

Evolutionary Computation

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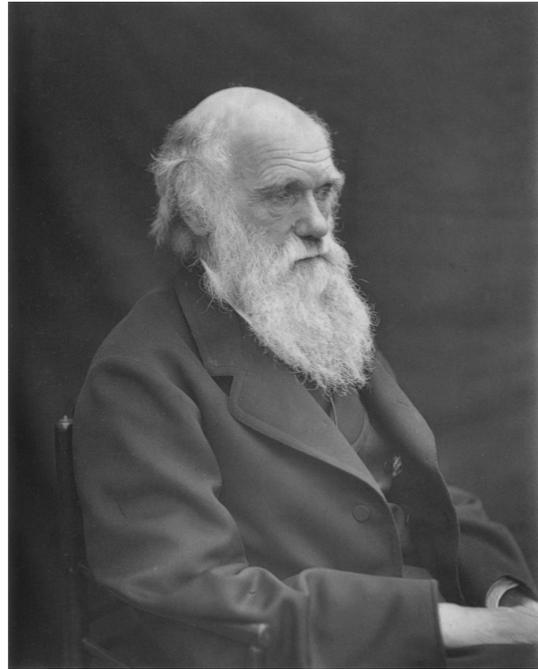


Outline

- What is evolutionary computation?
- What can it do?
- Improving it
- Connections

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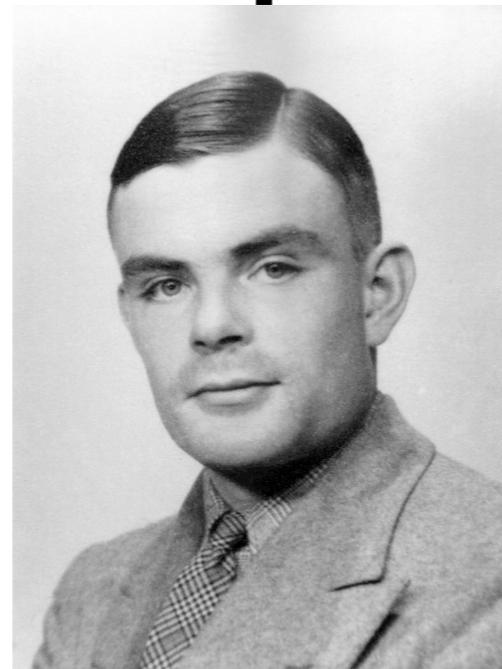
- * **What is evolutionary computation?**
- What can it do?
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Charles
Darwin



Ada
Lovelace



Alan
Turing

https://en.wikipedia.org/wiki/Charles_Darwin

<https://www.cnn.com/ampstories/tech/meet-ada-lovelace-the-first-computer-programmer>

<https://www.britannica.com/biography/Alan-Turing>

EvoBio & CompSci

- Bioinformatics
- Modeling & simulation
- Artificial life
- Evolutionary algorithms

