Evolving evolution, in computers, to solve some of the world's hardest problems

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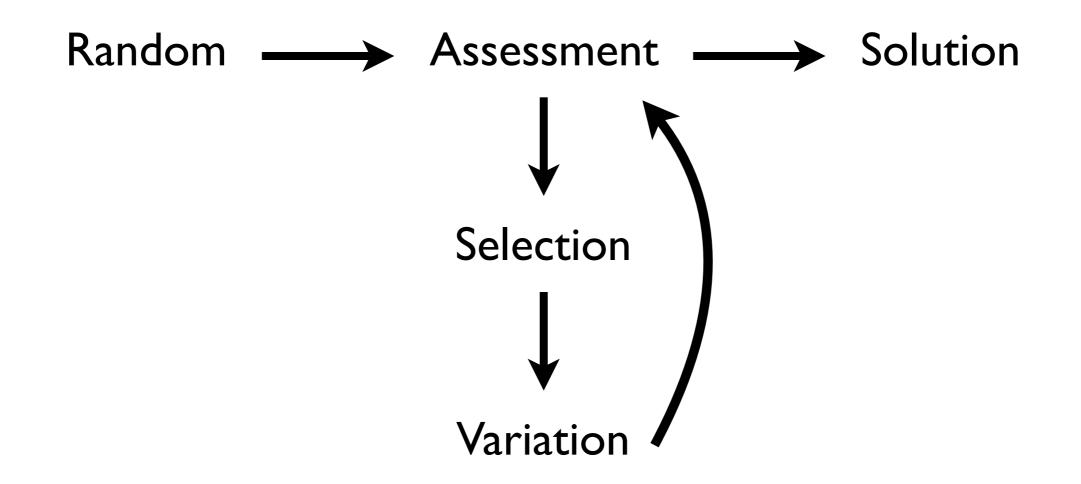
Outline

- Evolutionary computation
- Evolved solutions
- Evolving evolution



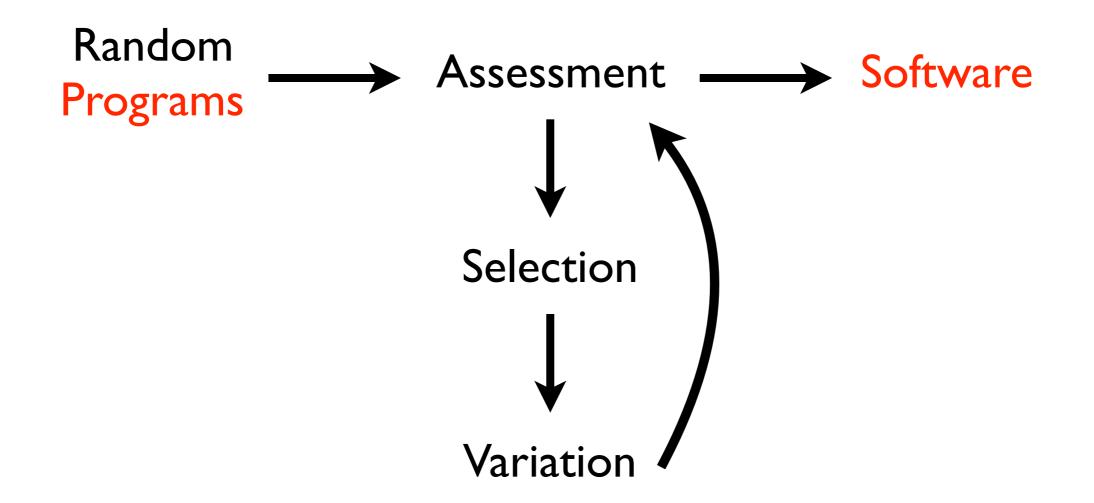
"Coral reef" by Les Chatfield is licensed under CC BY 2.0

Evolutionary Computation



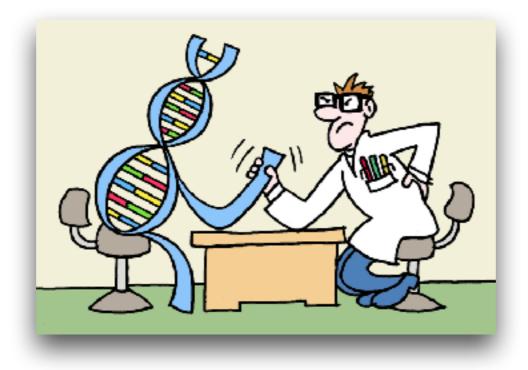
Evolving Lego Bridges

Genetic Programming



Analyzing a Decade of Human-Competitive ("HUMIE") Winners: What Can We Learn?

Karthik Kannappan, Lee Spector, Moshe Sipper, Thomas Helmuth, William Lacava, Jake Wisdom, Omri Bernstein



- Competition for evolutionary computation results
- Up to \$10,000 in cash prizes
- Held annually since 2004

Humies Criteria

The result was **patented as an invention** in the past is an improvement over a patented invention or would qualify today as a patentable new invention.

The result is equal to or better than a result that was accepted as a **new scientific result** at the time when it was published in a peer-reviewed scientific journal.

The result is equal to or better than a result that was placed into a database or archive of results maintained by an **internationally recognized panel of scientific experts**.

The result is **publishable in its own right** as a new scientific result independent of the fact that the result was mechanically created.

The result is equal to or better than the **most recent human-created** solution to a longstanding problem for which there has been a succession of increasingly better human-created solutions.

The result is equal to or better than a result that was considered an **achievement in its field** at the time it was first discovered.

The result solves a problem of **indisputable difficulty** in its field.

The result holds its own or wins a regulated **Competition involving human**

Contestants (in the form of either live human players or human-written computer programs).

Humies Winners

- Application areas: antennas, biology, chemistry, computer vision, electrical engineering, electronics, games, image processing, mathematics, mechanical engineering, medicine, operations research, optics, optimization, photonics, physics, planning, polymers, quantum computing, security, software engineering
- Winningest technique: genetic programming
- Winningest problem type: design





VOLUME 22

SUMMER 2008

NUMBER 3

SPECIAL ISSUE

Genetic Programming for Human-Competitive Designs

Guest Editor

LEE SPECTOR

Evolution, the Designer

And now, digital evolution

The Boston Globe

By Lee Spector | August 29, 2005

RECENT developments in computer science provide new perspective on "intelligent design," the view that life's complexity could only have arisen through the hand of an intelligent designer. These developments show that complex and useful designs can indeed emerge from random Darwinian processes.

