

The Evolution of Identity and Modularity in Nature and Computation

Lee Spector

Cognitive Science

Hampshire College

Oberlin class of 1984

(Philosophy, TIMARA, WOBC, Tank Coop, ...)

Overview

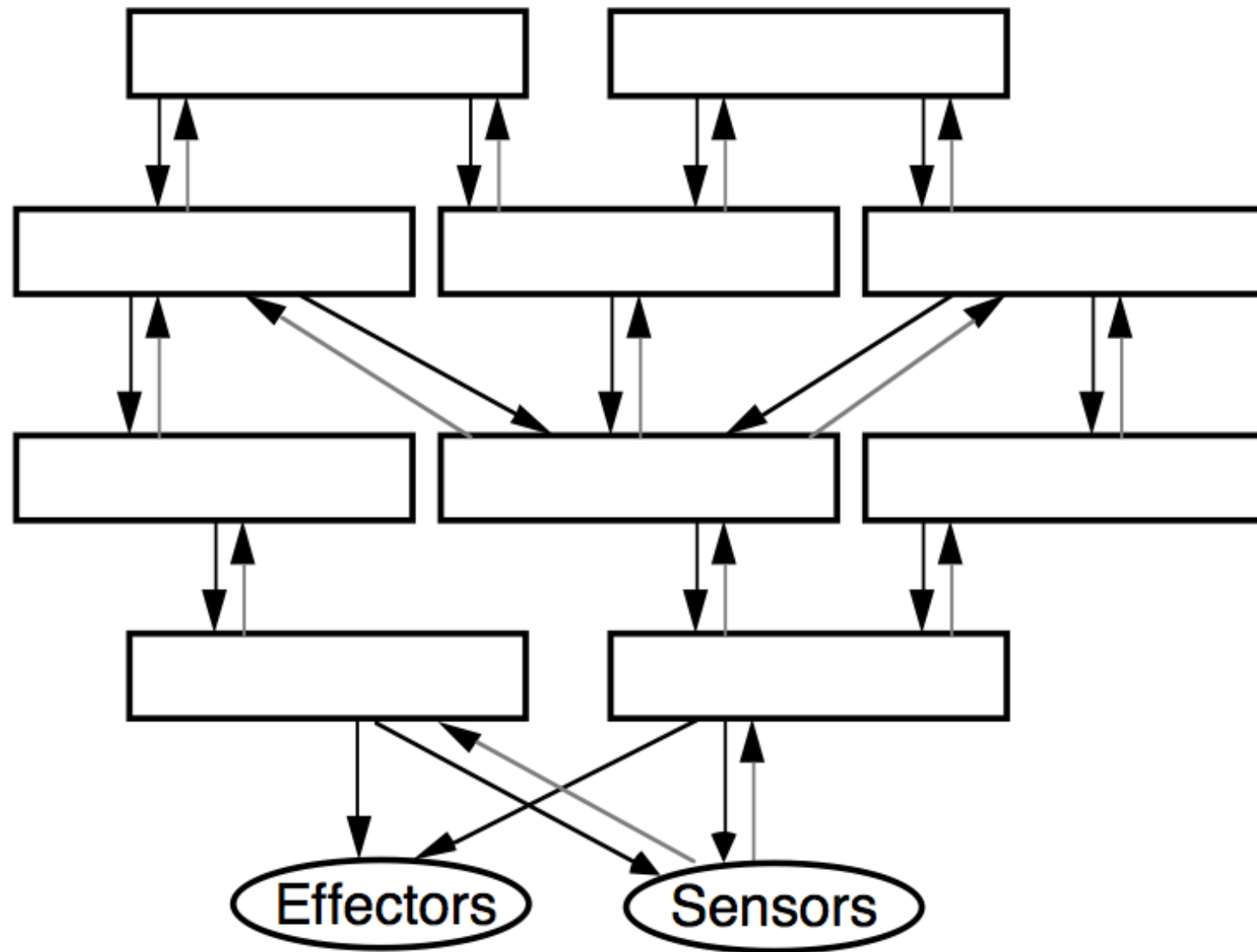
- Modularity
- Identity
- Evolving computer programs
- Evolving modular programs
- Implications

Hampshire College



- Undergraduate only
- Experimental/experimenting
- Five Colleges consortium
- No grades, credits, majors, departments, ...
- School of Cognitive Science

Modularity is Everywhere





<http://equitygreen.typepad.com/blog/2007/08/hybrid-seattle-.html#more>



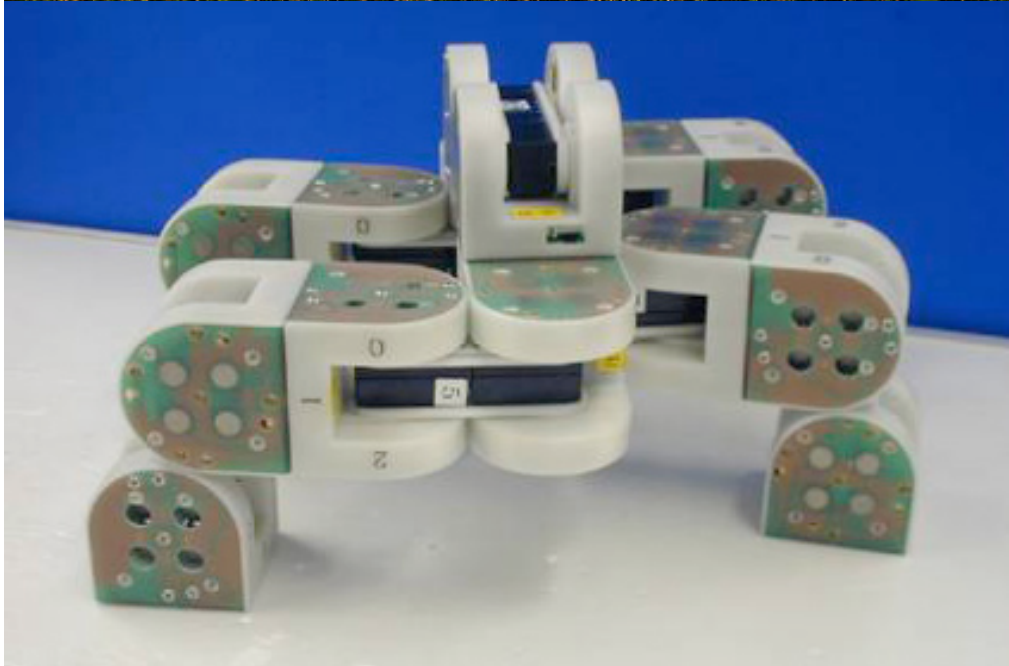
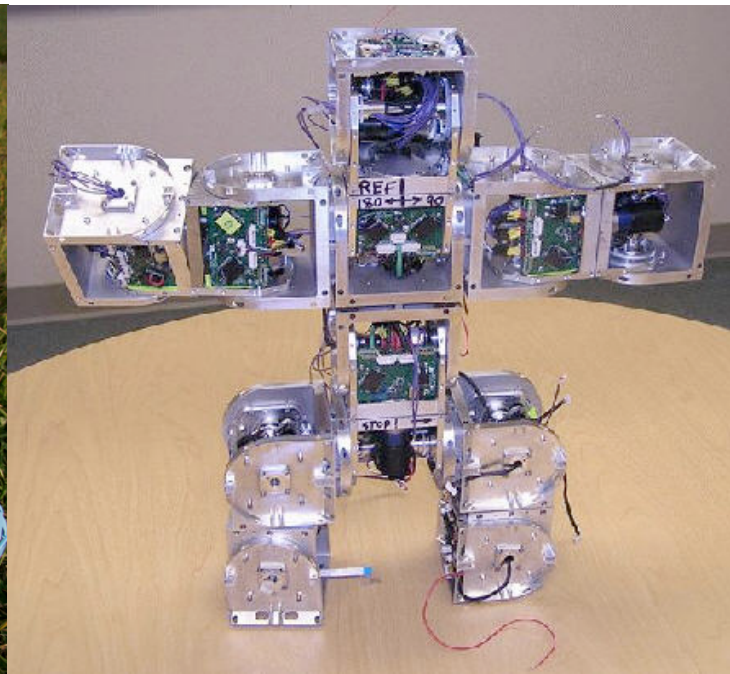
<http://www.flickrfotos.com/modular-44-plastic-coffee-table-design/>