EvoBio & CompSci

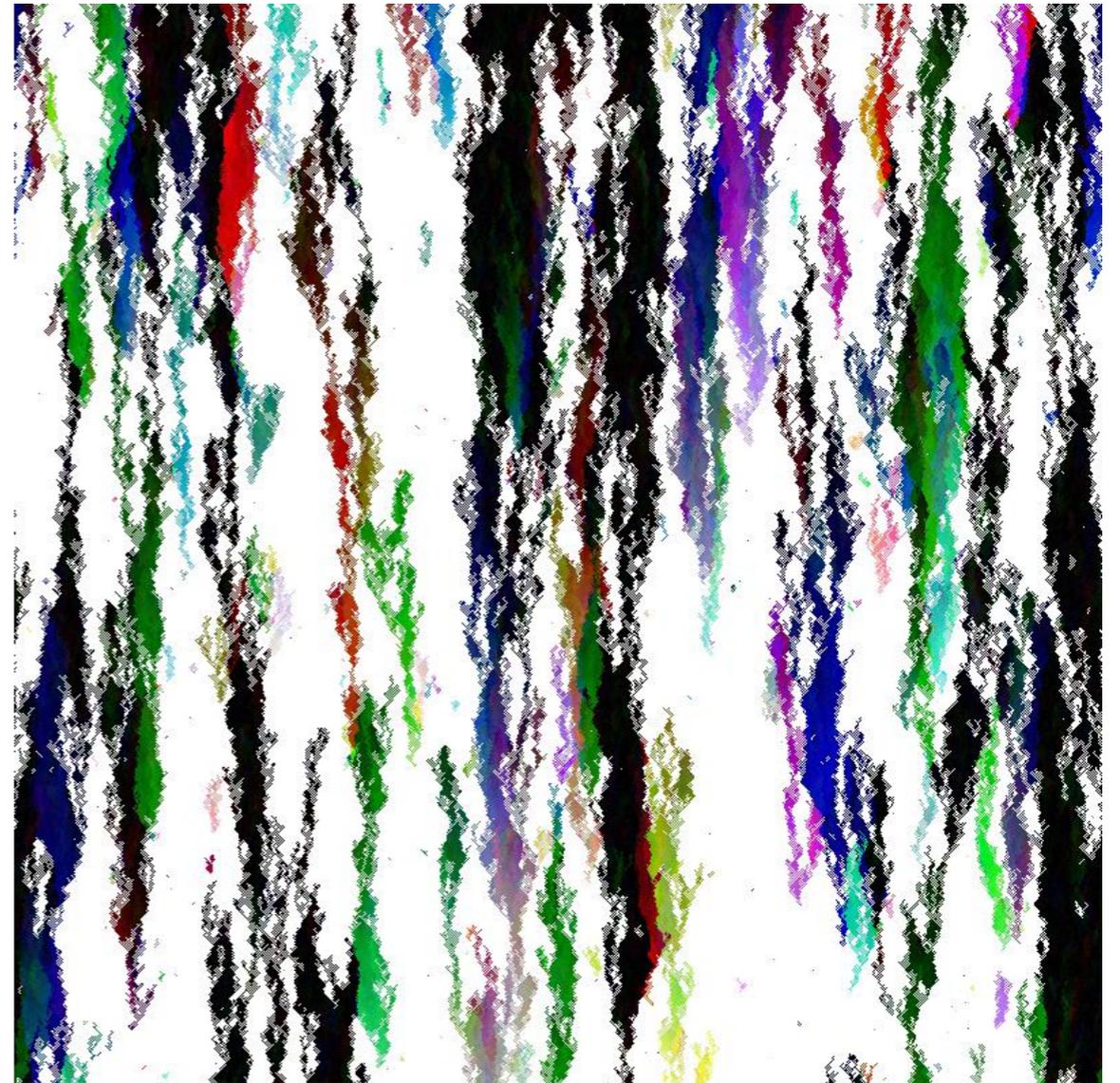
* **Bioinformatics**

- Modeling & simulation
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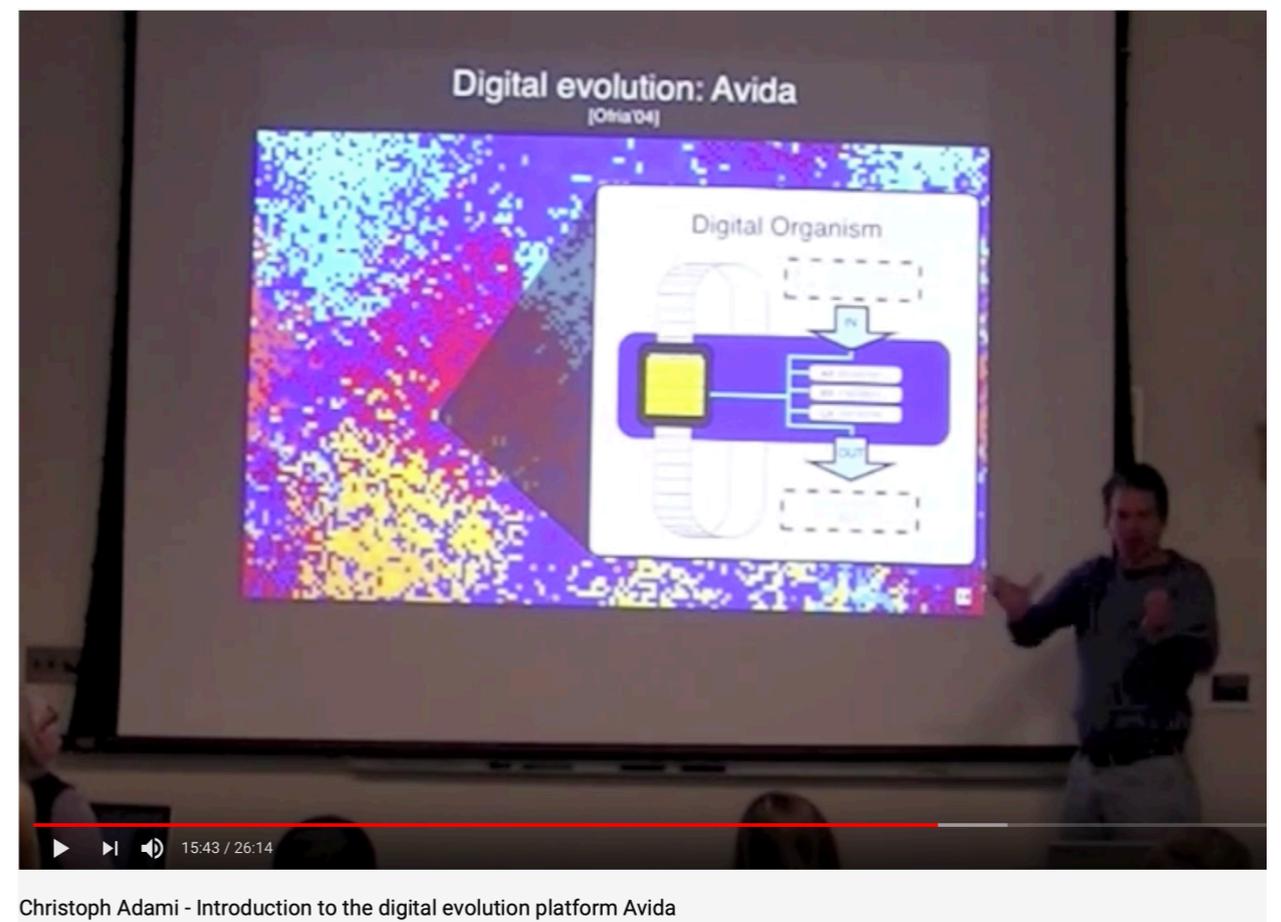
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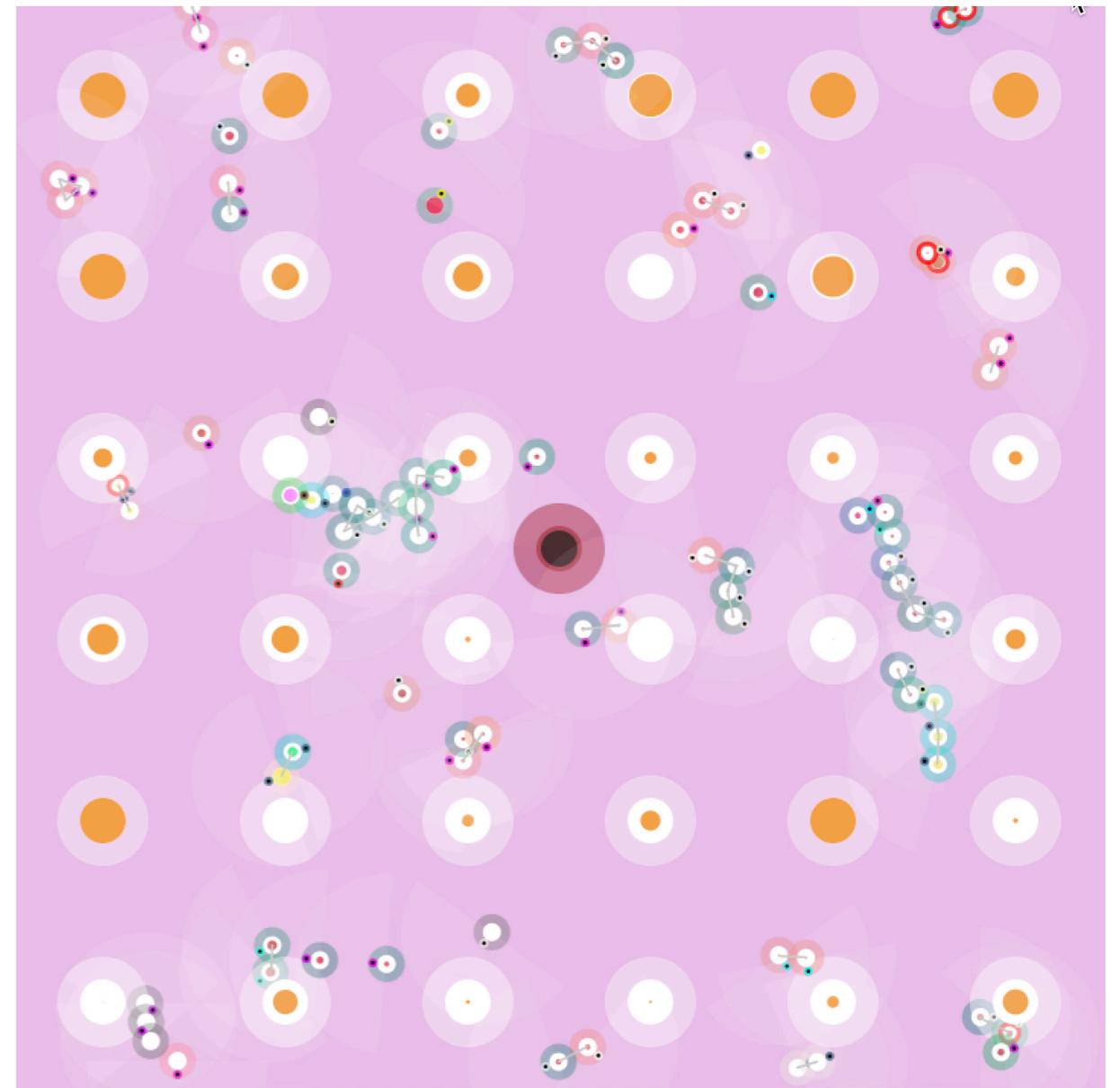
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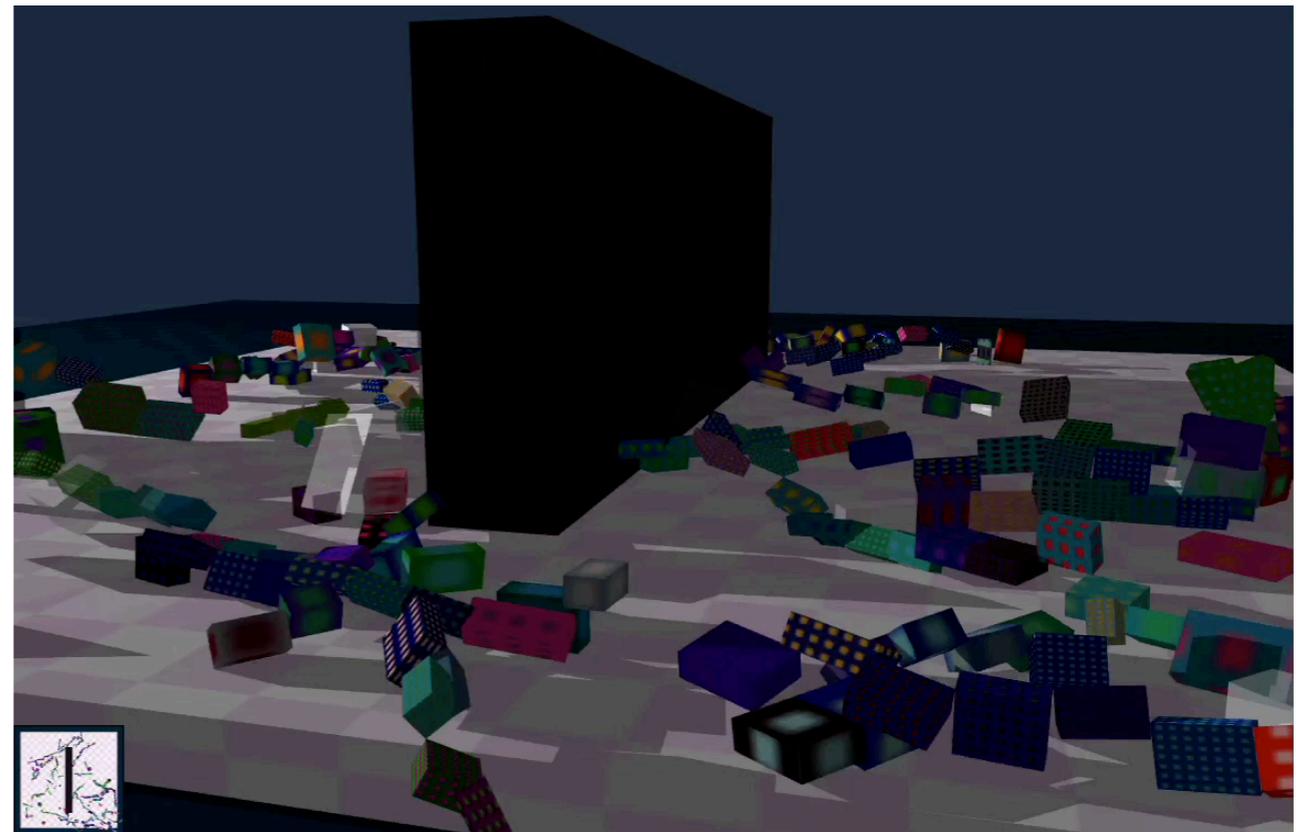
