# The calculator problem and the evolutionary synthesis of arbitrary software 

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## Tests



Software


## Outline

- Arbitrary software
- Requirements and ways to meet them
- The calculator problem
- Other problems and prospects


## Arbitrary Software

- OS utilities
- Word processors
- Web browsers
- Accounting systems
- Image processing systems
- Everything


## Arbitrary Software

- May be stateful, with multiple entry points
- May have a variety of interfaces involving a variety of types
- May require arbitrary Turing-computable functions
- Can be specified with behavioral tests


## Evolutionary Computation



## Genetic Programming (GP)

- Evolutionary computing to produce executable computer programs
- Programs are assessed by executing them
- Automatic programming; producing software
- Potential (?): evolve software at all scales, including and surpassing the most ambitious and successful products of human software engineering


## Program Representations

- Lisp-style symbolic expressions (Koza, ...).
- Purely functional/lambda expressions (Walsh, Yu, ...).
- Linear sequences of machine/byte code (Nordin et al., ...).
- Artificial assembly-like languages (Ray,Adami, ...).
- Stack-based languages (Perkis, Spector, Stoffel,Tchernev, ...).
- Graph-structured programs (Teller, Globus, ...).
- Object hierarchies (Bruce,Abbott, Schmutter, Lucas, ...)
- Fuzzy rule systems (Tunstel, Jamshidi, ...)
- Logic programs (Osborn, Charif, Lamas, Dubossarsky, ...).
- Strings, grammar-mapped to arbitrary languages (O'Neill, Ryan, ...).


## Evolvability

The fact that a computation can be expressed in a formalism does not imply that a correct program can be produced in that formalism by a human programmer or by an evolutionary process.

## Requirements

- Represent and evolve arbitrary computable functions on arbitrary types (Push)
- Represent and evolve arbitrary computational architectures (e.g. modules, interfaces; tags and tagged entry points)
- Drive evolution with performance tests (lexicase selection)
- Permit self-adaptation of evolutionary mechanisms (flexible representations, autoconstruction)


## Push

## Push

- Stack-based postfix language with one stack per type
- Types include: integer, float, Boolean, name, code, exec, vector, matrix, quantum gate, [add more as needed]
- Missing argument? NOOP
- Minimal syntax:
program $\rightarrow$ instruction | literal | ( program*)


## Why Push?

- Highly expressive: data types, data structures, variables, conditionals, loops, recursion, modules, ...
- Elegant: minimal syntax and a simple, stackbased execution architecture
- Evolvable
- Extensible
- Supports several forms of meta-evolution


## Sample Push Instructions

| Stack manipulation <br> instructions <br> (all types) | POP, SWAP, YANK, <br> DUP, STACKDEPTH, <br> Math <br> (INTEGER and FLOAT) |
| :--- | :--- |
| Logic (BOOLEAN) | $+,-, /, *,>,<$, <br> MIN, MAX |
| Code manipulation <br> (CODE) | AND, OR, NOT, <br> FROMINTEGER |
| COUOTE, CAR, CDR, CONS, <br> COntrol manipulation <br> (CODE and EXEC) | INSERT, LENGTH, LIST, <br> MEMBER, NTH, EXTRACT |

## Push(3) Semantics

- To execute program $P$ :

1. Push $P$ onto the EXEC stack.
2. While the EXEC stack is not empty, pop and process the top element of the EXEC stack, $E$ :
(a) If $E$ is an instruction: execute $E$ (accessing whatever stacks are required).
(b) If $E$ is a literal: push $E$ onto the appropriate stack.
(c) If $E$ is a list: push each element of $E$ onto the EXEC stack, in reverse order.

## ( 23 INTEGER.* 4.1 5.2 FLOAT.+ TRUE FALSE BOOLEAN.OR )

4.1 5.2
FLOAT.+
TRUE
FALSE
exec code

5.2
FLOAT.+
TRUE
FALSE
FLOAT.+
TRUE
FALSE
BOOLEAN.OR





## Same Results

## ( 23 INTEGER.* 4.1 5.2 FLOAT.+ TRUE FALSE BOOLEAN.OR )

( 2 BOOLEAN.AND 4.1 TRUE INTEGER./ FALSE 35.2 BOOLEAN.OR INTEGER.* FLOAT.+ )

## ( 3.14 CODE.REVERSE CODE.CDR IN IN 5.0 FLOAT.> (CODE.QUOTE FLOAT.*) CODE.IF )

$$
\mathrm{IN}=4.0
$$





FLOAT.>
FLOAT.>
 exec
code






(IN EXEC.DUP (3.13 FLOAT.*) 10.0 FLOAT./)

$$
I N=4.0
$$


exec
code
bool
int
float





## exec









# Auto-simplification 

Loop:
Make it randomly simpler
If it's as good or better: keep it
Otherwise: revert

## Modularity in Software

- Pervasive and widely acknowledged to be essential
- Modules may be functions, procedures, methods, classes, data structures, interfaces, etc.
- Modularity measures include coupling, cohesion, encapsulation, composability, etc.