<http://talkinterior.com/interior-design-vita-minimalist-modular-home/>



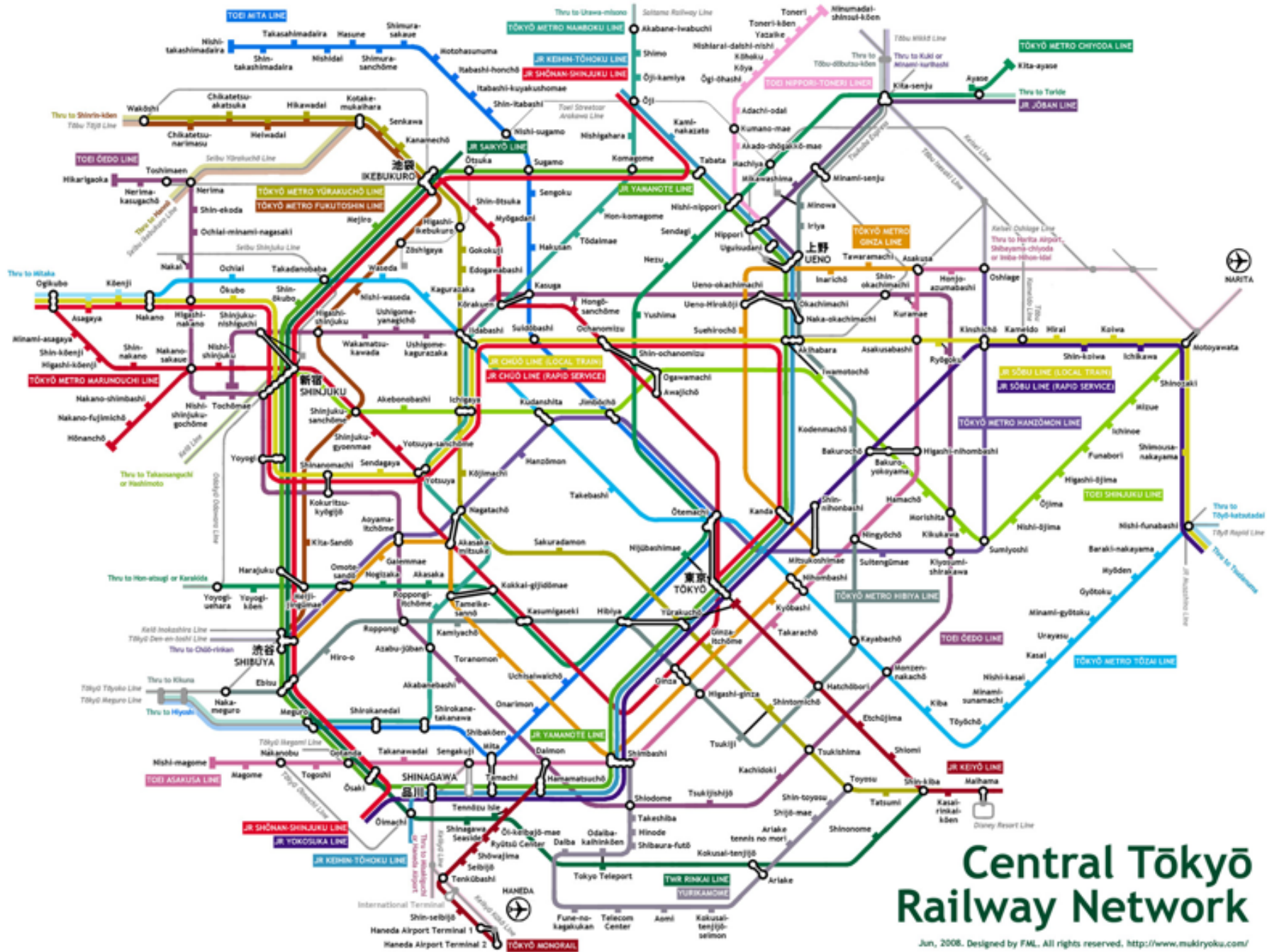
<http://www.e-potpourri.com/index.php/2008/02/02/octopus-studios-silverfish-aquarium-boasts-modern-modular-design/>



<http://wyss.harvard.edu/viewevent/37/wyss-seminar-series-kasper-stoy>
<http://www.technovelgy.com/ct/Science-Fiction-News.asp?NewsNum=953>
<http://www.engadget.com/2005/03/26/m-tran-self-reconfigurable-modular-robot/>
<http://www.hizook.com/blog/2012/01/16/ted-talks-about-robots-and-robotics-part-1>



<http://www.synthtopia.com/content/2007/04/04/moog-55-modular-synthesizer/>



<http://mappery.com/map-of/Tokyo-Metro-Map>

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.



<http://en.wikipedia.org/wiki/File:Sa-fern.jpg>



<http://a-z-animals.com/animals/centipede/>

Cognitive Science

- Long history of modularity theories: Gall, ... Simon, ... Fodor, ... Cermak and Craik, ... Gardner, ... Jackendoff, ... Grafman, ...
- Simon's "nearly decomposable systems"
- Fodor's features: domain specific, mandatory, fast, encapsulated, fixed architecture, characteristic patterns of ontogeny and failure
- Central vs. input systems
- Modest vs. massive

mod•ule | 'mäjool |

noun

each of a set of standardized parts or independent units that can be used to construct a more complex structure, such as an item of furniture or a building.

- [usu. with adj.] an independent self-contained unit of a spacecraft.
- Computing any of a number of distinct but interrelated units from which a program may be built up or into which a complex activity may be analyzed.

ORIGIN late 16th cent. (in the senses '*allotted scale*' and '*plan, model*'): from French, or from Latin *modulus* (see MODULUS). Current senses date from the 1950s.

Questions

- **Why** are modules everywhere?
- What are they good for?
- Where do they come from?
- What conditions permit or facilitate their emergence?

Identity

- How are modules recognized by other components of a system?
- Where do module identities come from?
- How can module identity co-evolve with modular architecture?

Holland's Tags

- Initially arbitrary identifiers that come to have meaning over time
- 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