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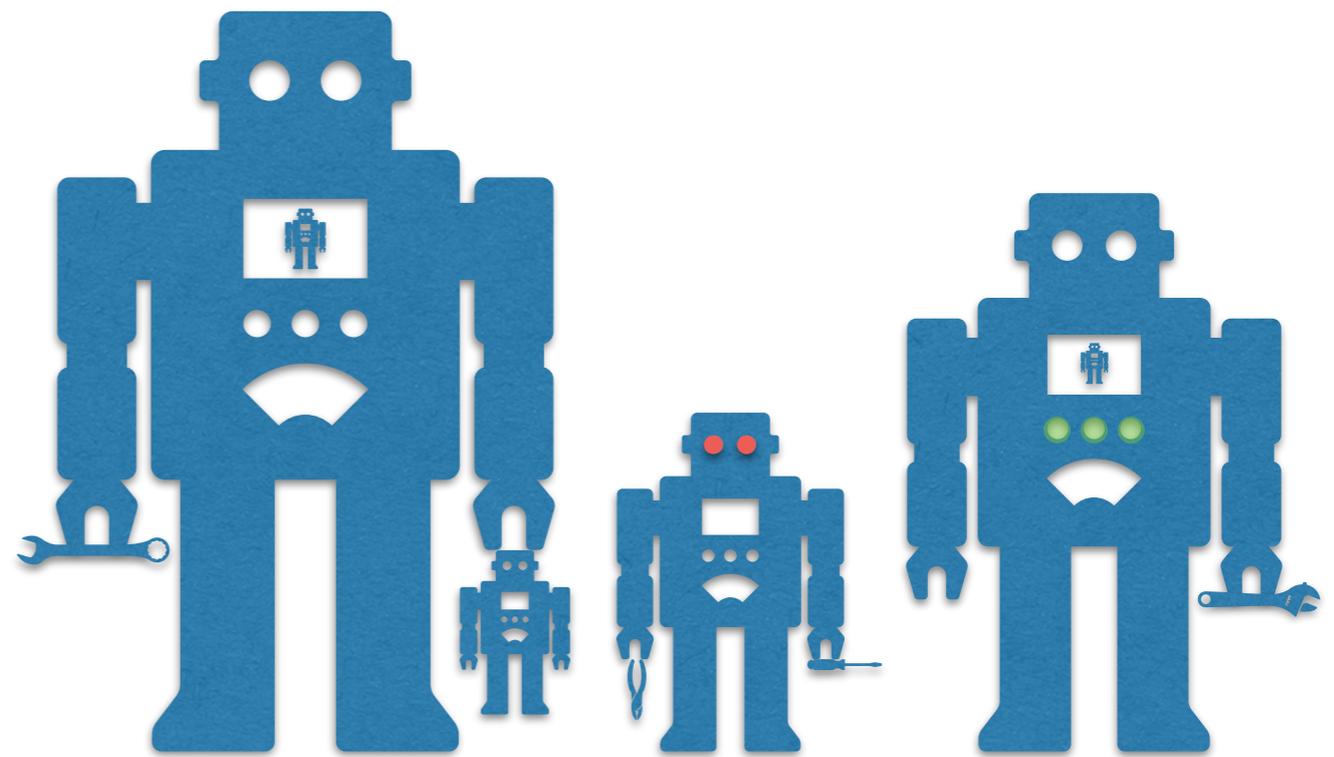
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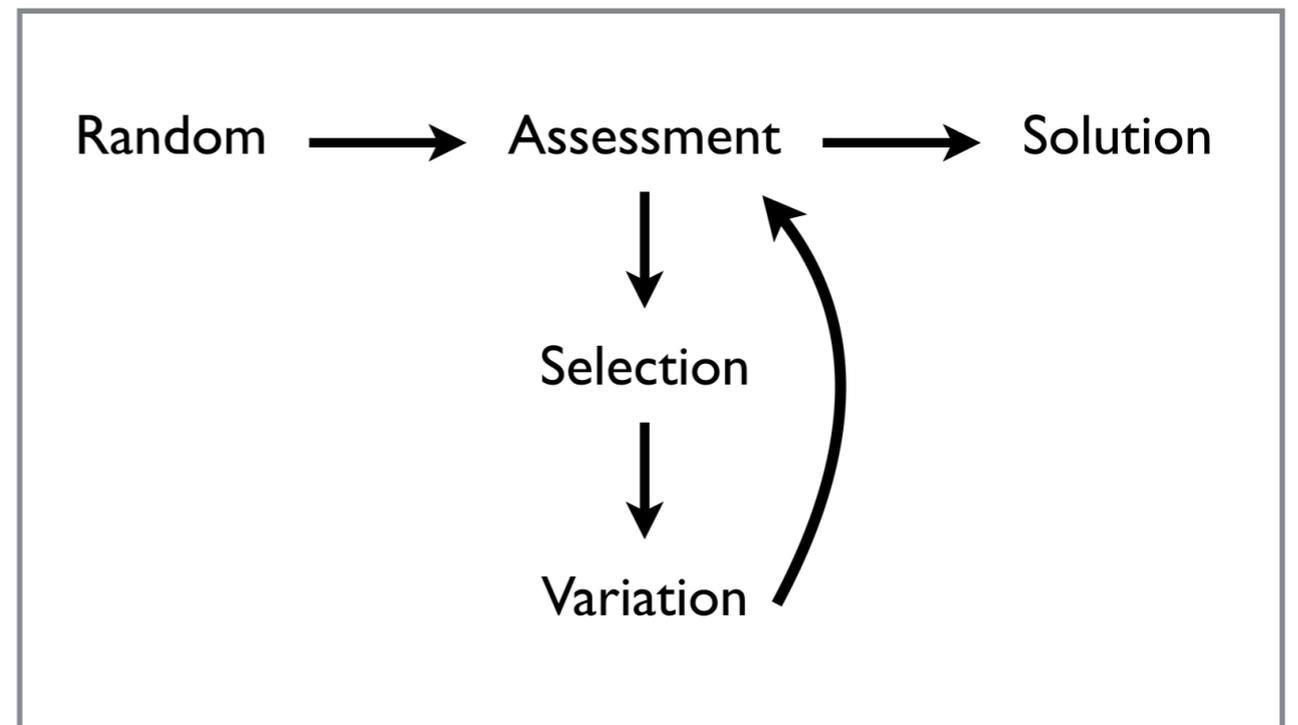
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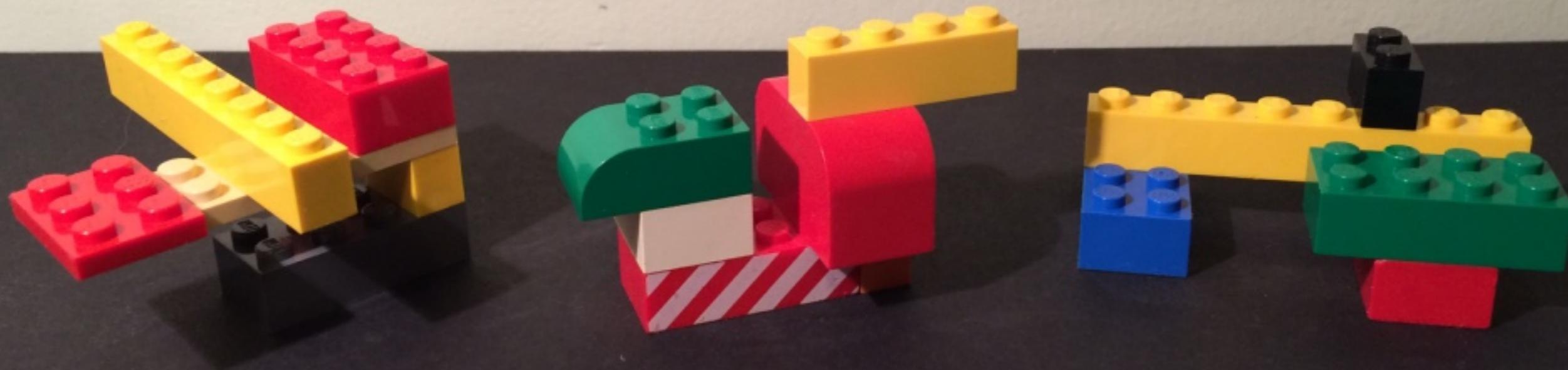
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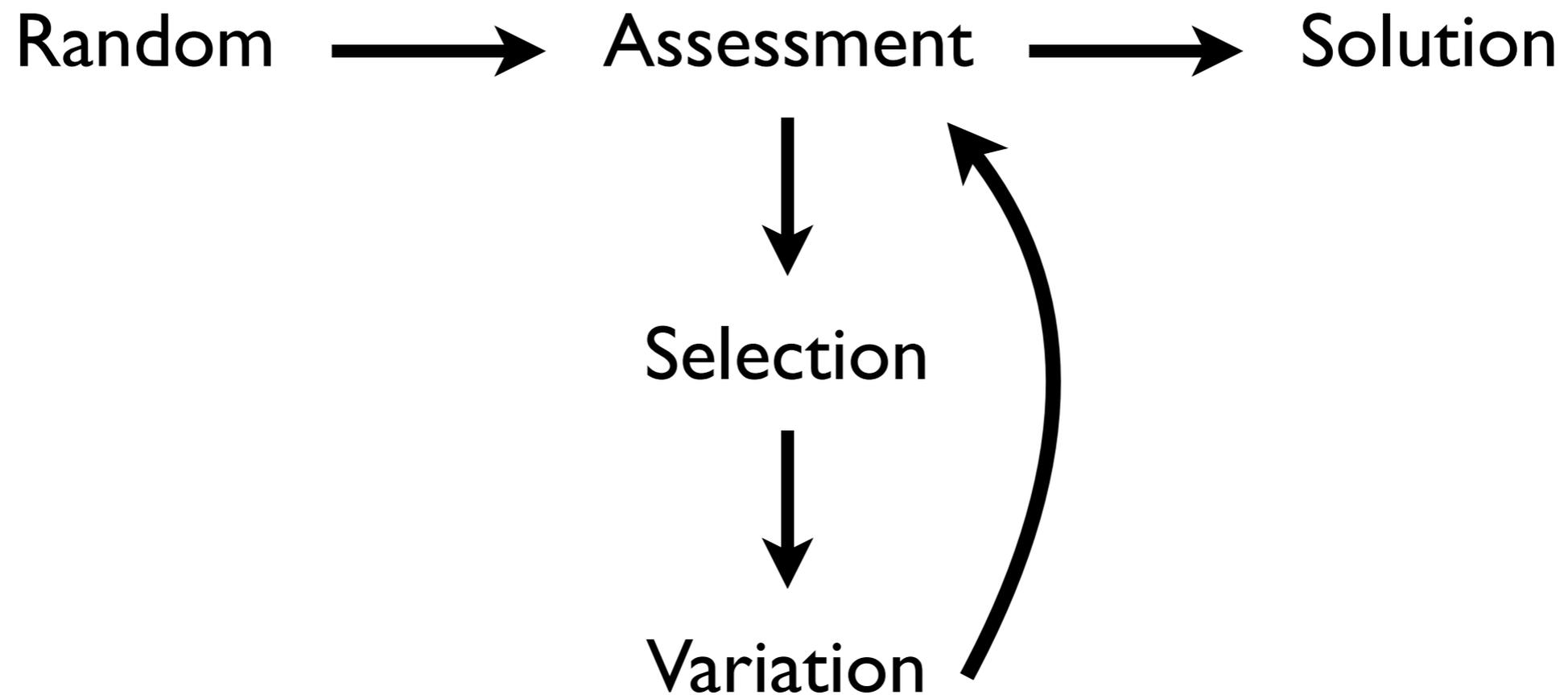