"Darwinian evolution is itself a designer worthy of significant respect, if not religious devotion."

Yavalath

Yavalath is an abstract board game for two or three players, invented by a computer program called LUDI. It has an easy rule set that any player can pick up immediately, but which produces surprisingly tricky emergent play.

Yavalath is available from <u>nestorgames</u>, making it the first — and still only — computer-generated game to be commercially published, together with its sister game <u>Pentalath</u>.

In October 2011, Yavalath was ranked in the top #100 abstract board games ever invented on the <u>BoardGameGeek</u> database. This helped it win the GECCO "<u>Humies</u>" gold medal for human-competitive results in evolutionary computation for 2012.

Here is a Yavalath article in the November 2013 issue of Bitcoin magazine.

Rules

The board starts empty.

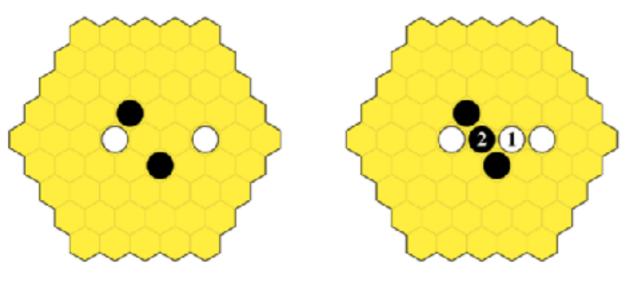
Two players take turns adding a piece of their colour to an empty cell.

Win by making a line-of-4 (or more) pieces of your colour. Lose by making a line-of-3 pieces of your colour beforehand. Draw if the board otherwise fills up.

No, players are not allowed to pass.

Tactics and Strategy

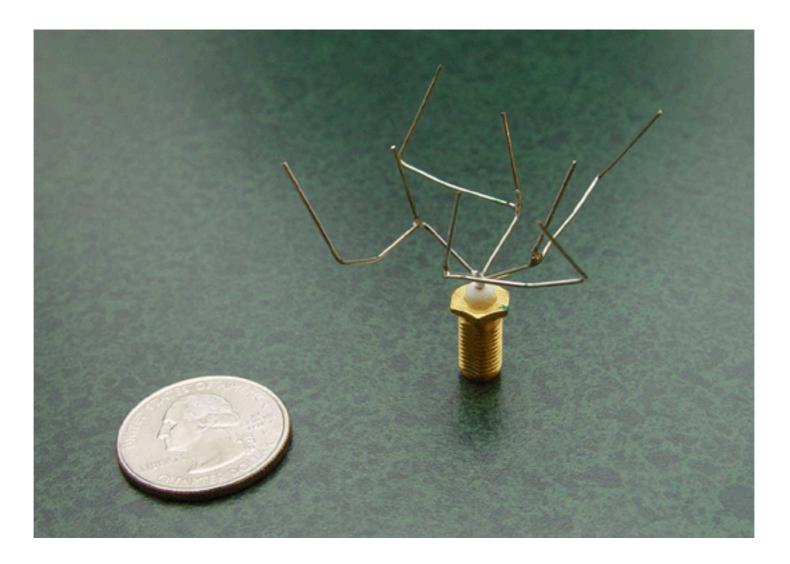
The key tactical play in Yavalath is the forcing move, as shown below. White move 1 forces Black to lose with the blocking move 2.





Humies Gold Medal, 2012

Evolved Antenna

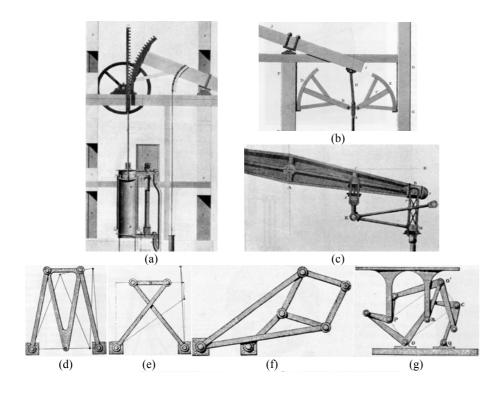


NASA Space Technology 5 Mission, Lohn, Hornby, and Linden

Humies Gold Medal, 2004

Line-Drawing Mechanism

- Without reference to an existing straight line
- Human-competitive results; challenged world's greatest inventors for a century (spanning 18th and 19th)



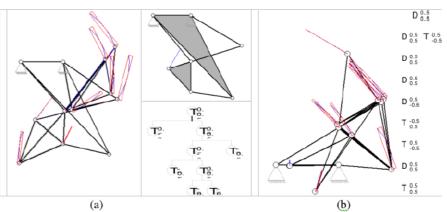
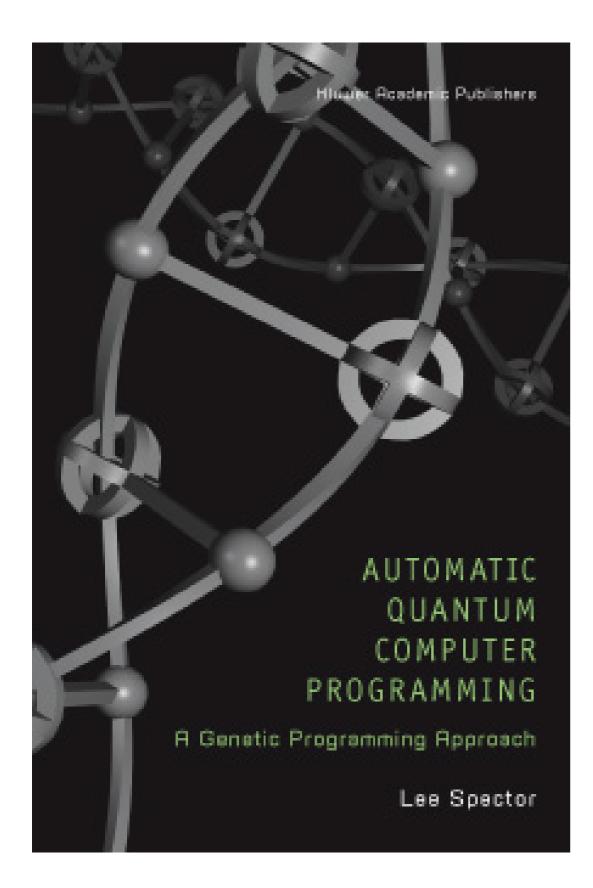


Fig. 10. Two Evolved mechanisms and their tree representations (a) Linearity 1:12819; The simplified equivalent shown top right, and (b) Linearity 1:4979.