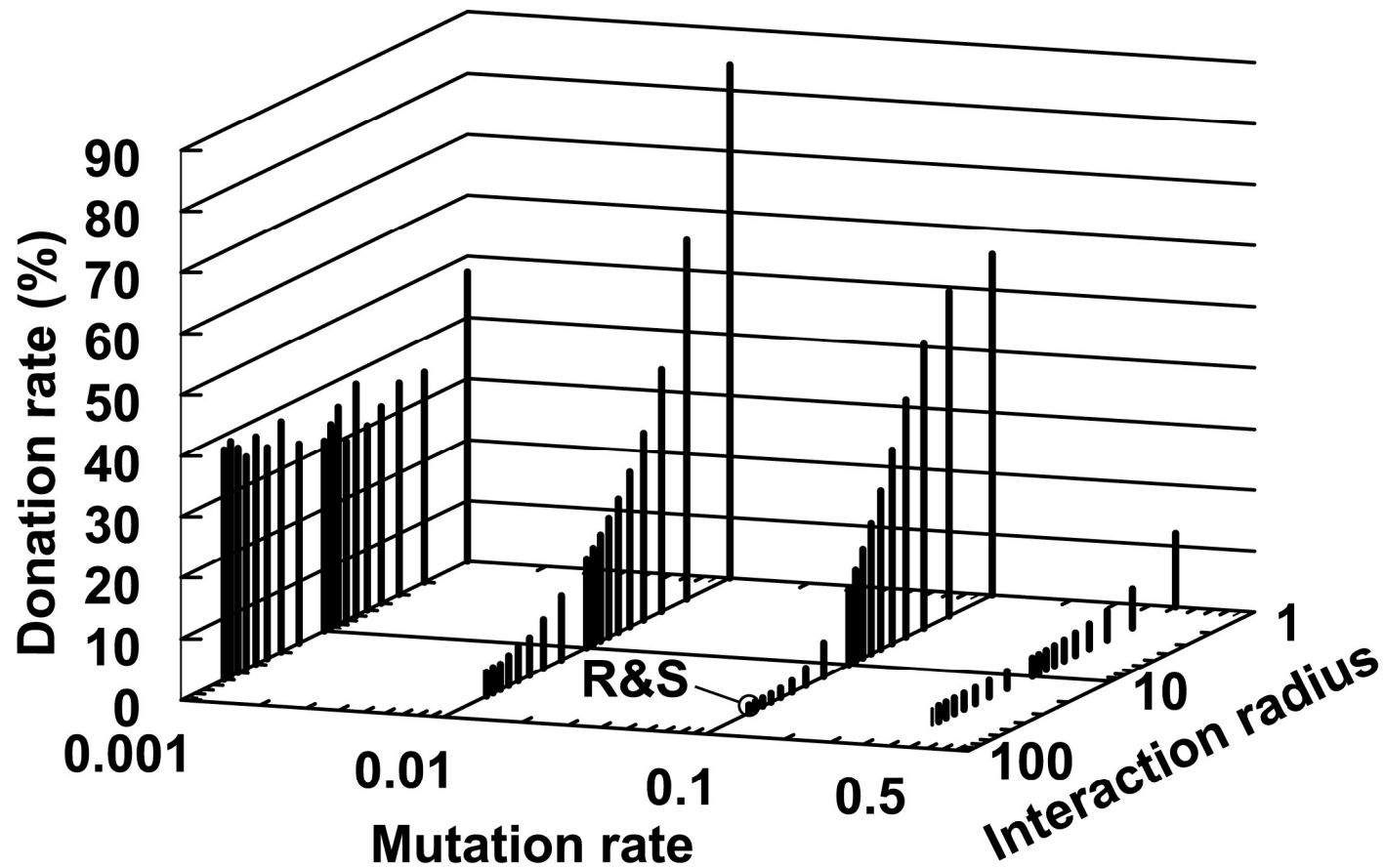
Evolution of Altruism

- Puzzles/challenges/results since Darwin
- Explanations of altruism toward:
 - Kin
 - Reciprocating partners
 - Agents with good reputations



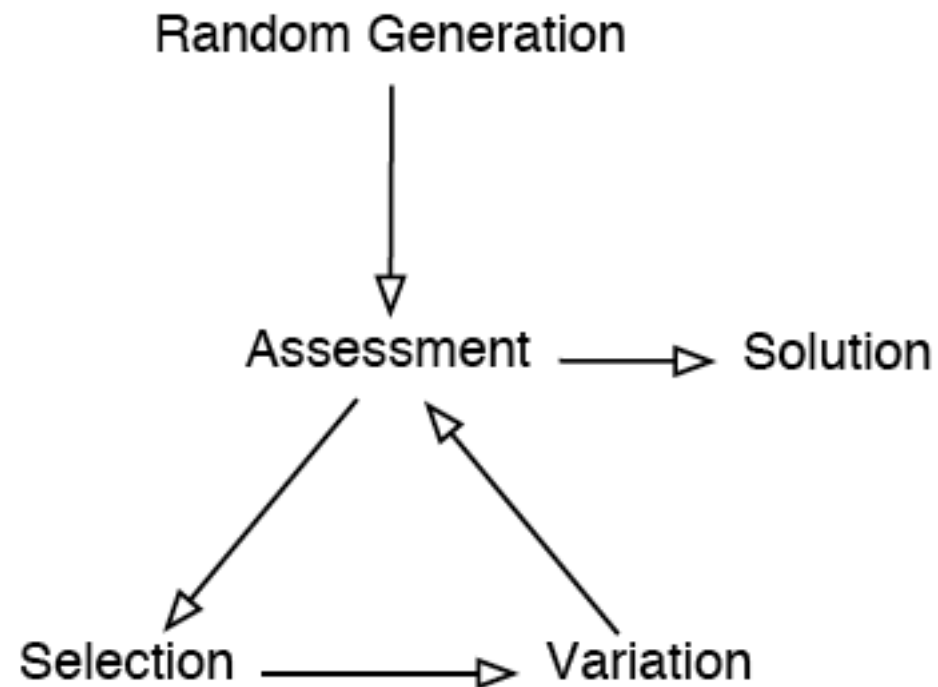
Tag-Based Altruism

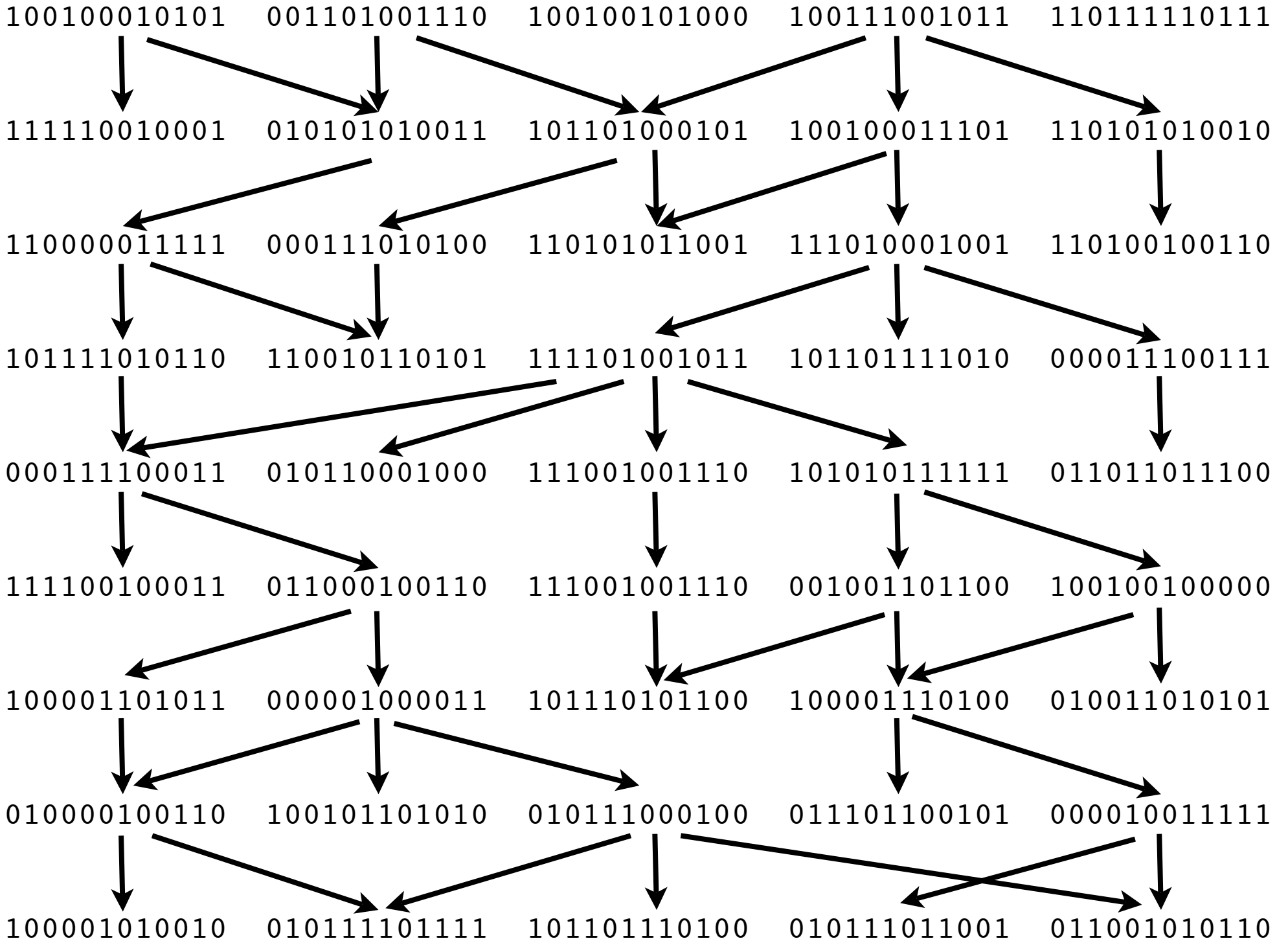
- Individuals have tags and tag-difference tolerances
- Donate when $\Delta\text{tags} \leq \text{tolerance}$
- Riolo *et al.* (*Nature*, 2001) showed that tag-based altruism can evolve; Roberts & Sherratt (*Nature*, 2002) claimed it would not evolve under more realistic conditions



Spector, L., and Klein, J. Genetic stability and territorial structure facilitate the evolution of tag-mediated altruism. In *Artificial Life*.