* **Evolutionary algorithms**

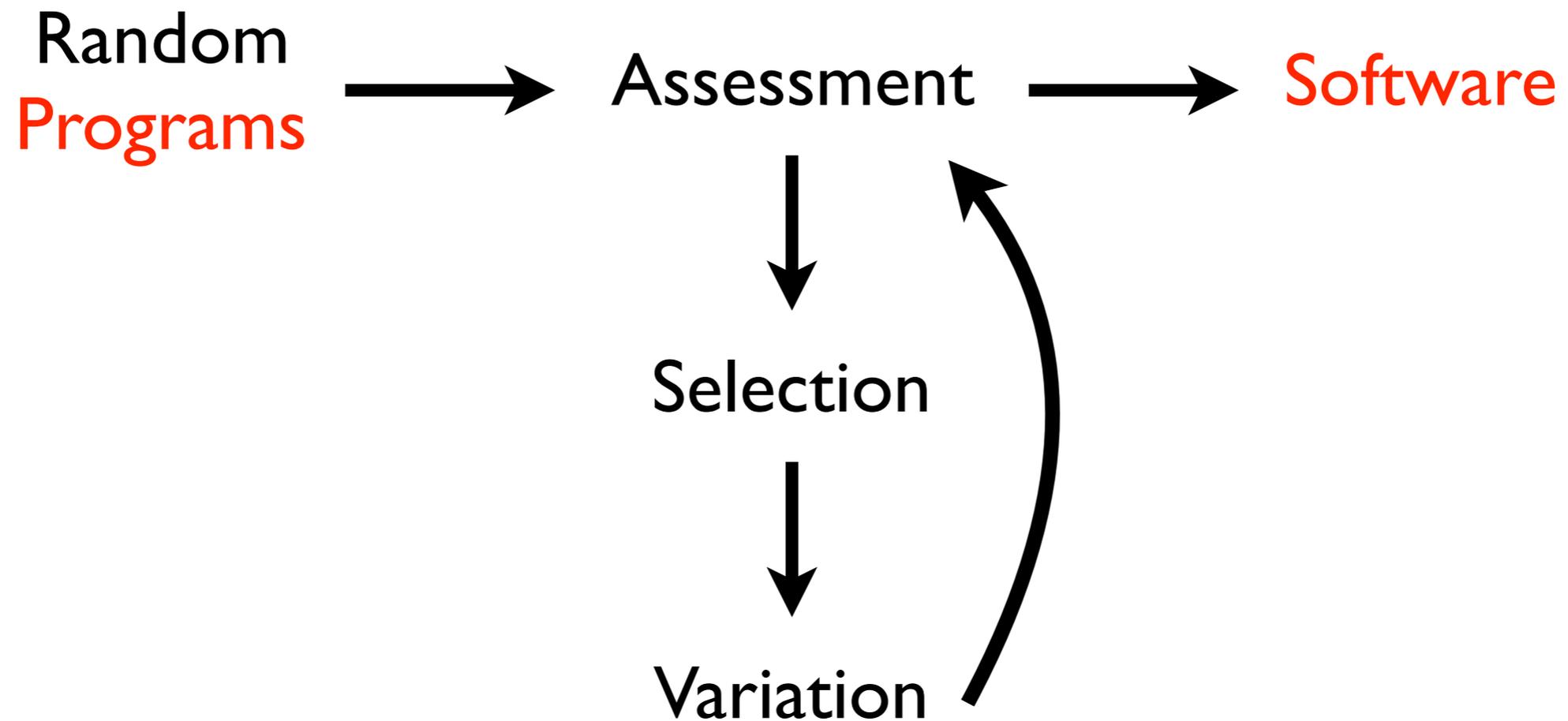
Evolving LEGO bridges



Evolutionary Algorithms



Genetic Programming



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Annual "Humies" Awards For Human-Competitive Results

Produced By Genetic And Evolutionary Computation

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***.

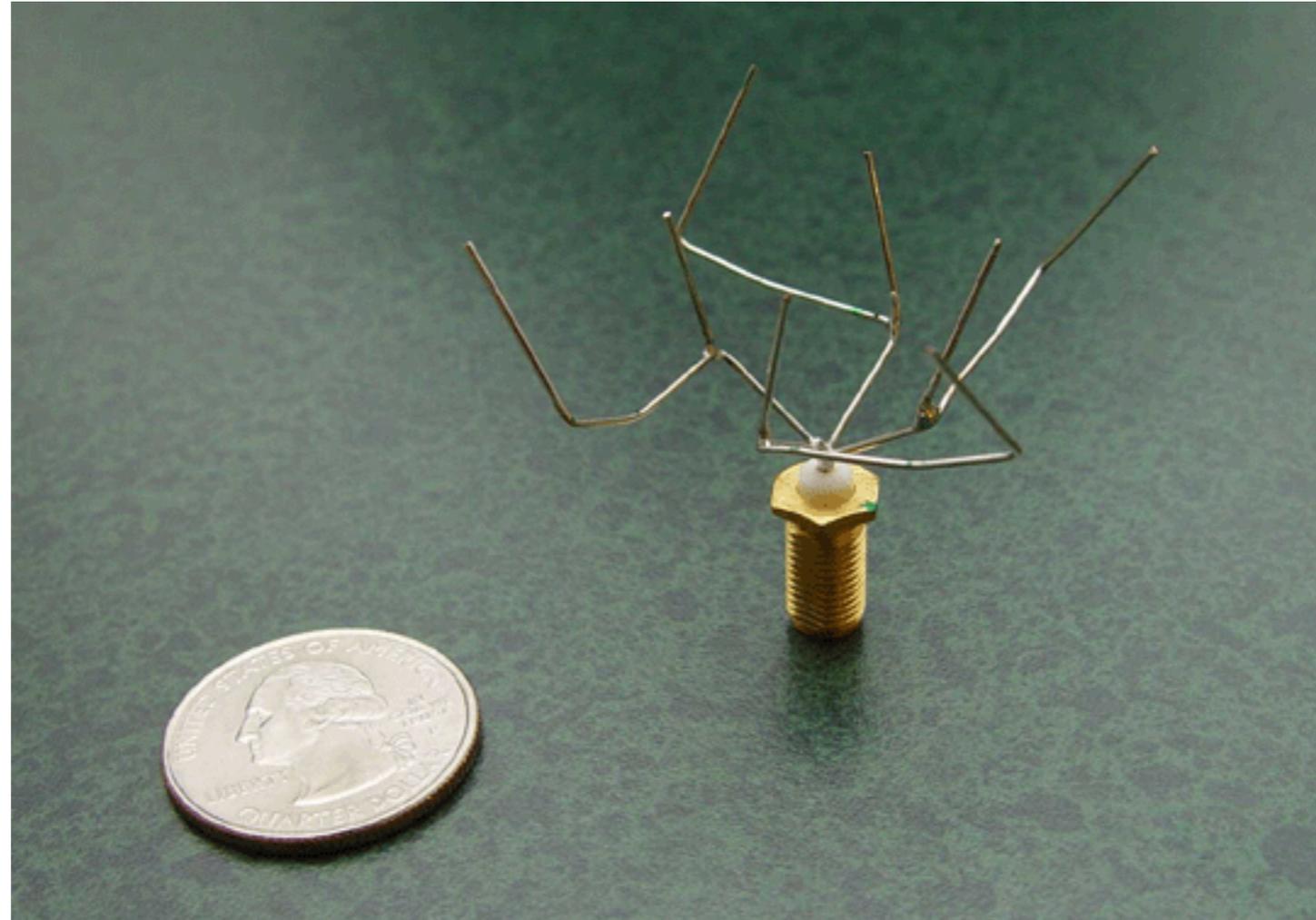
The result is ***publishable in its own right*** as a new scientific result independent of the fact that the result was mechanically created.

The result is ***equal to or better than the most recent human-created solution*** to a long-standing 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).