Lipson, H. 2004. How to Draw a Straight Line Using a GP: Benchmarking Evolutionary Design Against 19th Century Kinematic Synthesis. GECCO-2004.

Humies Silver Medal, 2012



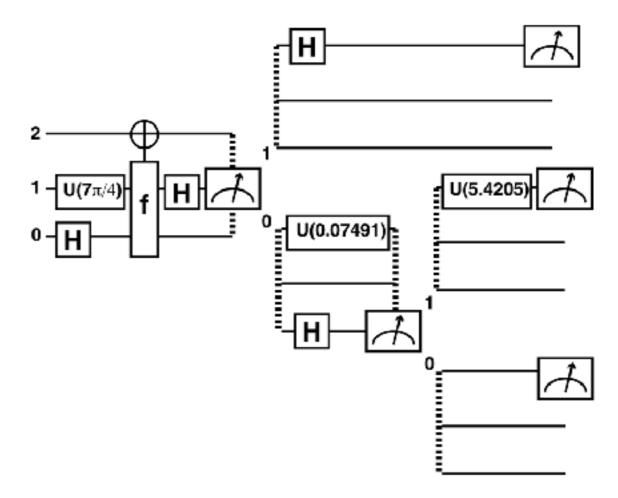


Figure 8.11. A gate array diagram for an evolved solution to the AND/OR oracle problem. The gate marked "f" is the oracle. The sub-diagrams on the right represent the possible execution paths following the intermediate measurements.

Humies Gold Medal, 2004

Genetic Programming for Finite Algebras

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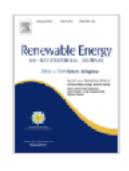
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Humies Gold Medal, 2008



Renewable Energy

Volume 87, Part 2, March 2016, Pages 892–902



Automatic identification of wind turbine models using evolutionary multiobjective optimization

William La Cava^{a,} ^I, ^{III}, Kourosh Danai^a, Lee Spector^b, Paul Fleming^c, Alan Wright^c, Matthew Lackner^a **Show more**

https://doi.org/10.1016/j.renene.2015.09.068

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Highlights

- Accurate, succinct models of wind turbine dynamics are identified from normal operating data.
- · A novel evolutionary multi-objective optimization system is described.
- The proposed method produces physically meaningful models without prior knowledge of the system.
- The method is bench-marked against other modeling techniques.

Genetic programming representations for multi-dimensional feature learning in biomedical classification

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Abstract. We present a new classification method that uses genetic programming (GP) to evolve feature transformations for a deterministic, distanced-based classifier. This method, called M4GP, differs from common approaches to classifier representation in GP in that it does not enforce arbitrary decision boundaries and it allows individuals to produce multiple outputs via a stack-based GP system. In comparison to typical methods of classification, M4GP can be advantageous in its ability to produce readable models. We conduct a comprehensive study of M4GP, first in comparison to other GP classifiers, and then in comparison to six common machine learning classifiers. We conduct full hyper-parameter optimization for all of the methods on a suite of 16 biomedical data sets, ranging in size and difficulty. The results indicate that M4GP outperforms other GP methods for classification. M4GP performs competitively with other machine learning methods in terms of the accuracy of the produced models for most problems. M4GP also exhibits the ability to detect epistatic interactions better than the other methods.