Evolutionary Computation





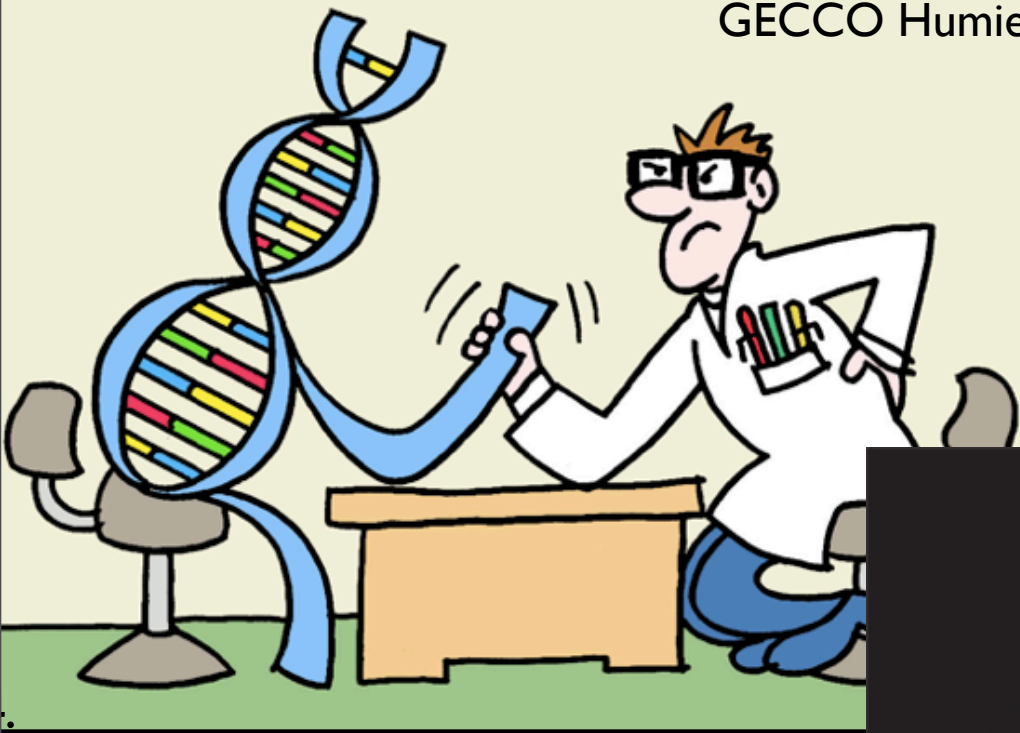
Traditional Genetic Algorithms

- Interesting dynamics
- Rarely solve interesting hard problems

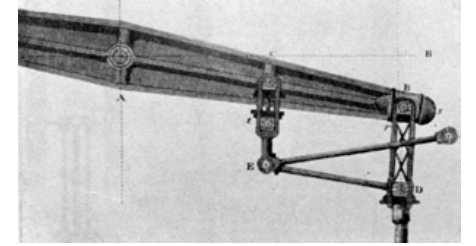
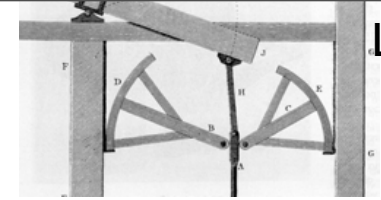
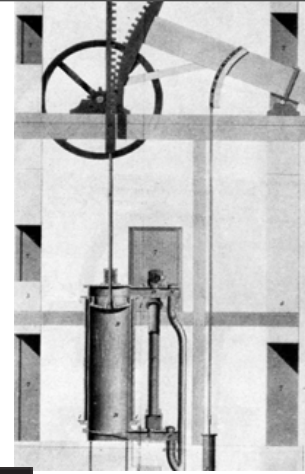
Genetic Programming

- Evolutionary computing to produce executable computer programs.
- Programs are tested by executing them.

GECCO Humies

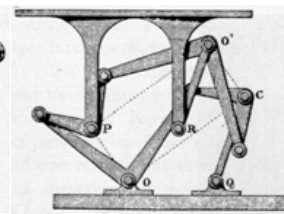
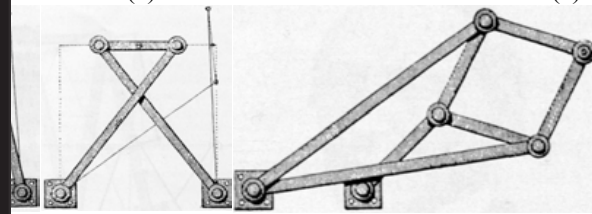


Lipson



(a)

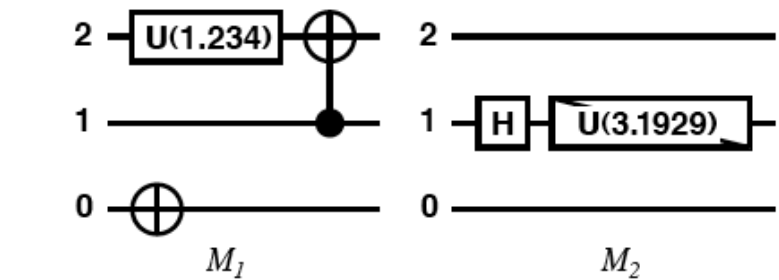
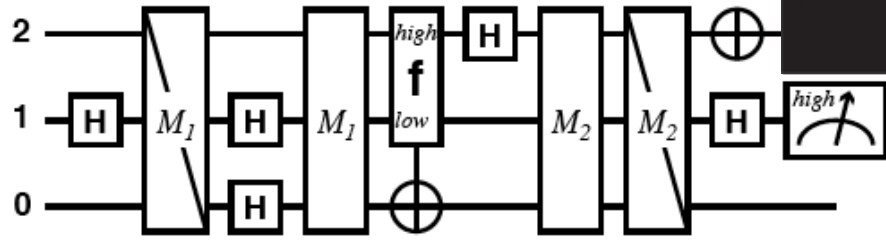
(c)



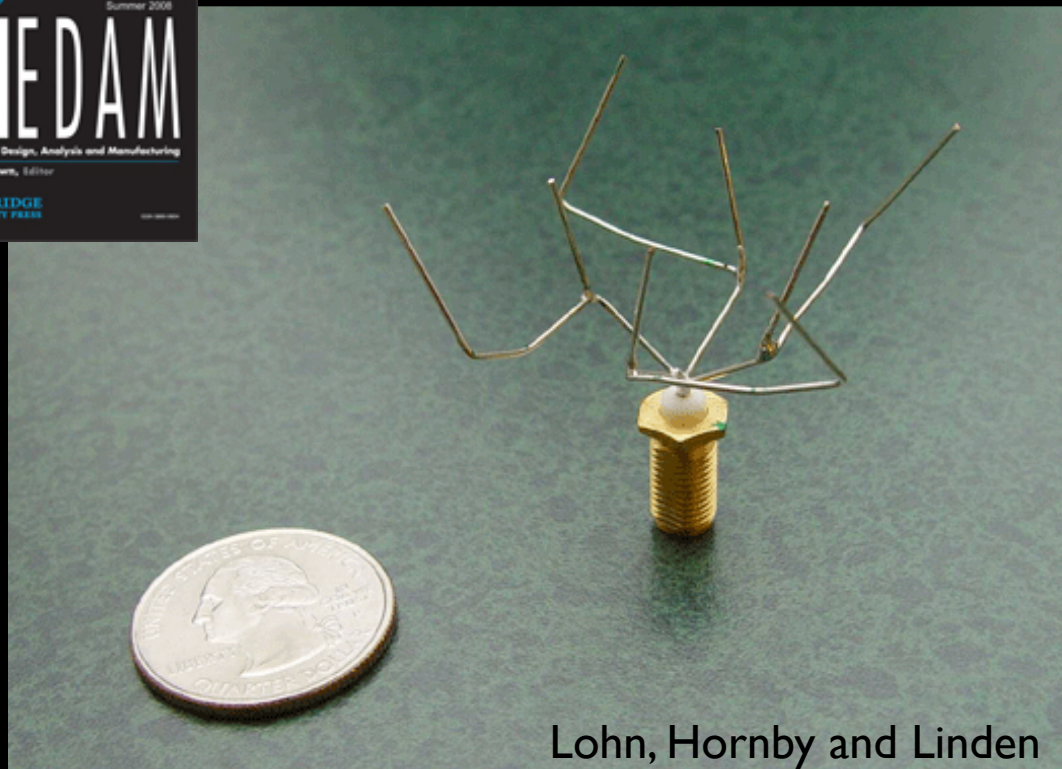
(e)