## Data/Control Structure

- Data abstraction and organization

Data types, variables, name spaces, data structures, ...

- Control abstraction and organization

Conditionals, loops, modules, threads, ...

## Structure via GP (I)

- Specialize GP techniques to directly support human programming language abstractions
- Strongly typed genetic programming
- Module acquisition/encapsulation systems
- Automatically defined functions
- Automatically defined macros
- Architecture altering operations


## Structure via GP (2)

- Specialize GP techniques to indirectly support human programming language abstractions
- Constrain genetic change, or repair after genetic change, to satisfy abstraction syntax
- Map from unstructured genomes to programs in languages that support abstraction (e.g. via grammars)


## Structure via GP (3)

- Develop new program encodings, represented most generally as graphs
- Develop abstraction mechanisms for these representations
- Specialize GP techniques to directly or indirectly support abstraction in these new program encodings


## Structure via GP (4)

- Evolve programs in a minimal-syntax language that nonetheless supports a full range of data and control abstractions
- For example: orchestrate data flows via stacks, not via syntax
- Push


## Tags

## Holland's Tags

- Initially arbitrary identifiers that come to have meaning over time
- Matches may be inexact
- Appear to be present in some form in many different kinds of complex adaptive systems
- Examples range from immune systems to armies on a battlefield
- A general tool for the support of emergent complexity


## Evolving Modular Programs

## With tags

- Include instructions that tag code (modules)
- Include instructions that recall and execute modules by closest matching tag
- If a single module has been tagged then all tag references will recall modules
- The number of tagged modules can grow incrementally over evolutionary time
- Expressive and evolvable


## Tags in Push

- Tags are integers embedded in instruction names
- Instructions like tag.exec. 123 tag values
- Instructions like tagged. 456 recall values by closest matching tag
- If a single value has been tagged then all tag references will recall (and execute) values
- The number of tagged values can grow incrementally over evolutionary time


# Calculator Execution Architecture 

## Lexicase Selection

- Each parent is selected by filtering the entire population, one one case at a time (in random order), keeping only the elite at each stage
- Useful for "modal" problems, which require qualitatively different responses to different inputs
- Useful for "uncompromising" problems, in which solutions must be optimal on each case
- All comparisons are "within case," so may be useful whenever cases are non-comparable


## Lexicase Selection

## Initialize:

Candidates $=$ the entire population
Cases $=$ a list of all of the test cases in random order

## Loop:

Candidates $=$ the subset of Candidates with exactly the best performance of any current candidate for the first case in Cases

If Candidates or Cases contains just a single element then return a randomly selected individual from Candidates

Otherwise remove the first case from Cases and go to Loop

## Finite Algebras

| $\mathbf{A}_{1} *$ | 0 | 1 | 2 |
| ---: | :---: | :---: | :---: |
| 0 | 2 | 1 | 2 |
| 1 | 1 | 0 | 0 |
| 2 | 0 | 0 | 1 |


| $\mathbf{A}_{2} *$ | 0 | 1 | 2 |
| ---: | :---: | :---: | :---: |
| 0 | 2 | 0 | 2 |
| 1 | 1 | 0 | 2 |
| 2 | 1 | 2 | 1 |

## AI Mal'cev Term

| Selection | Successes | CE | MBF |
| :--- | ---: | ---: | ---: |
| Tournament Size 2 | 35 | 532,000 | 0.75 |
| Tournament Size 3 | 43 | 420,000 | 0.70 |
| Tournament Size 4 | 31 | 440,000 | 0.75 |
| Tournament Size 5 | 22 | 616,000 | 0.77 |
| Tournament Size 6 | 25 | 750,000 | 0.90 |
| Tournament Size 7 | 23 | 403,000 | 0.92 |
| Tournament Size 8 | 26 | 464,000 | 0.94 |
| Tournament Size 9 | 21 | 550,000 | 1.06 |
| Lexicase | 94 | 90,000 | 0.05 |

