitrarily organized repository of building blocks: floating-point arithmetics, trigonometric functions, and polar-rectangular coordinates conversion

Java Wilderness

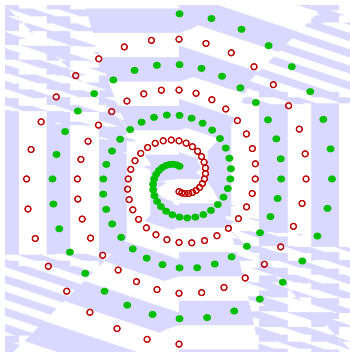
Intertwined Spirals



Intertwined spirals:



(e) Initial setup



(f) Koza's evolved solution

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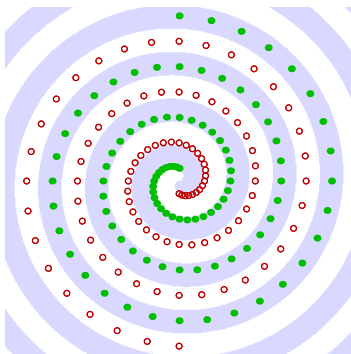
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Intertwined Spirals



A perfect solution:



Computes the (approximate) sign of $\sin\left(\frac{9}{4}\pi^2\sqrt{x^2 + y^2} - \tan^{-1}\frac{y}{x}\right)$ as the class predictor of (x, y)

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Koza's best-of-run:

```
(sin (iflte (iflte (+ Y Y) (+ X Y) (- X Y) (+ Y Y)) (* X X)
(sin (iflte (% Y Y) (% (sin (sin (% Y 0.30400002)))) X) (% Y
0.30400002) (iflte (iflte (% (sin (% (% Y (+ X Y))
0.30400002)) (+ X Y)) (% X -0.10399997) (- X Y) (* (+
-0.12499994 -0.15999997) (- X Y))) 0.30400002 (sin (sin
(iflte (% (sin (% (% Y 0.30400002) 0.30400002)) (+ X Y))
(% (sin Y) Y) (sin (sin (sin (% (sin X) (+ -0.12499994
-0.15999997)))))) (% (+ (+ X Y) (+ Y Y)) 0.30400002))))
(+ (+ X Y) (+ Y Y)))) (sin (iflte (iflte Y (+ X Y) (- X Y)
(+ Y Y)) (* X X) (sin (iflte (% Y Y) (% (sin (sin (% Y
0.30400002))) X) (% Y 0.30400002) (sin (sin (iflte (iflte
(sin (% (sin X) (+ -0.12499994 -0.15999997))) (% X
-0.10399997) (- X Y) (+ X Y)) (sin (% (sin X) (+
-0.12499994 -0.15999997))) (sin (sin (% (sin X) (+
-0.12499994 -0.15999997)))) (+ (+ X Y) (+ Y Y)))))) (%
Y 0.30400002))))))
```

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Our best-of-run:

```
boolean isFirst(double x, double y) {
    double a, b, c, e;
    a = Math.hypot(x, y);  e = y;
    c = Math.atan2(y, b = x) +
        -(b = Math.atan2(a, -a))
        * (c = a + a) * (b + (c = b));
    e = -b * Math.sin(c);
    if (e < -0.0056126487018762772) {
        b = Math.atan2(a, -a);
        b = Math.atan2(a * c + b, x);  b = x;
        return false;
    }
    else
        return true;
}
```

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Parameters

- Objective: summation of numbers in an input array
- Fitness: differences from actual sums on test inputs
- **Time limit:** 5000 backward branches

Code instrumentation

- Bytecode is instrumented with calls to time-limit check
- Before each backward branch and method invocation
- Robust and portable technique

Initialization

- “Weird” program that does **not** compute the sum

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Array sum: array loop solution

```
int sumlist(int list[]) {
    int sum = 0;
    int size = list.length;
    for (int tmp = 0; tmp < list.length; tmp++) {
        size = tmp;
        sum = sum + (list[tmp]);
    }
    return sum;
}
```

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Array sum: List loop solution

```
int sumlist(List list) {
    int sum = 0;
    int size = list.size();
    for (Iterator iterator = list.iterator();
         iterator.hasNext(); ) {
        int tmp = ((Integer) iterator.next())
                 .intValue();

        tmp = tmp + sum;
        if (tmp == list.size() + sum)
            sum = tmp;
        sum = tmp;
    }
    return sum;
}
```

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Array sum: List-recursive 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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The Tale of Alta Del



Parameters

- Objective: learn to play Tic-Tac-Toe
- Fitness: rounds won in single-elimination tournament

Initialization

- Negamax algorithm with α - β pruning and one of four (plausibly) insidious imperfections

Performance

- All imperfections are easily swept away (with interesting quirks!)

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The Tic-Tac-Toe seed:

```
1 int negamaxAB(TicTacToeBoard board,
2     int alpha, int beta, boolean save) {
3     Position[] free = getFreeCells(board);
4     // utility is derived from the number of free cells left
5     if (board.getWinner() != null)
6         alpha = utility(board, free);
7     else if (free.length == 0)
8         alpha = 0; save = false;
9     else for (Position move: free) {
10        TicTacToeBoard copy = board.clone();
11        copy.play(move.row(), move.col(),
12            copy.getTurn());
13        int utility = - (removed) negamaxAB(copy,
14            -beta, -alpha, false save);
15        if (utility > alpha) {
16            alpha = utility;
17            if (save)
18                // save the move into a class instance field
19                chosenMove = move;
20            if (alpha >= beta || beta >= alpha)
21                break;
22        }
23    }
24    return alpha;
25 }
```

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

Loops and recursion are not a problem!

Extant (bad) Java programs can be **improved**

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Future Work

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- Actively searching for consistent bytecode segments during compatibility checks
- Class-level evolution: cross-method crossover, introduction of new methods
- Development of mutation operators
- Applying FINCH to additional hard problems
- Designing an IDE plugin to leverage FINCH for **software engineers**
- Applying FINCH to meta-evolution
- Automatic improvement of existing applications: the realm of **extant software**

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Evolving Game Heuristics



- A row of cells: k black pieces, empty cell, k white pieces.
- Pieces move towards the opposite direction, striving to reverse the initial board situation.
- Pieces can move one step towards the opposite direction, or jump over one complementary-color piece.
- FINCH successfully evolved a `getMove` method, solving the problem consistently and effortlessly.
- Additionally, we had significant progress evolving heuristic evaluation functions for the game of *Connect Four*.

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Evolved program solving the k -empty- k game:

```
Move getMove(Board board) {
    int i = board.findEmpty();
    Piece left1  = board.getPlace(i-1);
    Piece left2  = board.getPlace(i-2);
    Piece right1 = board.getPlace(i+1);
    Piece right2 = board.getPlace(i+2);
    if (left1 == Piece.BLACK && left2 == Piece.WHITE)
        return Move.RIGHT;
    if (right2 == Piece.BLACK)
        return Move.LEFT;
    if (left1 == right2)
        return Move.RIGHT;
    if (right1 == Piece.BLACK)
        return Move.LEFT;
    return Move.RIGHT;
}
```

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