(f)

(g)



Spector



Lohn, Hornby and Linden

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, ...).

Mutating Lisp

```
(+ (* X Y)
   (+ 4 (- Z 23)))
```

```
(+ (* X Y)
   (+ 4 (- Z 23)))
```

```
(+ (- (+ 2 2) Z)
   (+ 4 (- Z 23)))
```

Recombining Lisp

Parent 1: (+ (* **X Y**)
 (+ 4 (- Z 23)))

Parent 2: (- (* 17 (+ 2 X))
 (* (- (* **2 Z**) **1**)
 (+ 14 (/ Y X))))

Child 1: (+ (- (* **2 Z**) **1**)
 (+ 4 (- Z 23)))

Child 2: (- (* 17 (+ 2 X))
 (* (* **X Y**)
 (+ 14 (/ Y X))))

Symbolic Regression

Given a set of data points, evolve a program that produces y from x .

Primordial ooze: +, -, *, %, x, 0.1

Fitness = error (smaller is better)

GP Parameters

Maximum number of Generations: 51

Size of Population: 1000

Maximum depth of new individuals: 6

Maximum depth of new subtrees for mutants: 4

Maximum depth of individuals after crossover: 17

Fitness-proportionate reproduction fraction: 0.1

Crossover at any point fraction: 0.3

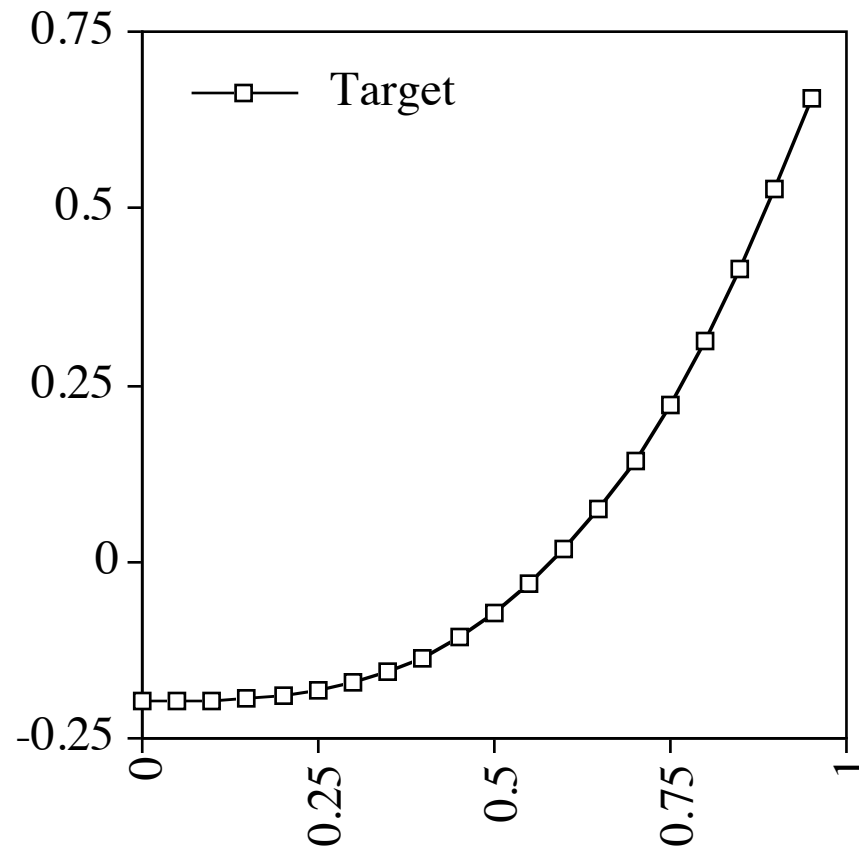
Crossover at function points fraction: 0.5