An Evolved Antenna for Deployment on NASA's Space Technology 5 Mission

Jason D. Lohn, Gregory S. Hornby, Derek S. Linden
NASA Ames Research Center

Humies Gold Medal, 2004

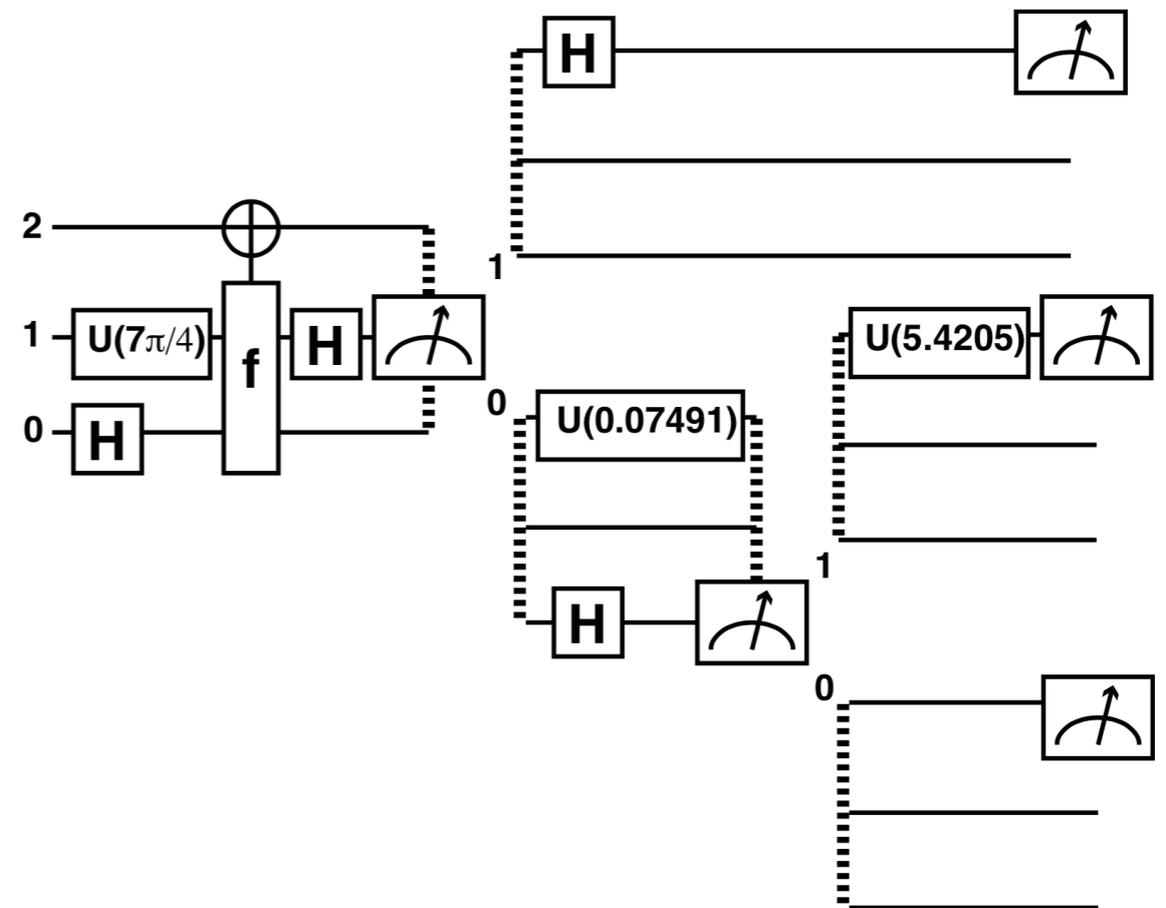
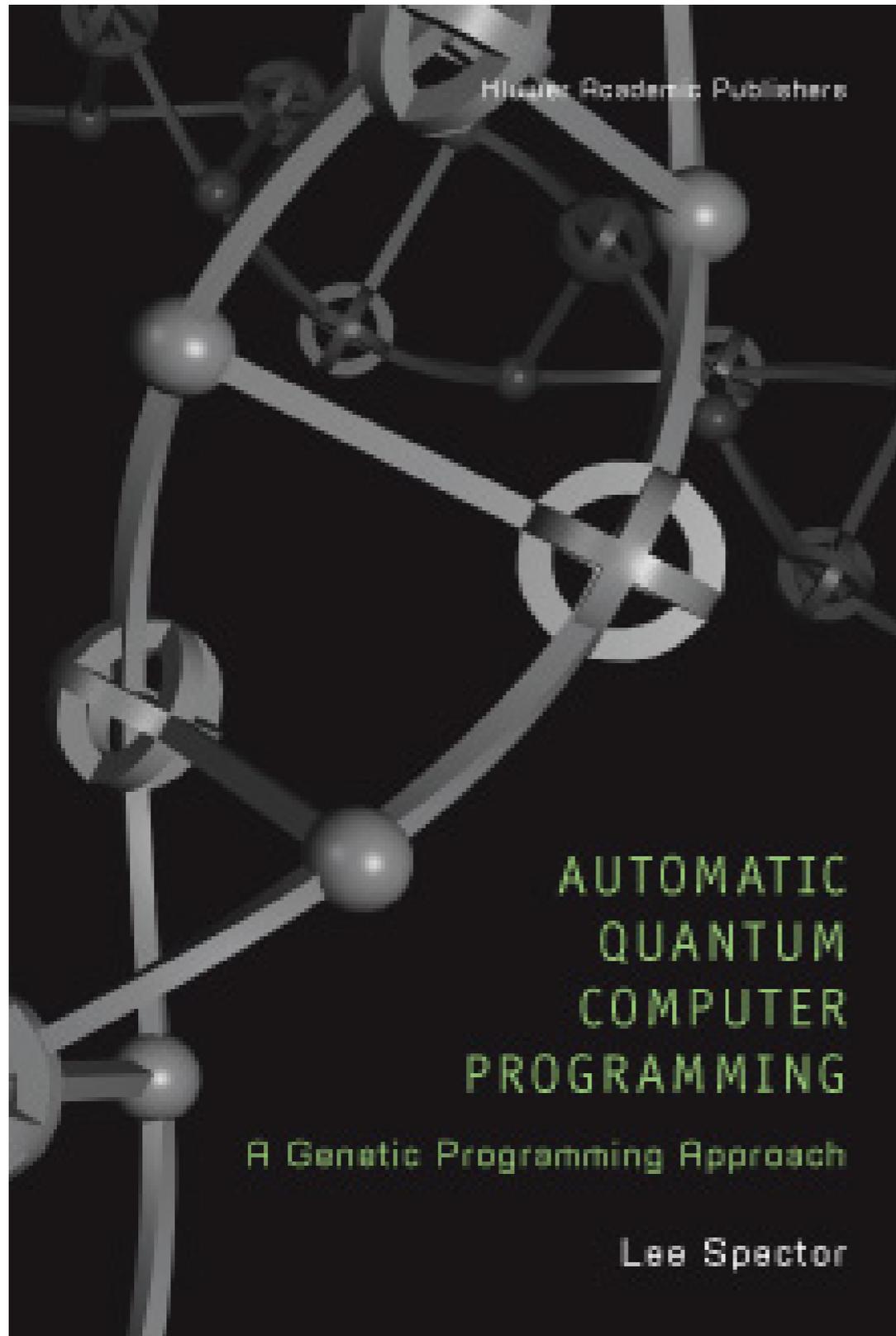


Figure 8.11. A gate array diagram for an evolved solution to the AND/OR oracle problem. The gate marked “f” is the oracle. The sub-diagrams on the right represent the possible execution paths following the intermediate measurements.

Lee Spector
Hampshire College

Humies Gold Medal, 2004

Genetic Programming for Finite Algebras

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Jon Klein
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$$\begin{aligned} &((((((((((X*(Y*X))*X)*Z)*(Z*X))*((X*(Z*(X*(Z*Y)))))*Z))* \\ &Z)*Z)*(Z*(((X*(((Z*Z)*X)*(Z*X))))*X)*Y)*(((Y*(Z*(Z* \\ &Y))))*((Y*Y)*X)*Z)*(X*(((Z*Z)*X)*(Z*(X*(Z*Y)))))) \end{aligned}$$

Humies Gold Medal, 2008

International Journal of Algebra and Computation | Vol. 28, No. 05, pp. 759-790 (2018)

Evolution of algebraic terms 3: Term continuity and beam algorithms

David M. Clark  and Lee Spector

Fixing software bugs in 10 minutes or less using evolutionary computation

University of New Mexico

Stephanie Forrest

ThanhVu Nguyen

University of Virginia

Claire Le Goues

Westley Weimer



THE UNIVERSITY *of*
NEW MEXICO

Humies Gold Medal, 2009

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 [nestorgames](http://nestorgames.com), making it the first — and still only — computer-generated game to be commercially published, together with its sister game [Pentalath](http://nestorgames.com).

In October 2011, Yavalath was ranked in the top #100 abstract board games ever invented on the [BoardGameGeek](http://boardgamegeek.com) database. This helped it win the GECCO "Humies" 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.

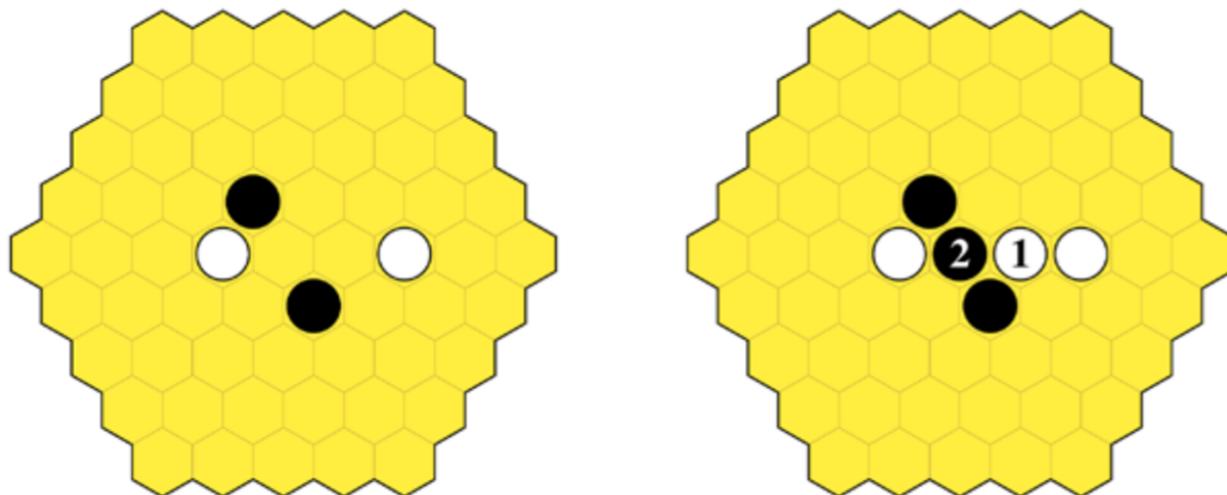
Draw if the board otherwise fills up.



No, players are not allowed to pass.