Keywords: genetic programming, feature learning, classification

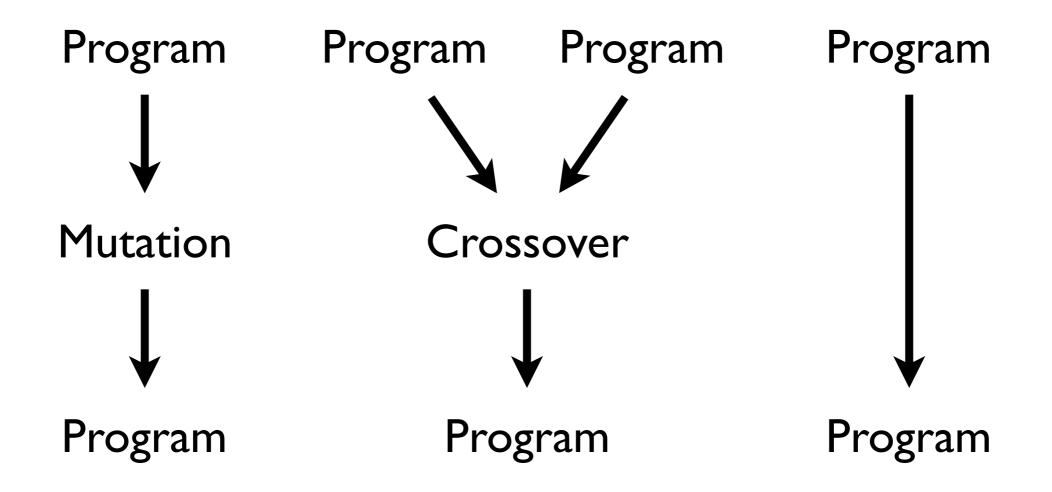
Intro Programming

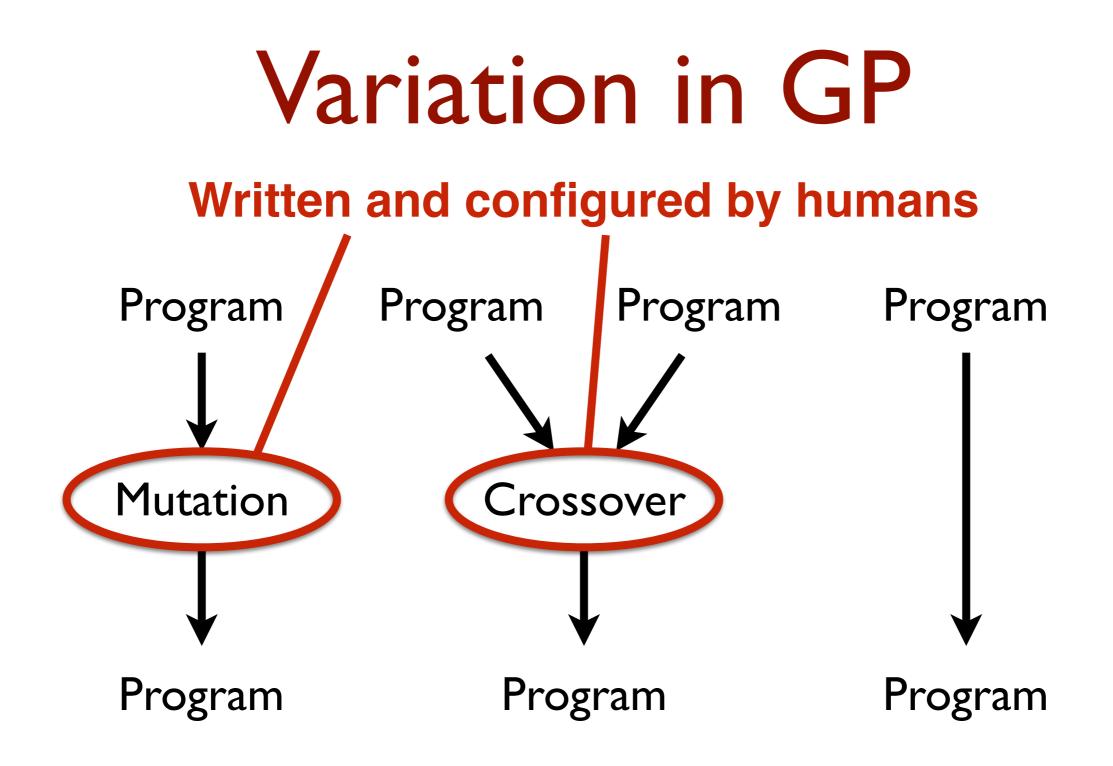
Number IO, Small or Large, For Loop Index, Compare String Lengths, Double Letters, Collatz Numbers, Replace Space with Newline, String Differences, Even Squares, Vallis Pi, String Lengths Backwards, Last Index of Zero, Vector Average, Count Odds, Mirror Image, Super Anagrams, Sum of Squares, Vectors Summed, X-Word Lines, Pig Latin, Negative to Zero, Scrabble Score, Word Stats, Checksum, Digits, Grade, Median, Smallest, Syllables 7. Replace Space with Newline (P 4.3) Given a string input, print the string, replacing spaces with newlines. Also, return the integer count of the non-whitespace characters. The input string will not have tabs or newlines.

- Multiple types, looping, multiple tasks
- Simplified solution:

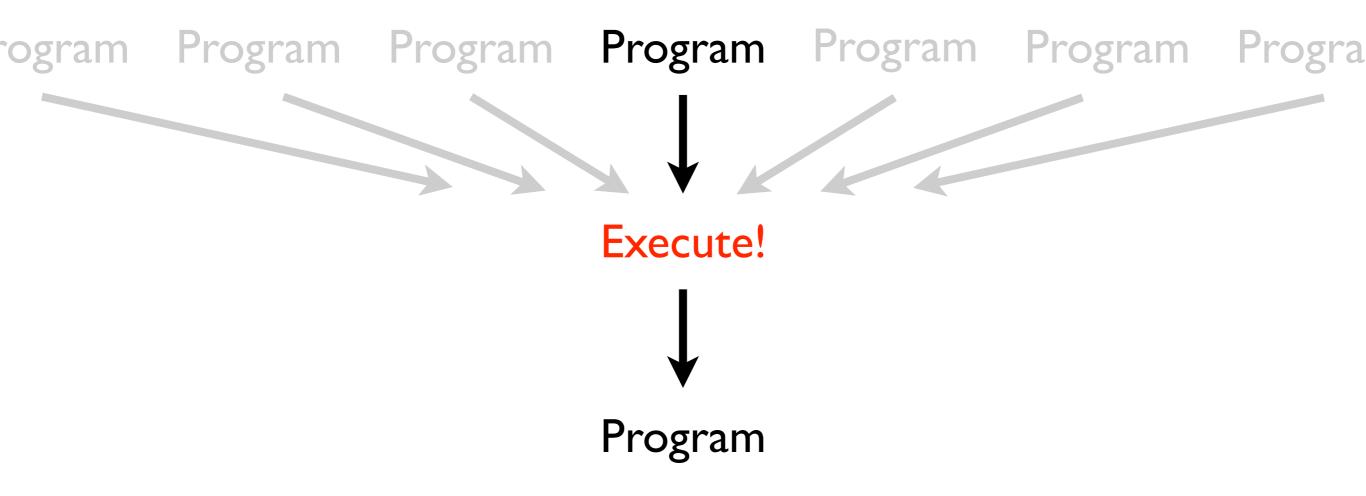
(\space char_dup exec_dup in1 \newline string_replacechar
print_string string_removechar string_length)