Selection method: FITNESS-PROPORTIONATE

Generation method: RAMPED-HALF-AND-HALF

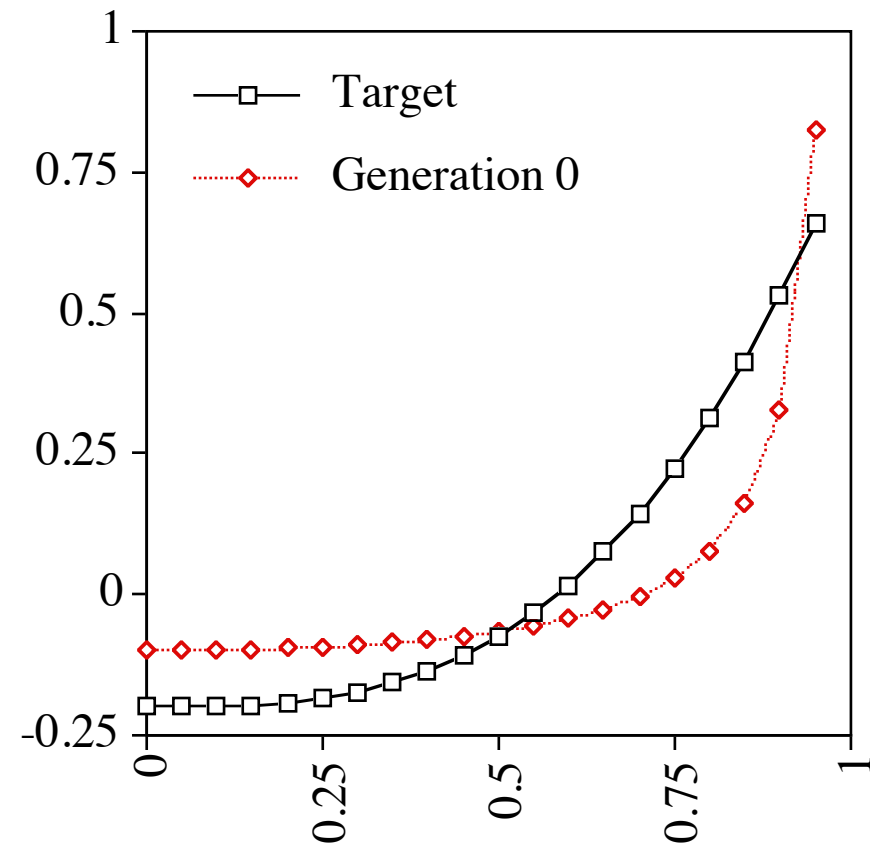
Randomizer seed: 1.2

Evolving $y = x^3 - 0.2$



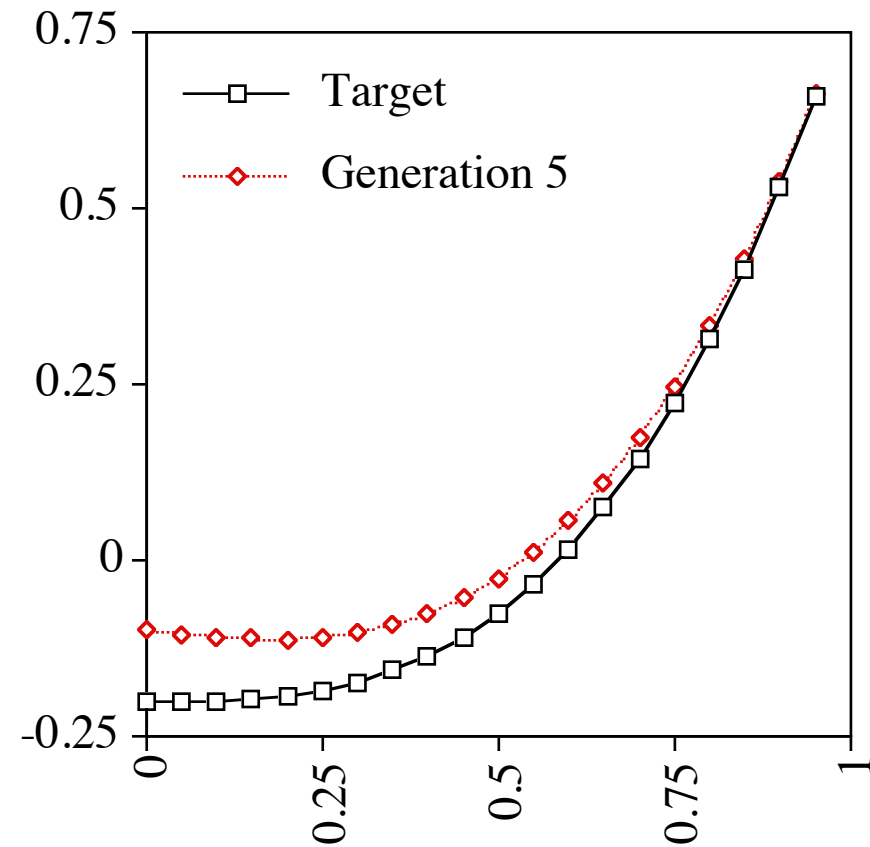
Best Program, Gen 0

```
(- (% (* 0.1
      (* X X) )
  (- (% 0.1 0.1)
      (* X X) ) )
0.1)
```



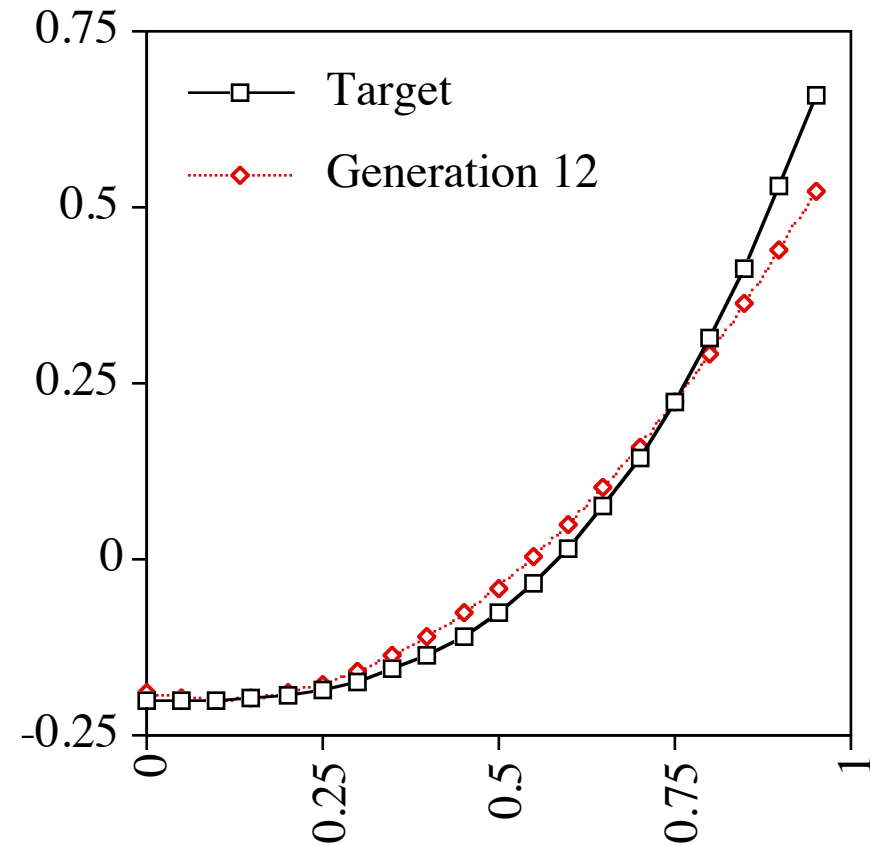
Best Program, Gen 5

```
(- (* (* (% X 0.1)
          (* 0.1 X))
   (- X
      (% 0.1 X)))
0.1)
```



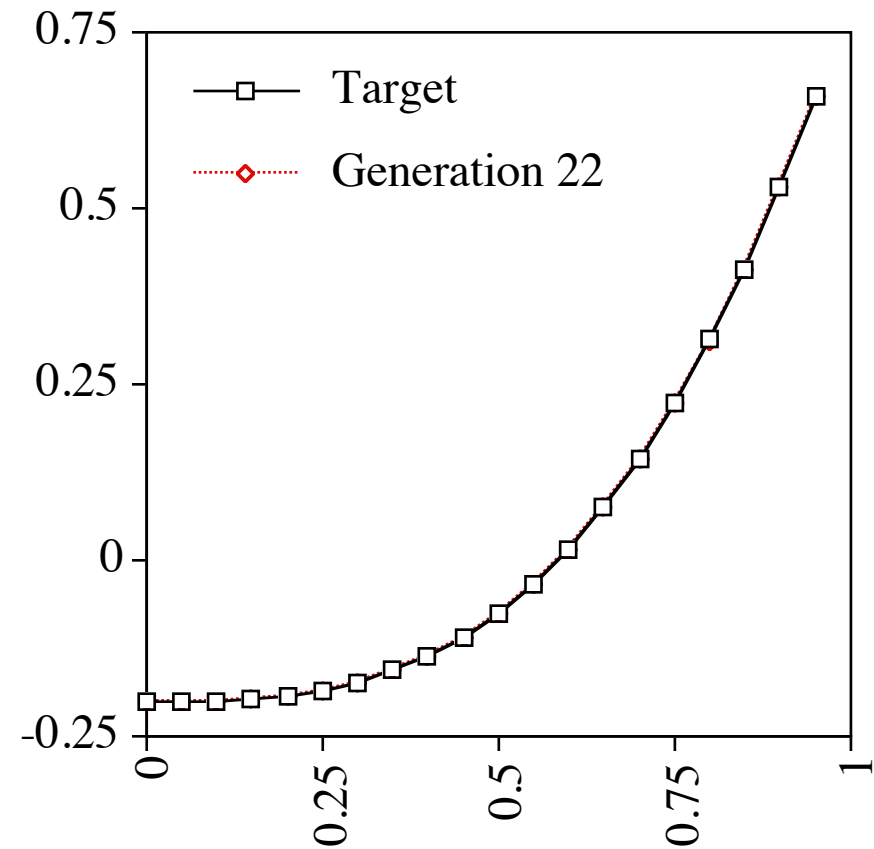
Best Program, Gen 12

```
(+ (- (- 0.1
      (- 0.1
        (- (* X X)
          (+ 0.1
            (- 0.1
              (* 0.1
                0.1)))))))
(* X
  (* (% 0.1
      (% (* (* (- 0.1 0.1)
              (+ X
                (- 0.1 0.1)))
        X)
      (+ X (+ (- X 0.1)
              (* X X))))))
  (+ 0.1 (+ 0.1 X))))
(* X X))
```



Best Program, Gen 22

```
(- (- (* X (* X X))  
      0.1)  
  0.1)
```



Genetic Programming for Finite Algebras

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Humies 2008
GOLD MEDAL

Goal

- Find finite algebra terms that have certain special properties
- For decades there was no way to produce these terms in general, short of exhaustive search
- Current best methods produce enormous terms