Tactics and Strategy

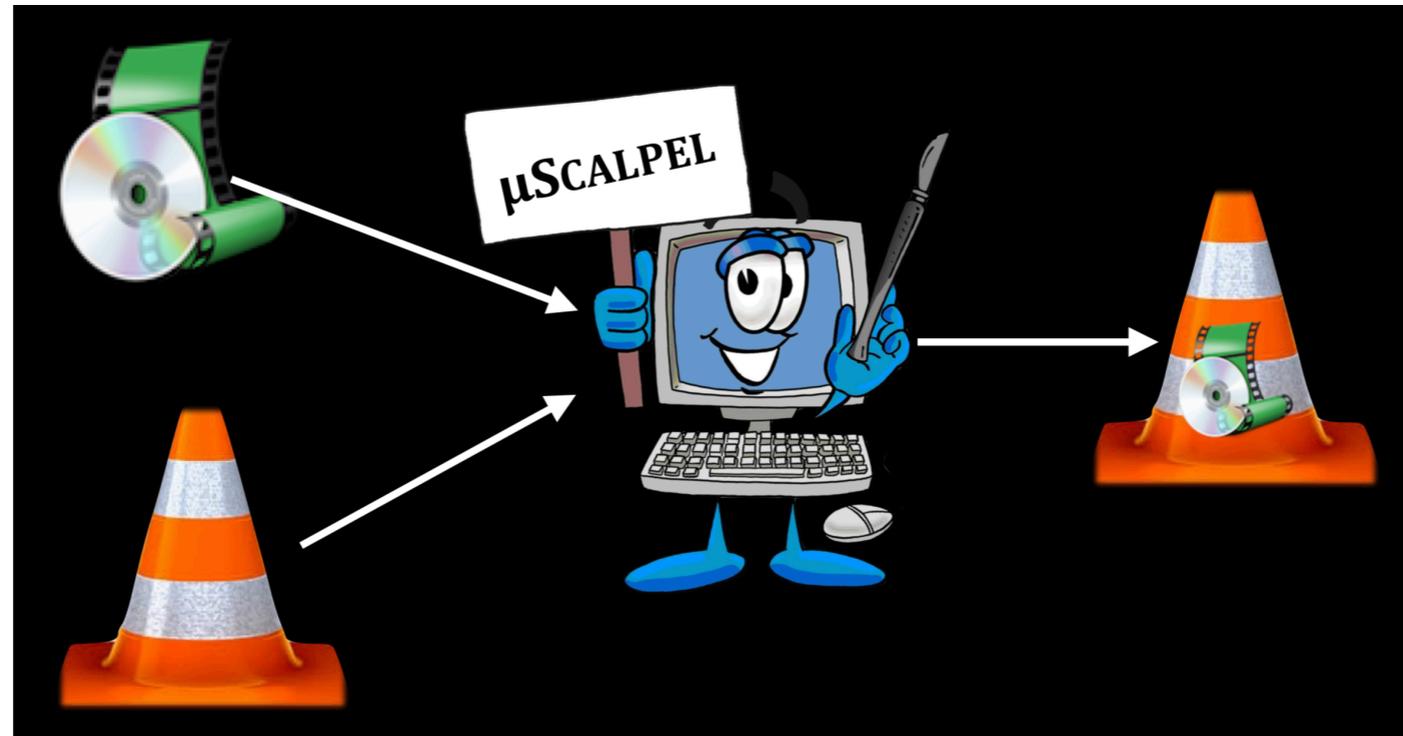
The key tactical play in Yavalath is the *forcing move*, as shown below. White move 1 forces Black to lose with the blocking move 2.



Cameron Browne
Imperial College London

Humies Gold Medal, 2012

Automated Software Transplantation



Earl T. Barr, Mark Harman, Yue Jia, Alexandru Marginean, Justyna Petke
University College London

Humies Gold Medal, 2016

Darwinian Data Structure Selection

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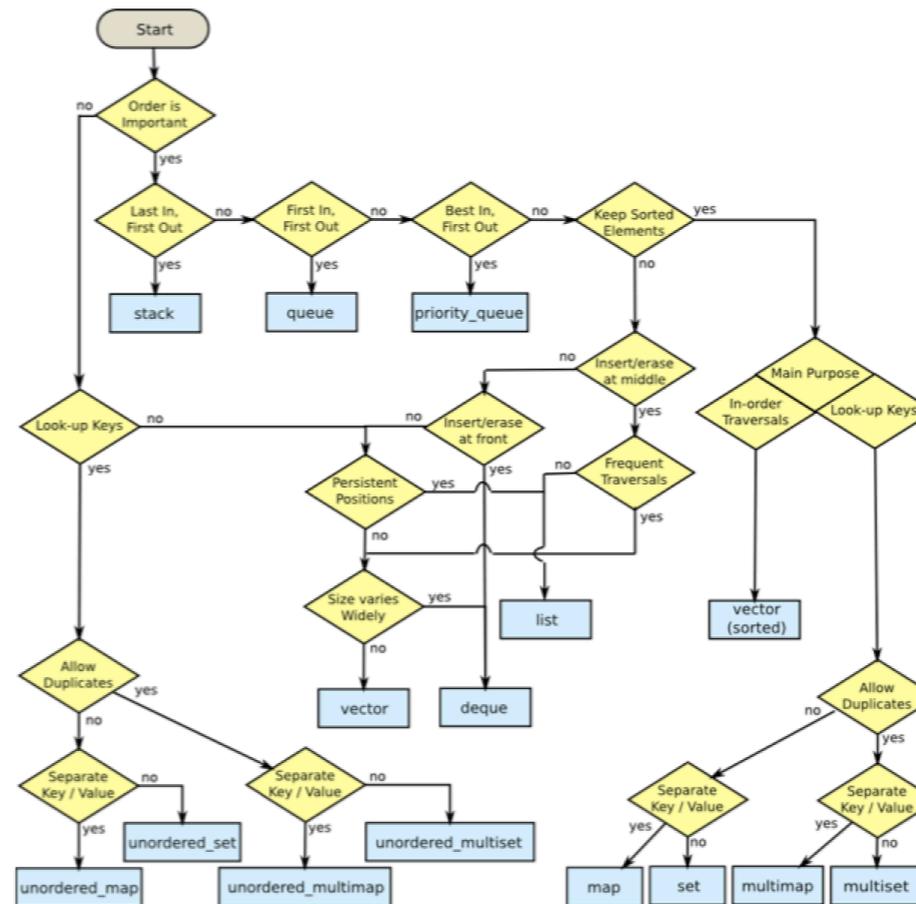
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Earl T. Barr
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Data Structure Selection/Optimisation Process



Hmmm, maybe just use defaults that work in most cases.



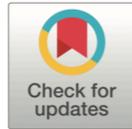
<https://www.human-competitive.org/sites/default/files/basiosslides.pptx>

Humies Bronze Medal, 2019

<https://www.human-competitive.org/awards>



ELSEVIER



Brachytherapy 18 (2019) 396–403

BRACHYTHERAPY

Physics

Evaluation of bi-objective treatment planning for high-dose-rate prostate brachytherapy—A retrospective observer study

Stefanus C. Maree^{1,*}, Ngoc Hoang Luong², Ernst S. Kooreman¹, Niek van Wieringen¹, Arjan Bel¹, Karel A. Hinnen¹, Henrike Westerveld¹, Bradley R. Pieters¹, Peter A.N. Bosman^{2,3}, Tanja Alderliesten¹

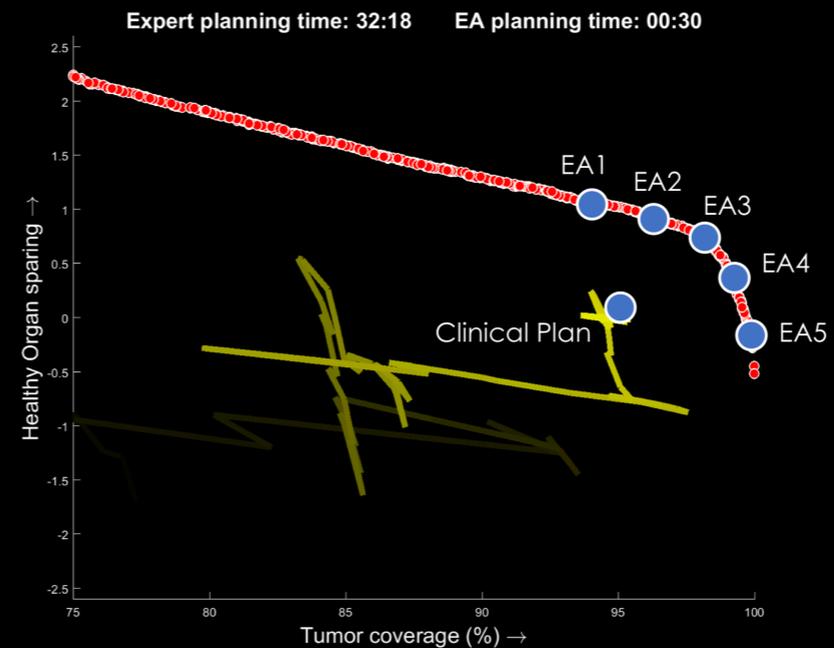
¹Department of Radiation Oncology, Amsterdam UMC, University of Amsterdam, Amsterdam, The Netherlands

²Life Sciences and Health Research Group, Centrum Wiskunde & Informatica, Amsterdam, The Netherlands

³Algorithmics group, Department of Software Technology, Faculty of Electrical Engineering, Mathematics and Computer Science, Delft University of Technology, Delft, The Netherlands