Variation in GP

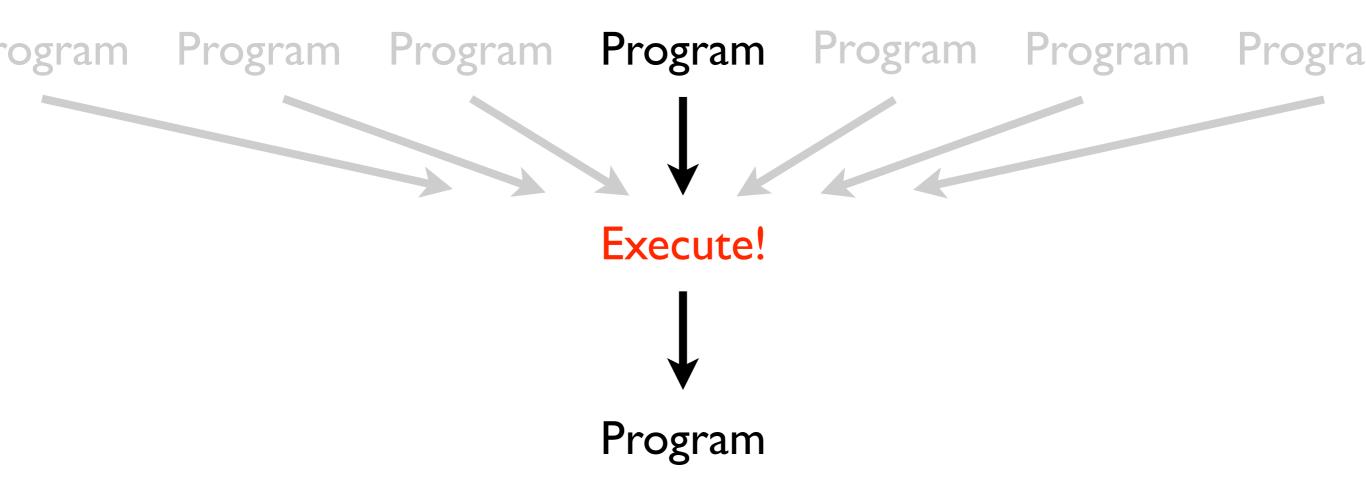




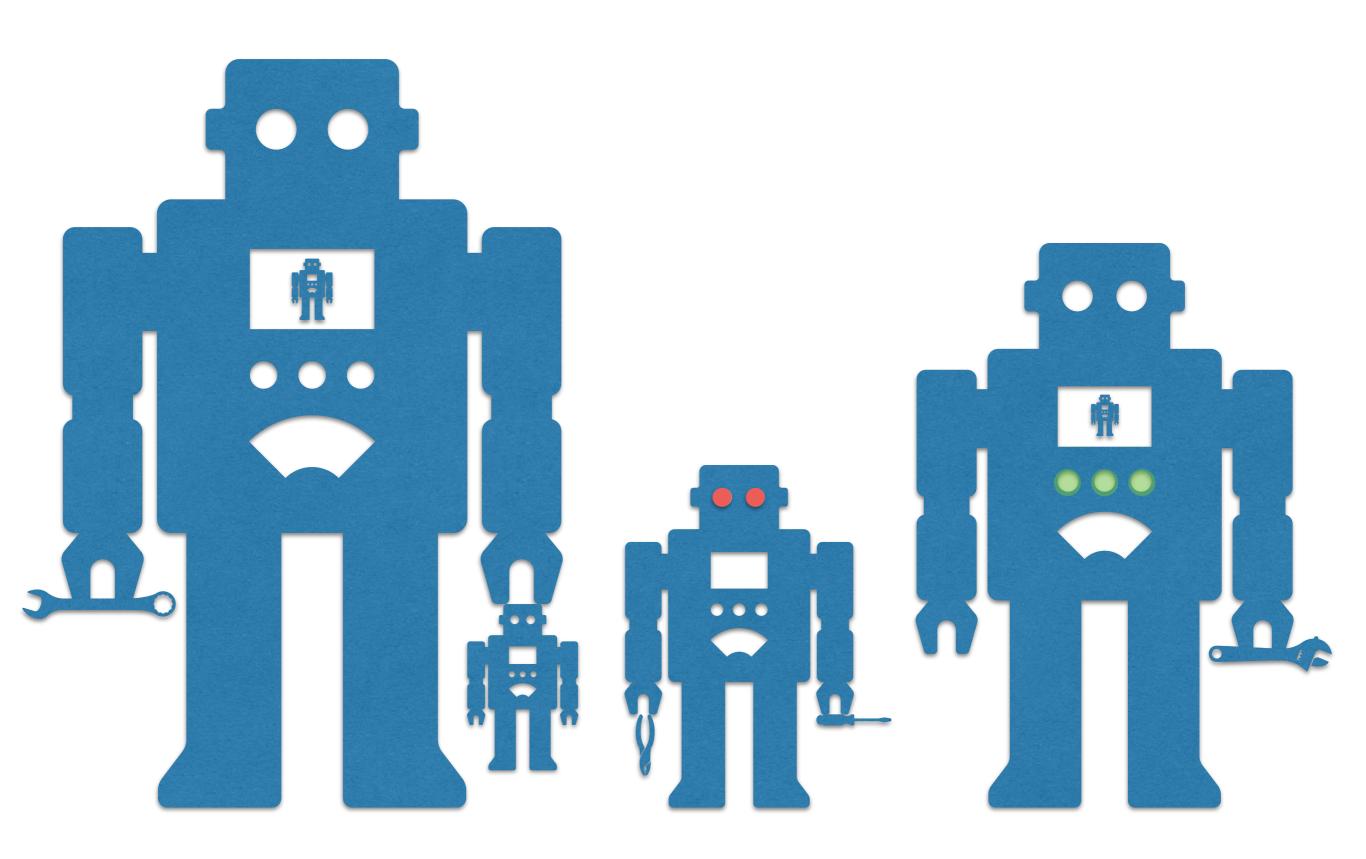
Autoconstruction



Autoconstruction



A bit more complicated when genomes distinguished from programs



Autoconstructive Evolution

- Evolve evolution while evolving solutions
- How? Individuals produce and vary their own children, with methods that are subject to variation
- Requires understanding the evolution of variation
- Hope: May produce EC systems more powerful than we can write by hand

Needed for Evolution²

- Diversity: Individuals vary
- Diversification: Individuals produce descendants that vary, in various ways
- Recursive Variance: Individuals produce descendants that vary in the ways that they vary their offspring