Significance, Time

	Uninformed Search Expected Time (Trials)
3 element algebras Mal'cev Pixley/majority discriminator	5 seconds ($3^{15} \approx 10^7$) 1 hour ($3^{21} \approx 10^{10}$) 1 month ($3^{27} \approx 10^{13}$)
4 element algebras Mal'cev Pixley/majority discriminator	10^3 years ($4^{28} \approx 10^{17}$) 10^{10} years ($4^{40} \approx 10^{24}$) 10^{24} years ($4^{64} \approx 10^{38}$)

Significance, Time

	Uninformed Search Expected Time (Trials)	GP Time
3 element algebras Mal'cev Pixley/majority discriminator	5 seconds ($3^{15} \approx 10^7$) 1 hour ($3^{21} \approx 10^{10}$) 1 month ($3^{27} \approx 10^{13}$)	1 minute 3 minutes 5 minutes
4 element algebras Mal'cev Pixley/majority discriminator	10^3 years ($4^{28} \approx 10^{17}$) 10^{10} years ($4^{40} \approx 10^{24}$) 10^{24} years ($4^{64} \approx 10^{38}$)	30 minutes 2 hours ?

Significance, Size

Term Type	Primality Theorem
Mal'cev	10,060,219
Majority	6,847,499
Pixley	1,257,556,499
Discriminator	12,575,109

(for A_1)

Significance, Size

Term Type	Primality Theorem	GP
Mal'cev	10,060,219	12
Majority	6,847,499	49
Pixley	1,257,556,499	59
Discriminator	12,575,109	39

(for A_1)

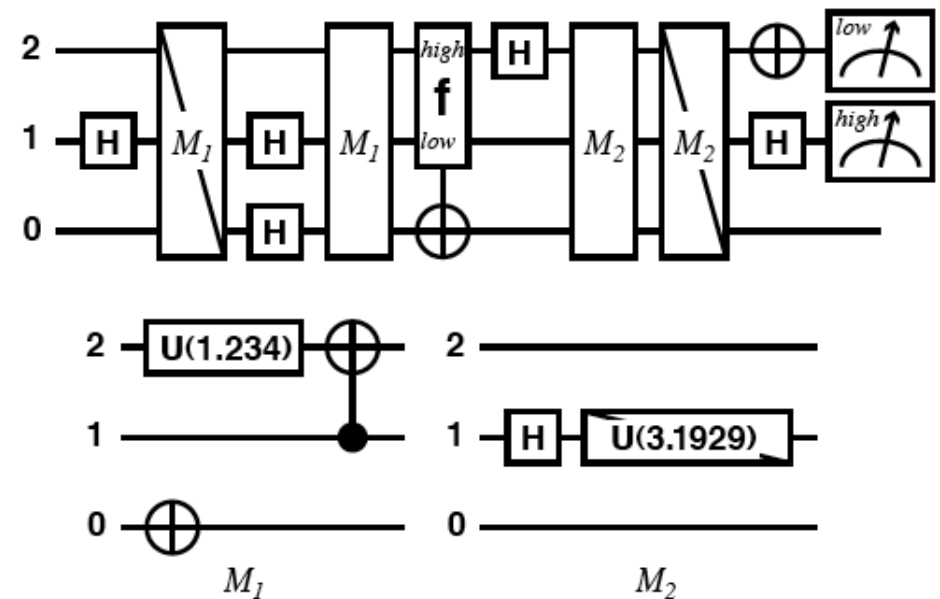
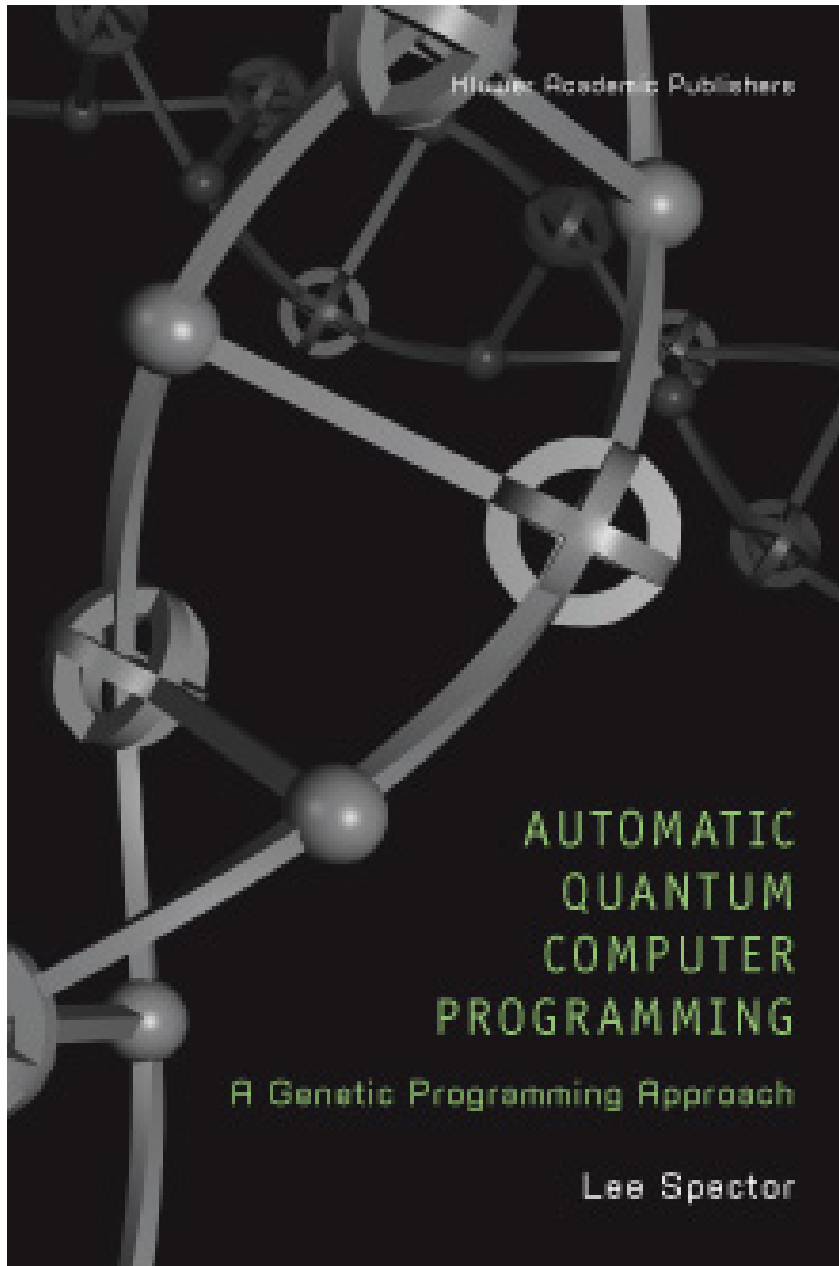


Figure 8.7. A gate array diagram for an evolved version of Grover's database search algorithm for a 4-item database. The full gate array is shown at the top, with M_1 and M_2 standing for the smaller gate arrays shown at the bottom. A diagonal line through a gate symbol indicates that the matrix for the gate is transposed. The "f" gate is the oracle.