Human competition

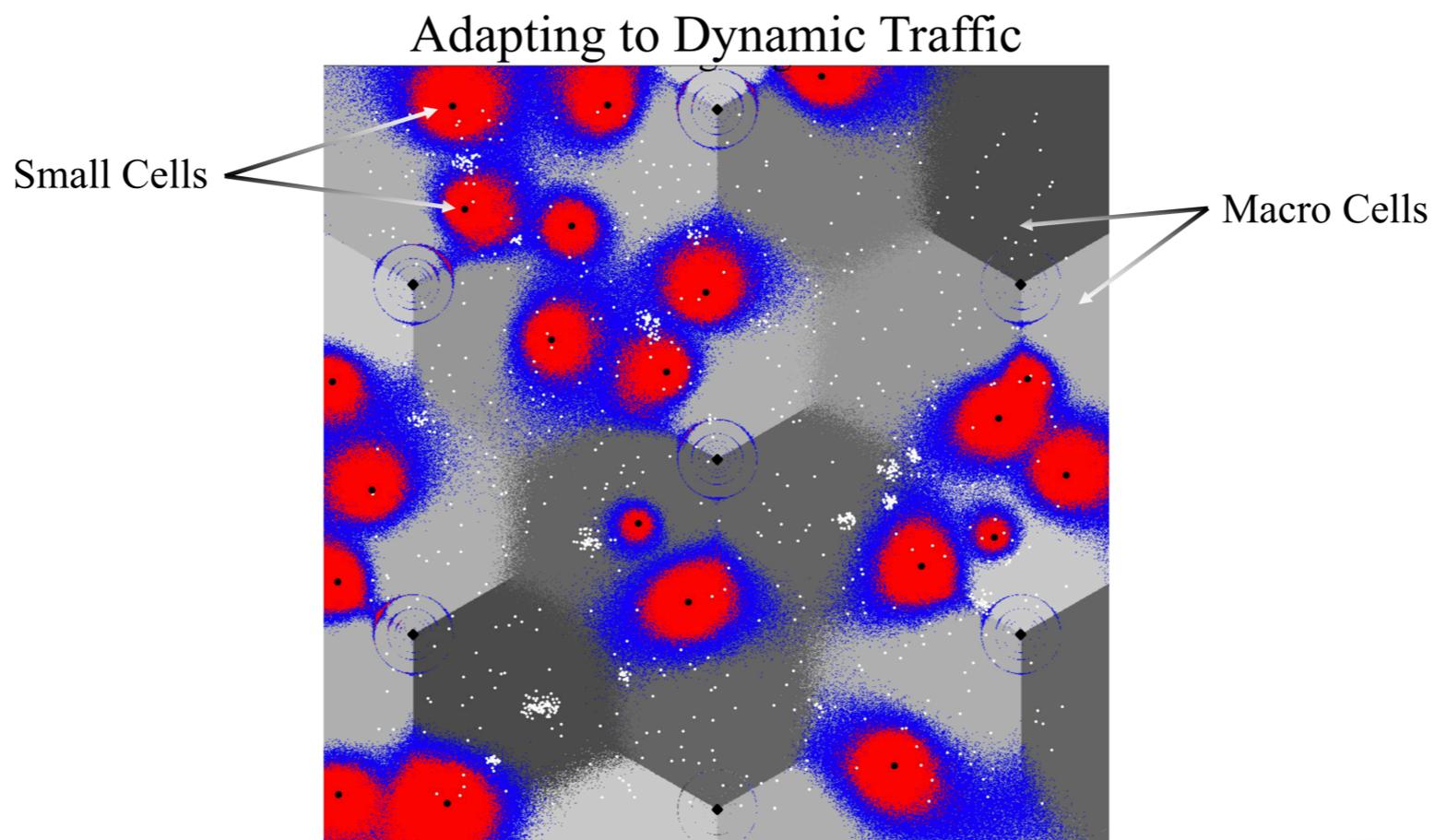
- 18 prostate cancer patients
- Blinded comparison between 6 plans
- 3 experienced radiation oncologists



Humies Silver Medal, 2019

Automated Self-Optimization in Heterogeneous Wireless Communications Networks

David Lynch^{ID}, Michael Fenton, David Fagan, Stepan Kucera, *Senior Member, IEEE*,
Holger Claussen, *Senior Member, IEEE*, and Michael O'Neill



Humies Gold Medal, 2019

Automatic identification of wind turbine models using evolutionary multiobjective optimization

William La Cava ^a  , Kouros Danai ^a, Lee Spector ^b, Paul Fleming ^c, Alan Wright ^c, Matthew Lackner ^a

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<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.

Multidimensional genetic programming for multiclass classification

William La Cava ^a  , Sara Silva ^{b, c, d}, Kouros Danai ^e, Lee Spector ^f, Leonardo Vanneschi ^c, Jason H. Moore ^a

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<https://doi.org/10.1016/j.swevo.2018.03.015>

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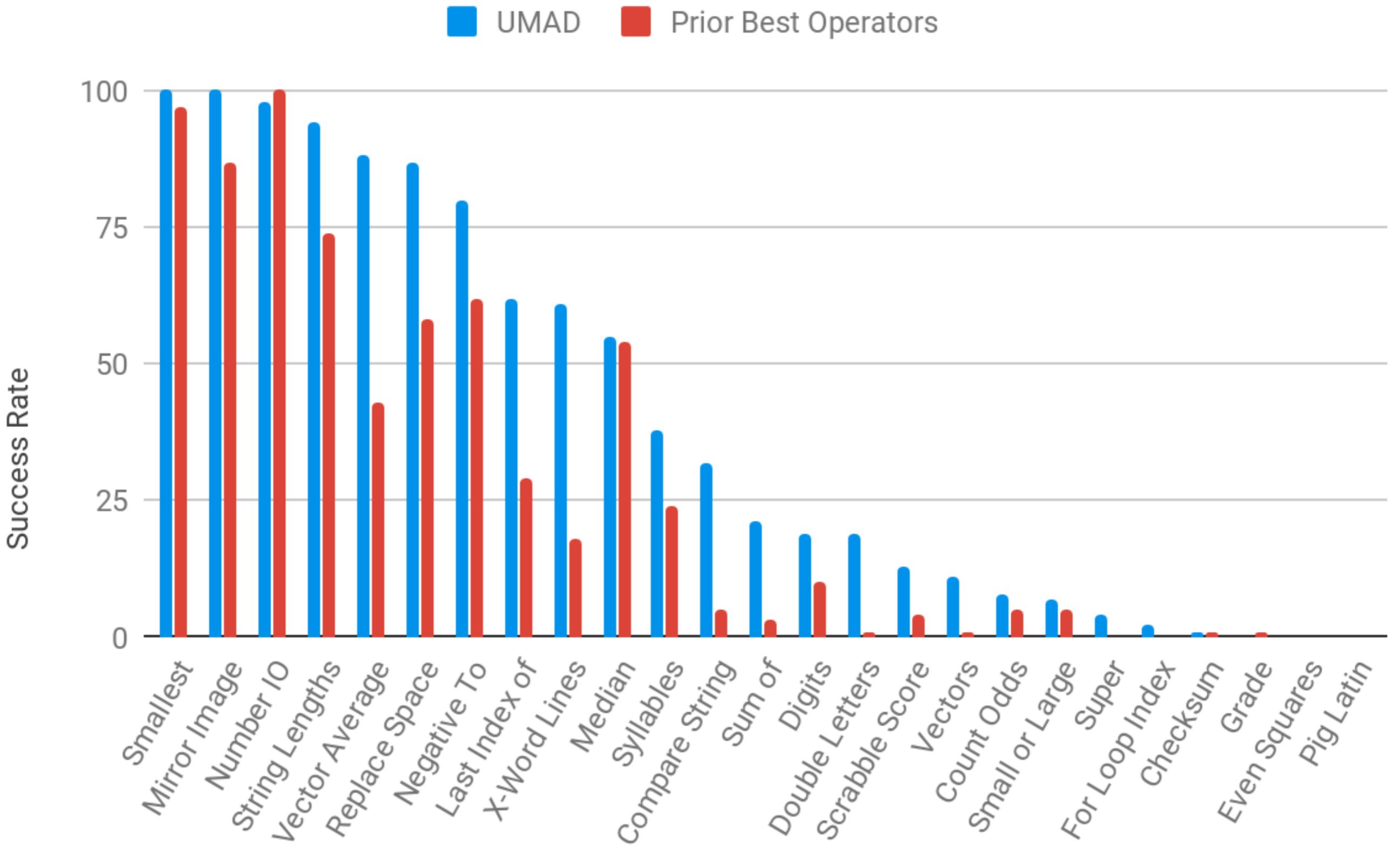
Abstract

We describe a new [multiclass classification](#) method that learns multidimensional feature transformations using [genetic programming](#). This method optimizes models by first performing a transformation of the feature space into a new space of potentially different [dimensionality](#), and then performing classification using a [distance function](#) in the transformed space. We analyze a novel program representation for using genetic programming to represent multidimensional features and compare it to other approaches. Similarly, we analyze the use of a [distance metric](#) for classification in comparison to simpler techniques more commonly used when applying genetic programming to multiclass classification. Finally, we compare this method to several state-of-the-art [classification techniques](#) across a broad set of problems and show that this technique achieves competitive test accuracies while also producing concise models. We also quantify the [scalability](#) of the method on problems of varying dimensionality, sample size, and difficulty. The results suggest the proposed method scales well to large feature spaces.

Software Synthesis

- 29 benchmark problems taken from intro CS textbooks
- Require multiple data types and control structures
- Driven by software tests, input/output pairs
- Used for studies of program synthesis, by us and by others