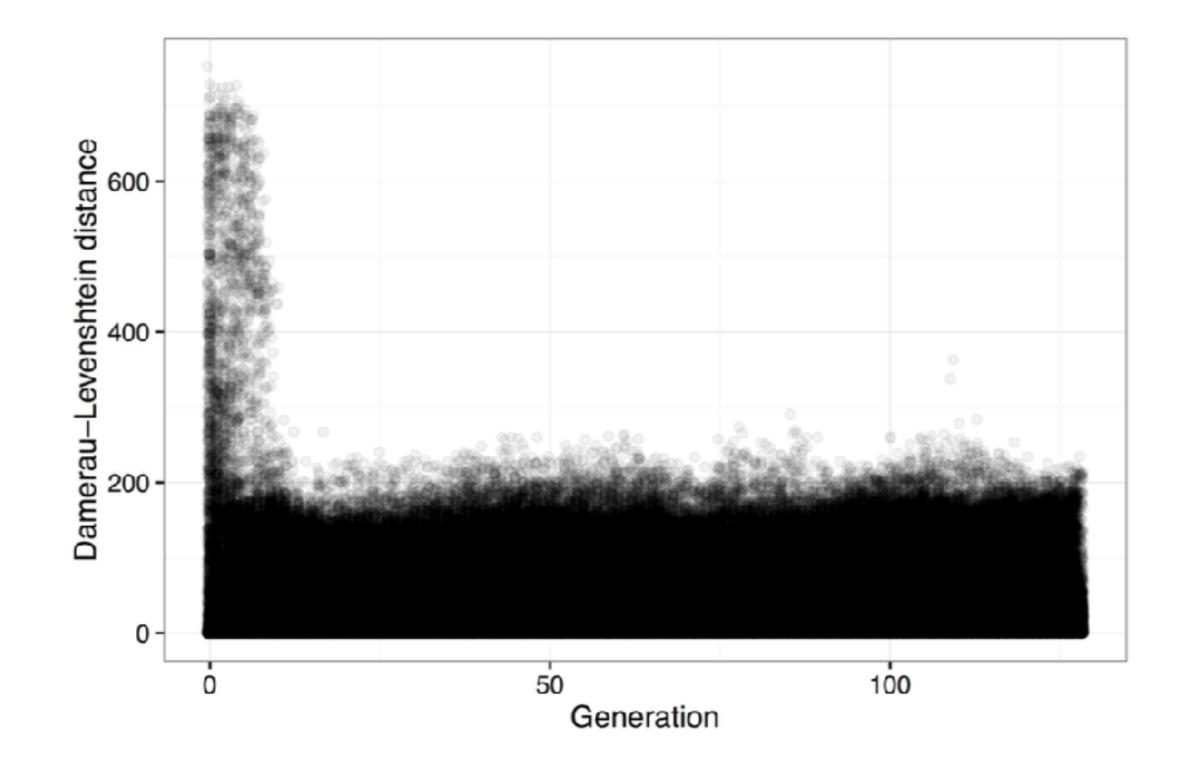


Figure 1: DL-distances between parent and child during a single non-autoconstructive run of GP on the Replace Space With Newline problem

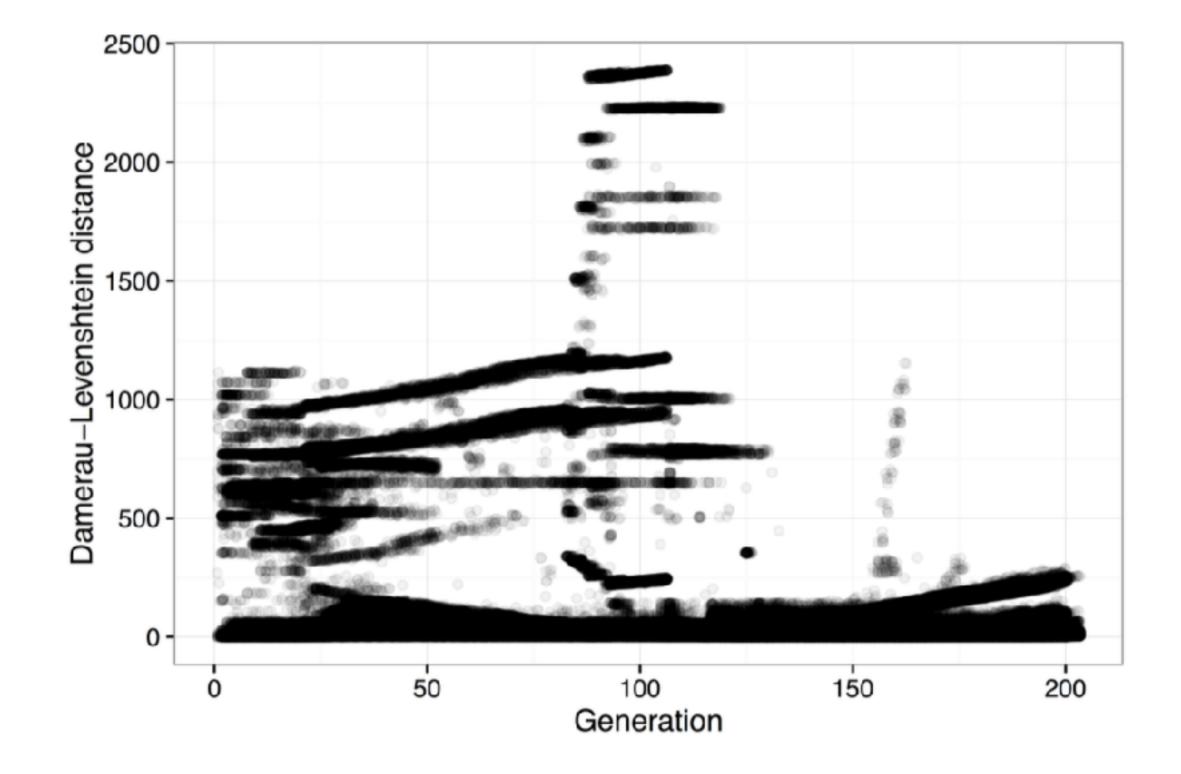


Figure 3: DL-distances between parent and child during a single autoconstructive run of GP on the Replace Space With Newline problem

Rivaling Ordinary GP

- Autoconstructive evolution can succeed as much and as fast as non-autoconstructive evolution
- In 20 runs in one configuration, 75% success within 300 generations on Replace Space With Newline (100% by generation 628)
- Surprising!