## A2 Mal'cev Term

| Selection | Successes | CE | MBF |
| :--- | ---: | ---: | ---: |
| Tournament Size 3 | 7 | $3,780,000$ | 1.50 |
| Tournament Size 4 | 5 | $3,648,000$ | 1.50 |
| Tournament Size 5 | 8 | $2,052,000$ | 1.51 |
| Tournament Size 6 | 9 | $1,921,000$ | 1.45 |
| Tournament Size 7 | 3 | $4,131,000$ | 1.59 |
| Tournament Size 8 | 9 | 990,000 | 1.64 |
| Tournament Size 9 | 10 | $1,356,000$ | 1.60 |
| Lexicase | 75 | 208,000 | 0.25 |

## The Digital Multiplier Problem

- Evolve a digital circuit to multiply two binary numbers
- $n$-bit digital multiplier: $2 \times n$ bits $\rightarrow 2 n$ bits
- Multiple outputs
- Scalable
- Recommended as a GP benchmark problem (McDermott, et al 2012, White et al 2013)


## 3-bit Digital Multiplier

Boolean Stack and, or, xor, invert_first_then_and, dup, swap, rot
Input / Output in0, ..., in2n, out0, ..., out2n

| Selection | Successes | MBF |
| :--- | ---: | ---: |
| Tournament Size 7 | 0 | 0.24 |
| Lexicase | 100 | 0 |

## Factorial

| Boolean Stack | and, dup, eq, frominteger, not, or, pop, rot, <br> swap |
| :--- | :--- |
| Integer Stack | add, div, dup, eq, fromBoolean, <br> greaterThan, lessThan, mod, mult, pop, rot, <br> sub, swap <br> dup, eq, if, noop, pop, rot, swap, when, k, <br> Exec Stack <br> s, y |
| Input | in |
| Constants | 0,1 |


| Selection | Successes | MBF |
| :--- | ---: | ---: |
| Tournament Size 7 | 0 | 74,545 |
| Lexicase | 61 | 28,980 |

## Calculator Test Cases

- digit-entry-tests
- digit-entry-pair-tests
- single-digit-math-tests
- ;single-digit-incomplete-math-tests
- ;single-digit-chained-math-tests
- ;division-by-zero-tests
- ;double-digit-float-entry-tests


## Autoconstructive Evolution

- Individuals make their own children
- Agents thereby control their own mutation rates, sexuality, and reproductive timing
- The machinery of reproduction and diversification (i.e., the machinery of evolution) evolves
- Radical self-adaptation


## ULTRA

## Uniform Variation

- All genetic material that a child inherits should be $\approx$ likely to be mutated
- Parts of both parents should be $\approx$ likely to appear in children (at least if they are $\approx$ in size), and to appear in a range of combinations
- Should be applicable to genomes of varying size and structure


## Why Uniformity?

- No hiding from mutation
- All parts of parents subject to variation and recombination
- Biological genetic variation, while not fully uniform, has uniformity properties that prevent some of the problems we see in
GP; e.g. just having more genes doesn't generally "protect" genes any of them


## Prior Work

- Point mutations or "uniform crossovers" that replace/swap nodes but only in restricted ways; cannot change structure, has depth biases (McKay et al, I995; Page et al, I998; Poli and Langdon, I998; Poli and Page, 2000; Semenkin and Semenkina, 2012)
- Uniform mutation via size-based numbers of tree replacements; depth biases, little demonstrated benefit (McKay et al, I995;Van Belle and Ackley, 2002)


## ULTRA

- Achieve uniformity by treating genomes as linear sequences, even if they are hierarchically structured
- Repair after transform to ensure structural validity


## The ULTRA Operator

- Uniform Linear Transformation with Repair and Alternation
- Linearize 2 parents, treating "(" and ")" as ordinary tokens
- Start at the beginning of one parent and copy tokens to the child, switching parents stochastically (according to the alternation rate, and subject to an alignment deviation)
- Post-process with uniform mutation (according to a mutation rate) and repair


## Parents:

$$
\begin{aligned}
& \text { ( a b ( c ( d ) ) e ( f g ) ) } \\
& \text { ( } 1 \text { ( } 2 \text { ( } 34 \text { ) } 5 \text { ) } 6 \text { ) }
\end{aligned}
$$

## Result of alternation:

$$
(\mathrm{a} b 2(34 \mathrm{~d}) \text { ) } 6 \text { ) }
$$

## Result of repair:

$$
(\mathrm{a}(\mathrm{~b} 2(34 \mathrm{~d})) 6)
$$

## ULTRA on the

## bioavailability problem





- Bowling
- WC
- generative tests