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.
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
```



| Application            | Count | Application Category |
|------------------------|-------|----------------------|
| Antennas               | 1     | Engineering (19)     |
| Biology                | 2     | Science (7)          |
| Chemistry              | 1     | Science (7)          |
| Computer vision        | 2     | Computer science (7) |
| Electrical engineering | 1     | Engineering (19)     |
| Electronics            | 5     | Engineering (19)     |
| Games                  | 6     | Games (6)            |
| Image processing       | 3     | Computer science (7) |
| Mathematics            | 2     | Mathematics (3)      |
| Mechanical engineering | 4     | Engineering (19)     |
| Medicine               | 2     | Medicine (2)         |
| Operations research    | 1     | Engineering (19)     |
| Optics                 | 2     | Engineering (19)     |
| Optimization           | 1     | Mathematics (3)      |
| Photonics              | 1     | Engineering (19)     |
| Physics                | 1     | Science (7)          |
| Planning               | 1     | Computer science (7) |
| Polymers               | 1     | Engineering (19)     |
| Quantum                | 3     | Science (7)          |
| Security               | 1     | Computer science (7) |
| Software engineering   | 3     | Engineering (19)     |

| Problem Type Count |    |
|--------------------|----|
| Classification     | 5  |
| Clustering         | 1  |
| Design             | 20 |
| Optimization       | 8  |
| Planning           | 1  |
| Programming        | 4  |
| Regression         | 3  |

Kannappan, K., L. Spector, M. Sipper, T. Helmuth, W. La Cava, J. Wisdom, and O. Bernstein. 2015. Analyzing a decade of Human-competitive ("HUMIE") winners -- what can we learn? In *Genetic Programming Theory and Practice XII*. New York: Springer.

# Evolution, the Designer

WHAT WOULD DARWIN SAY? | LEE SPECTOR

The Boston Globe

## And now, digital evolution

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.”

# What it is **Not** Good For

- Quite a lot
- It is solving "problems beyond the reach of other forms of AI"
- But also **not** solving problems **within** the reach of other forms of AI

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# Areas for Improvement

- Representation
- Variation
- Selection

# Areas for Improvement

- Representation
- Variation
- \* **Selection**

# Parent Selection

- Traditionally based on overall scores
- Roulette wheels or tournaments
- Unbalanced, qualitatively diverse test sets

STAPLES®

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|     |                |     |          |
|-----|----------------|-----|----------|
| %   | $\sqrt{\quad}$ | OFF | ON/C     |
| MRC | M-             | M+  | CE       |
| 7   | 8              | 9   | $\div$   |
| 4   | 5              | 6   | $\times$ |
| 1   | 2              | 3   | -        |
| 0   | .              | =   | +        |

# Biological Selection

- Survive challenges that you happen to face
- Until you can reproduce
- Each challenge may be competitive

# Lexicase Selection

- Don't use overall scores
- To select single parent:
  1. Shuffle test cases
  2. First test case – keep best\* individuals
  3. Repeat with next test case, etc.Until one individual remains
- Selected parent may be specialist, not great on average, but lead to generalists later

---

| Problem name               | Lexicase | Tournament |
|----------------------------|----------|------------|
| Replace Space With Newline | 57       | 13         |
| Syllables                  | 24       | 1          |
| String Lengths Backwards   | 75       | 18         |
| Negative To Zero           | 72       | 15         |
| Double Letters             | 5        | 0          |
| Scrabble Score             | 0        | 0          |
| Checksum                   | 0        | 0          |
| Count Odds                 | 4        | 0          |

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# Diversity

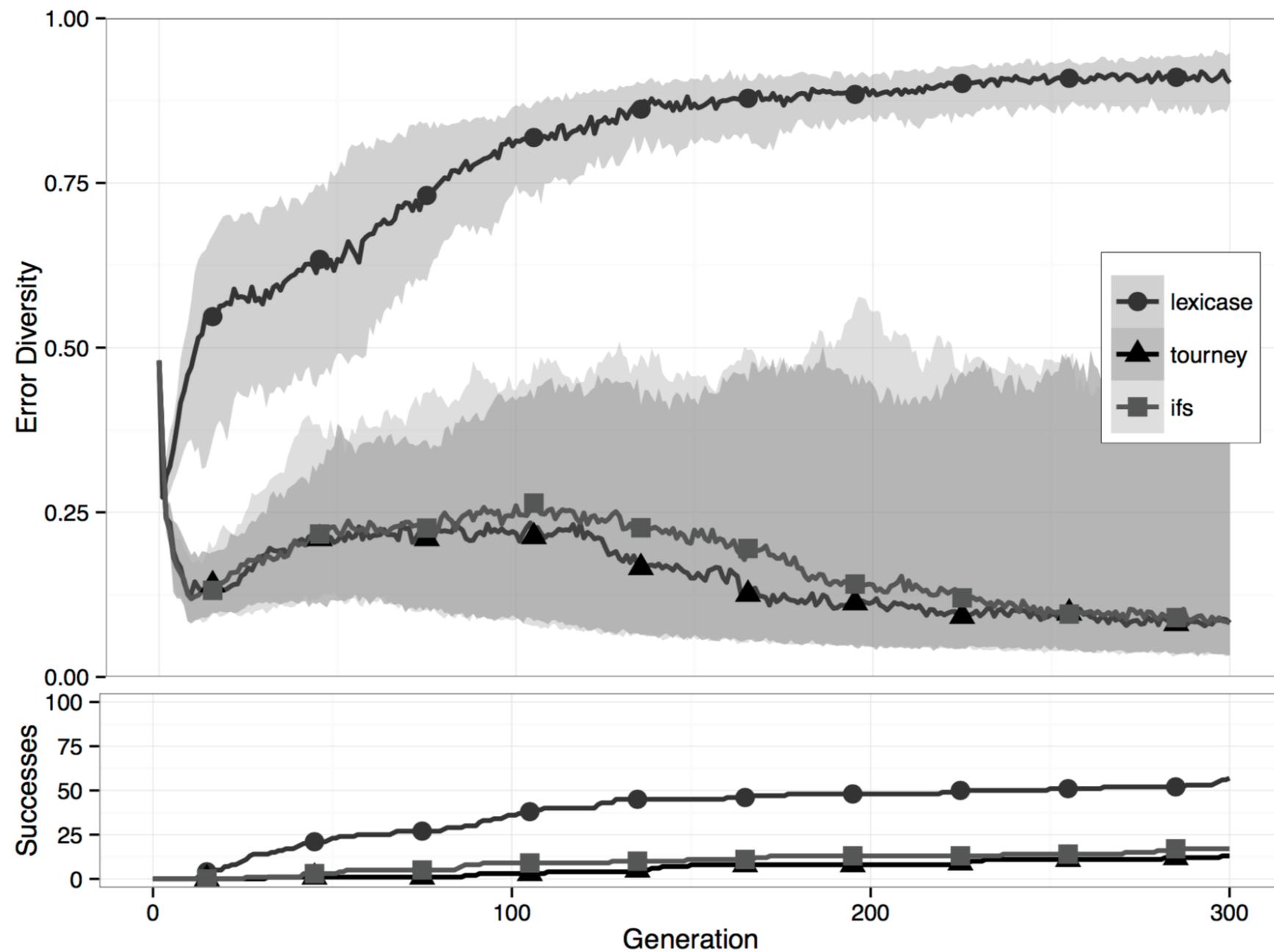


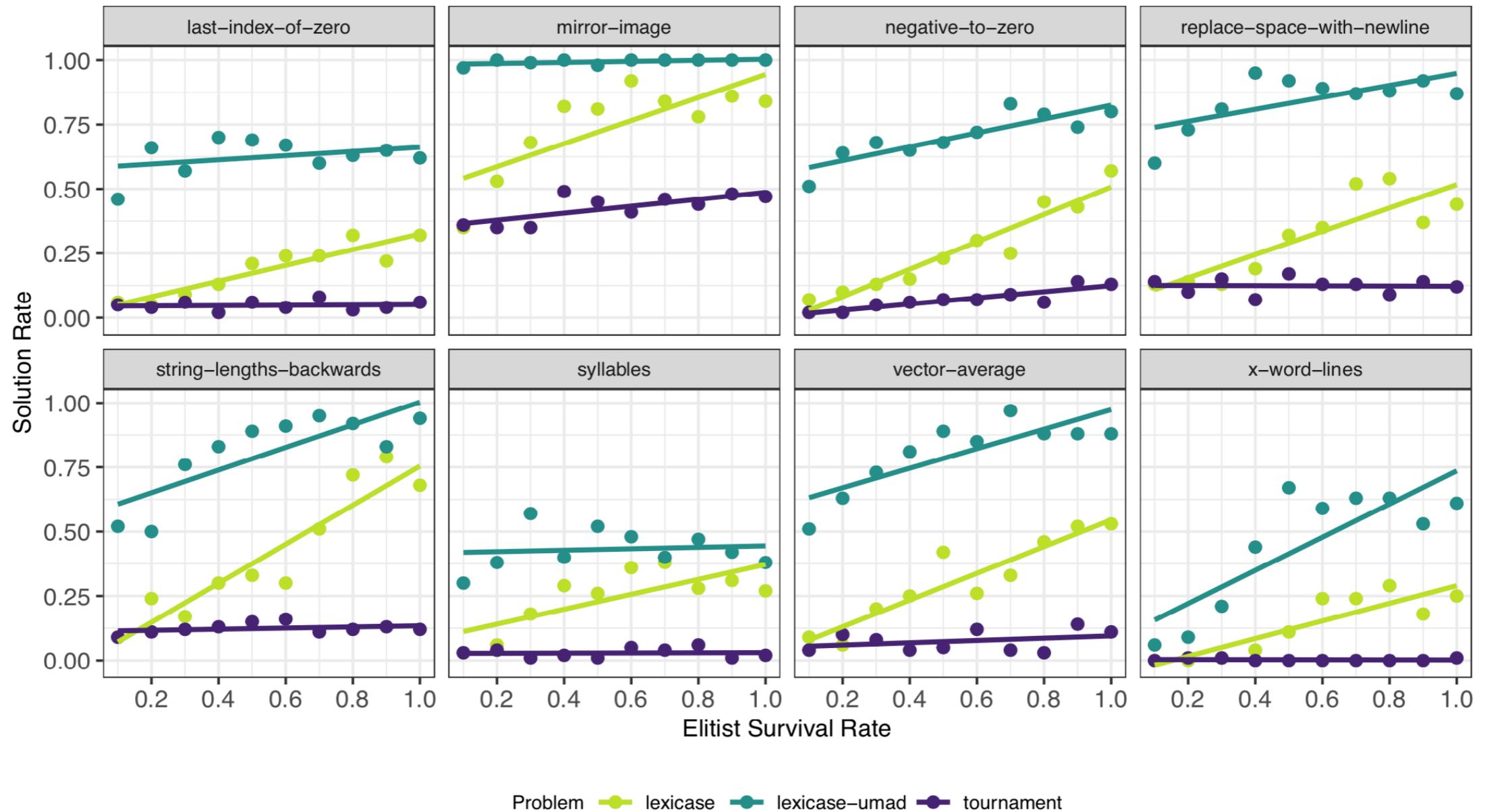
Fig. 1 Replace Space With Newline – error diversity

GPTP-2015

# Why Does it Work?

- Prior results: Diversity alone is not the answer
- Hypothesis: Selecting specialists is important
- Test by only allowing programs with good overall scores (good non-specialists) to be selected
- Degraded performance would suggest that specialists are important

# Specialists Help



# General Lessons?

In what ways might specialists be important in:

- Other forms of machine learning?
- Biological evolution?
- Engineering teams?
- Educational communities?

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# Connections

- Machine learning
- Software engineering
- Programming languages
- Theory
- Evolutionary biology
- Applications

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# Takeaways

- Many EvoBio/CompSci intersections
- Evolutionary algorithms use variation and selection to solve hard, interesting, and important problems
- Ample opportunities for improvement and application
- Specialists appear to be important

Thank you!

Questions?