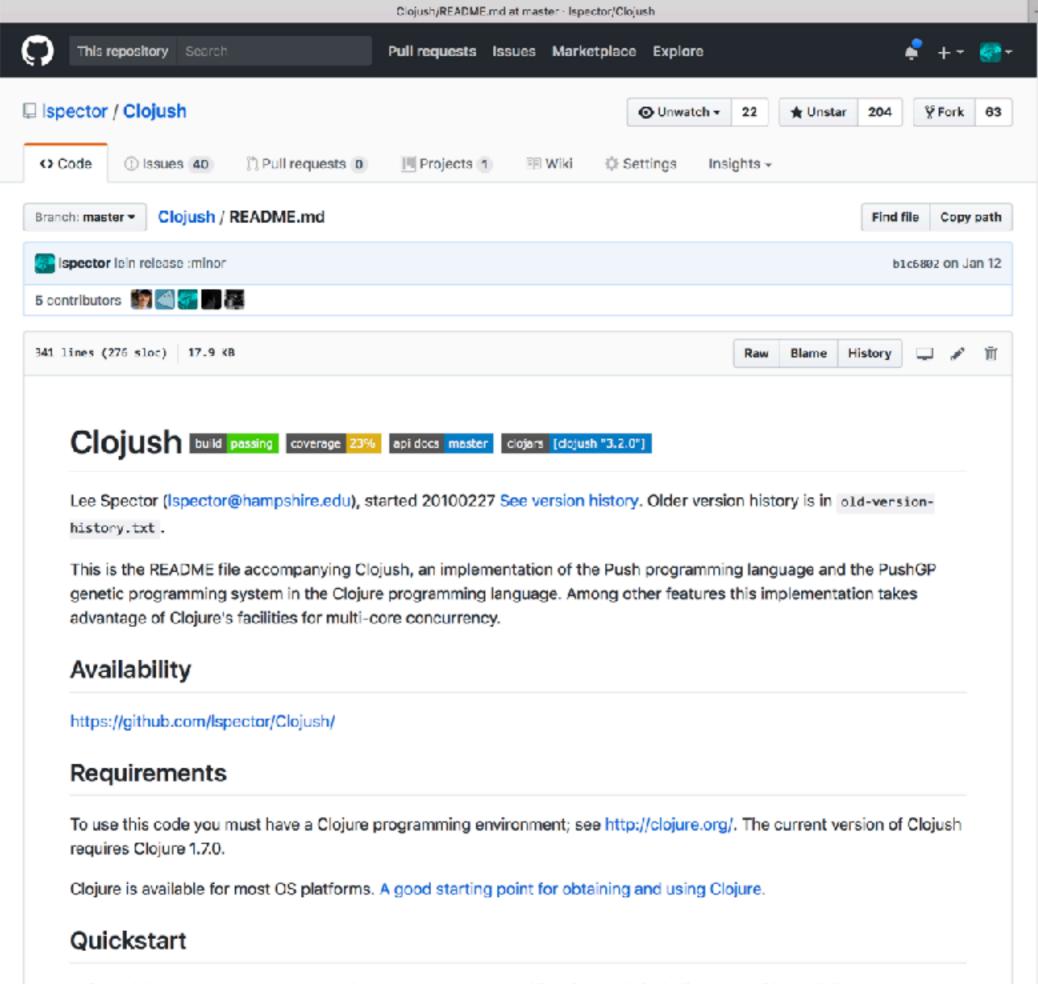
- 8. String Differences (P 4.4) Given 2 strings (without whitespace) as input, find the indices at which the strings have different characters, stopping at the end of the shorter one. For each such index, print a line containing the index as well as the character in each string. For example, if the strings are "dealer" and "dollars", the program should print:
 - 1 e o 2 a l 4 e a

Extending GP's Reach

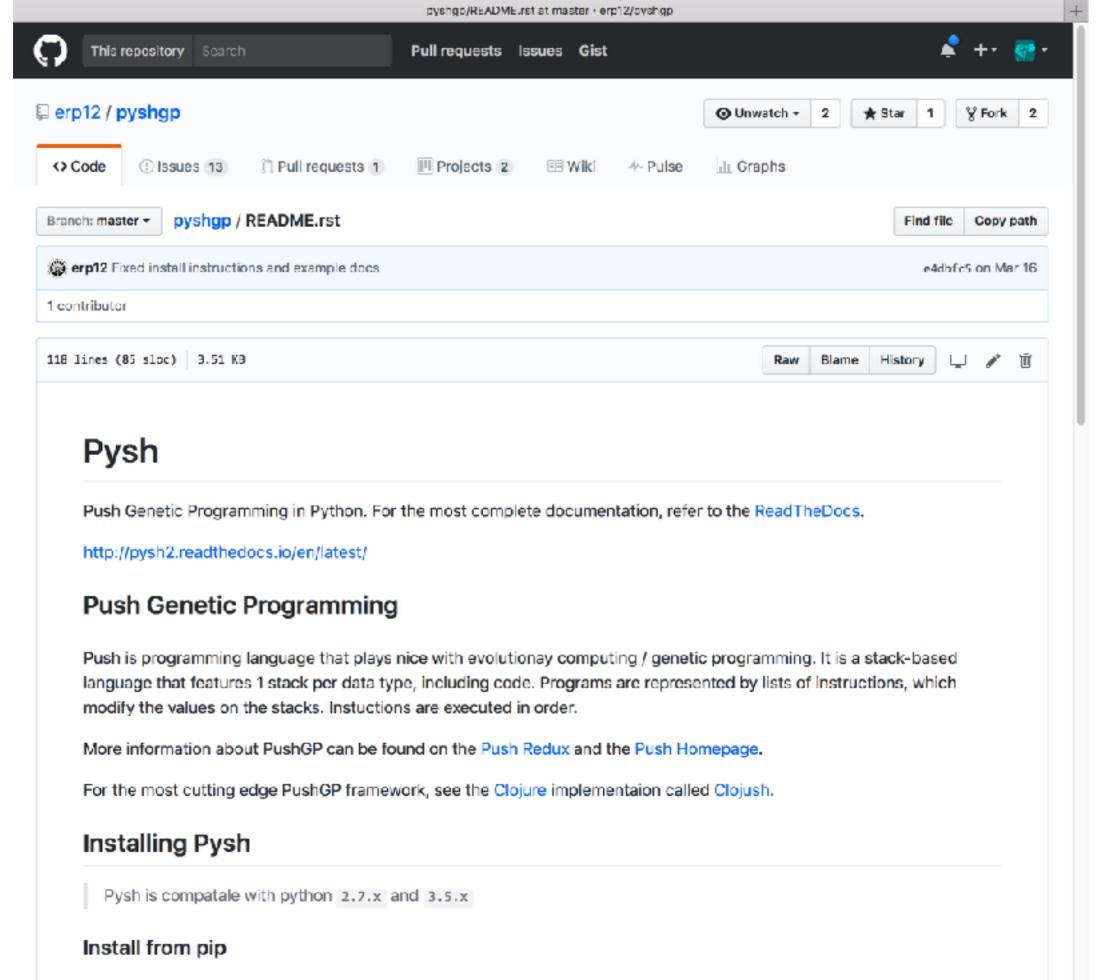
- Without autoconstruction, string difference not yet solved by GP, despite many efforts/configurations
- 3 autoconstructive solutions so far

Prospects

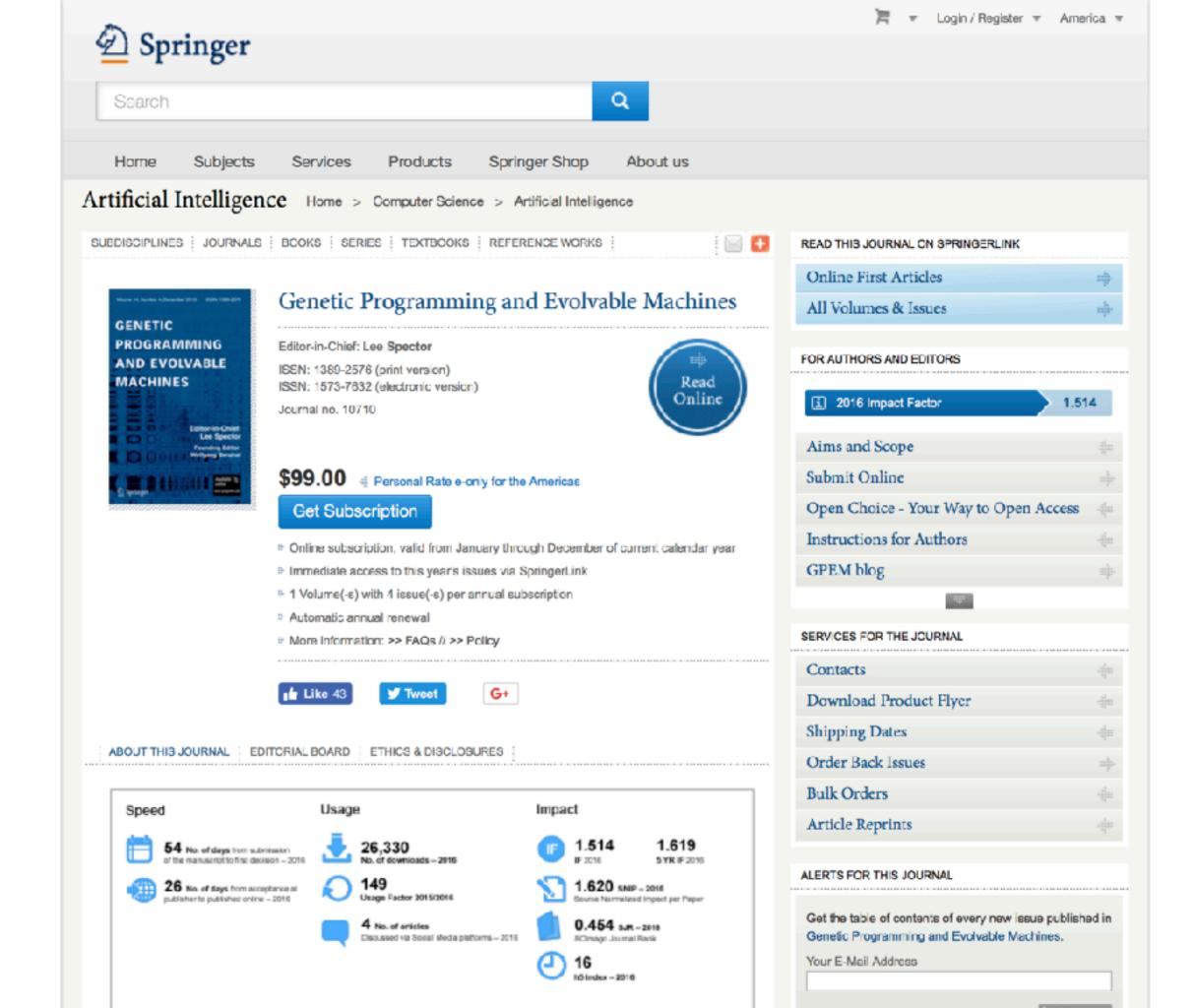
- Evolutionary computation is already solving important, hard problems
- If it can be applied to itself, to evolve as it runs, then it will be able to solve harder problems
- We are beginning to see how this might work
- Help would be appreciated!



Using Leiningen you can run an example from the OS command line (in the Clojush directory) with a call like:



Coming whith first beta release of pyshgp. Check the Roadmap to get a sense of how far off this is.



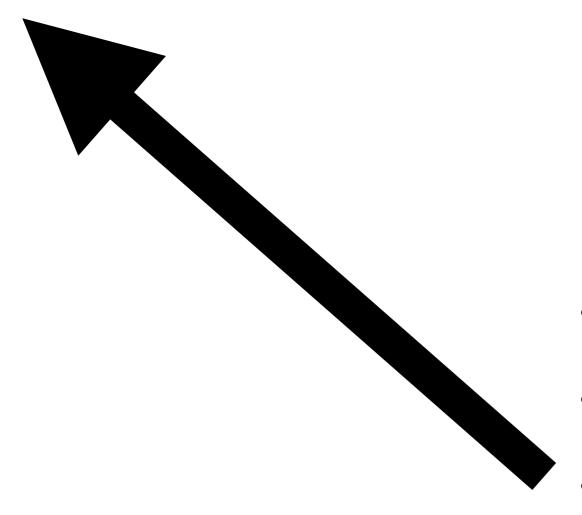


Technology that we don't understand



- Technology that we don't understand
- Like children and other life forms





- pushlanguage.org
- <u>http://hampshire.edu/lspector</u>
- My office

Thanks

- Members of the Hampshire College Computational Intelligence Lab
- This material is based upon work supported by the National Science Foundation under Grants No. 1017817, 1129139, and 1331283. Any opinions, findings, and conclusions or recommendations expressed in this publication are those of the authors and do not necessarily reflect the views of the National Science Foundation.