Humies 2004
GOLD MEDAL

Evolving Modular Programs

With “automatically defined functions”

- All programs in the population have the same, pre-specified architecture
- Genetic operators respect that architecture
- Complicated, brittle, limited...
- Architecture-altering operations: more so

Evolving Modular Programs

With “execution stack manipulation”

- Code queued for execution is stored on an “execution stack”
- Allow programs to duplicate and manipulate code that on the stack
- Simple types and uses of modules can be evolved easily
- Does not scale well to large/complex systems

Evolving Modular Programs

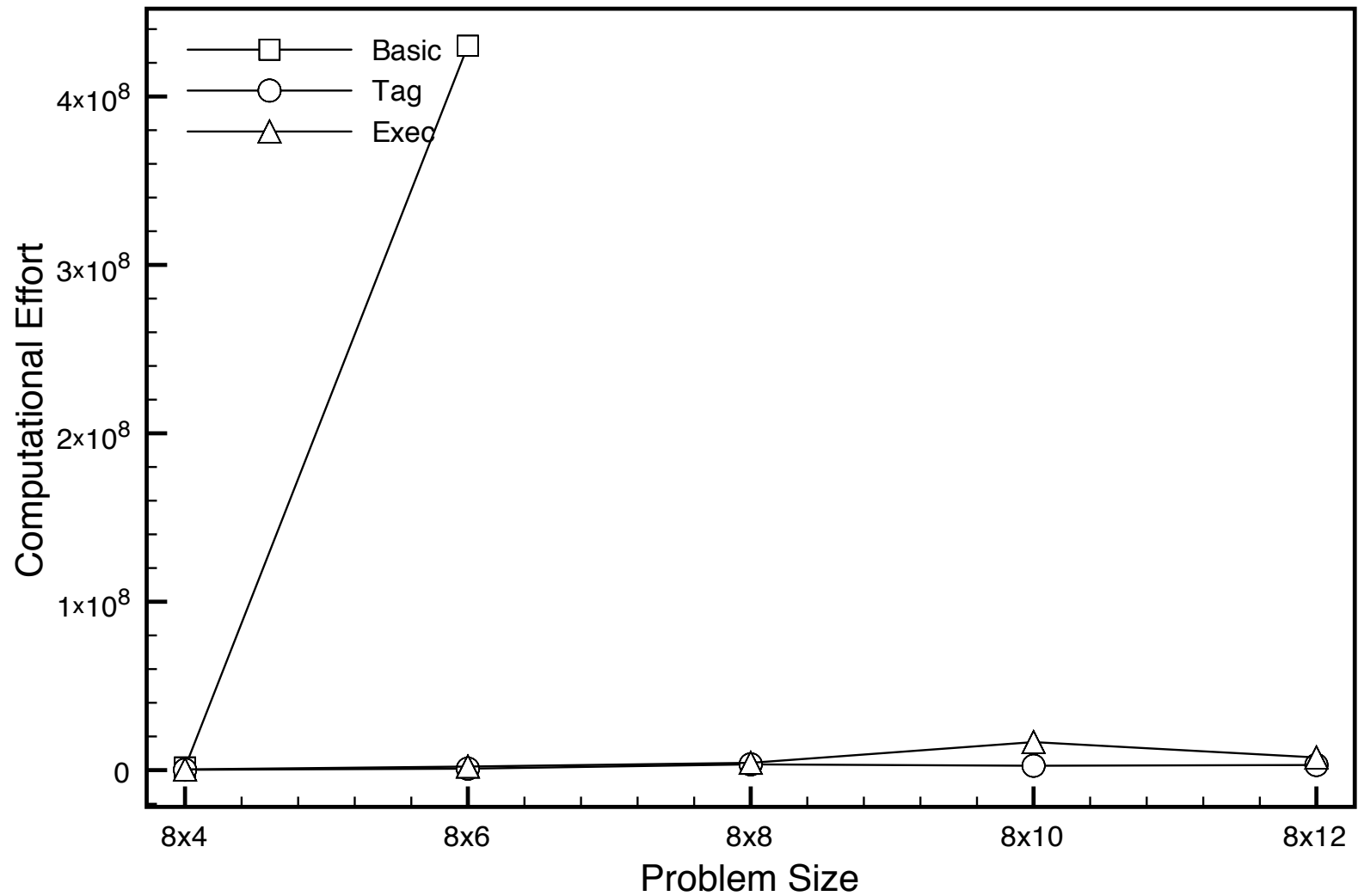
With tags

- 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

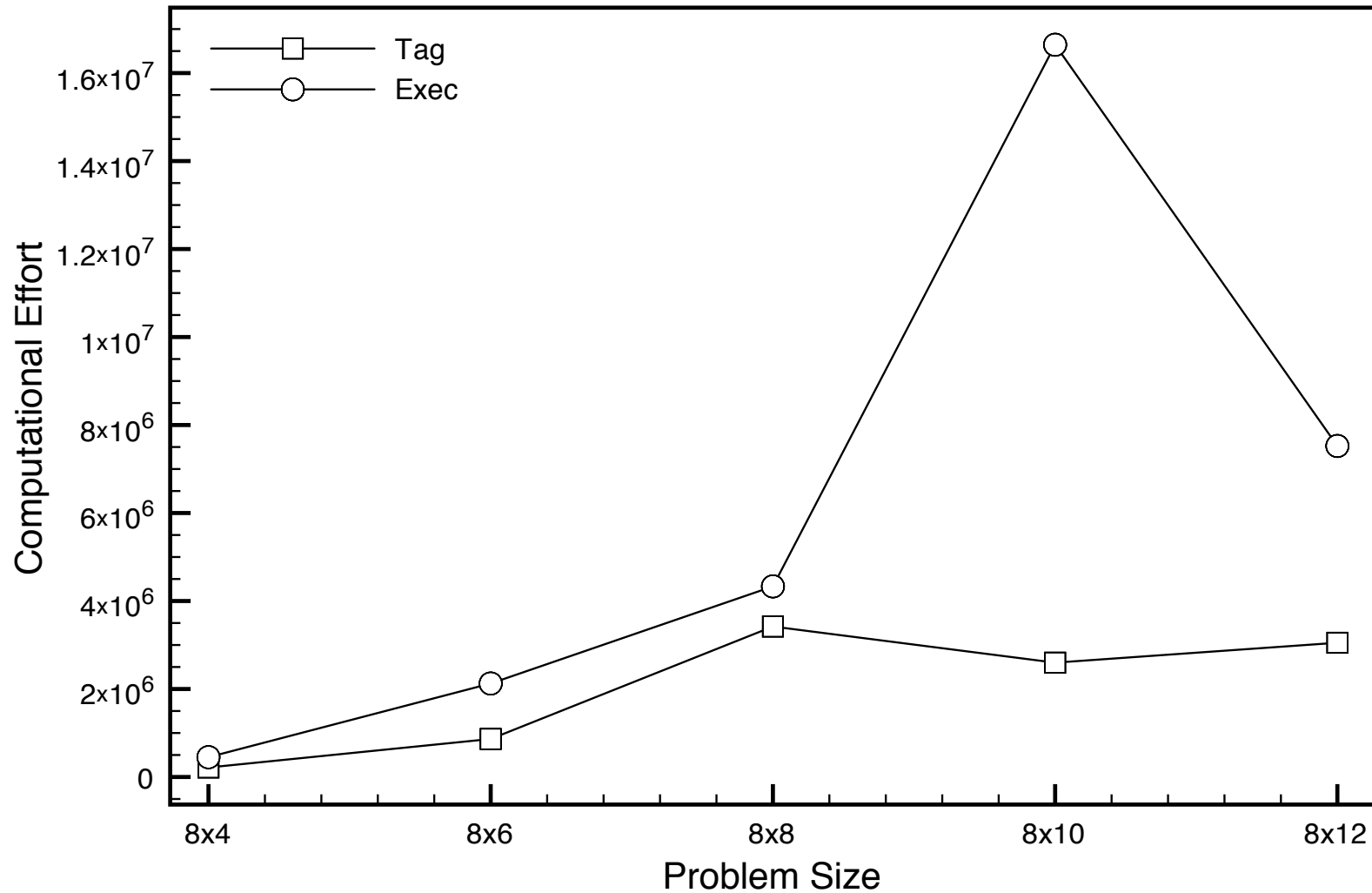
DSOAR Instructions

Condition	Instructions
Basic	if-dirty, if-obstacle, left, mop, v8a, frog, \mathcal{R}_{v8}
Tag	if-dirty, if-obstacle, left, mop, v8a, frog, \mathcal{R}_{v8} , tag.exec.[1000], tagged.[1000]
Exec	if-dirty, if-obstacle, left, mop, v8a, frog, \mathcal{R}_{v8} , exec.dup, exec.pop, exec.rot, exec.swap, exec.k, exec.s, exec.y

DSOAR Effort

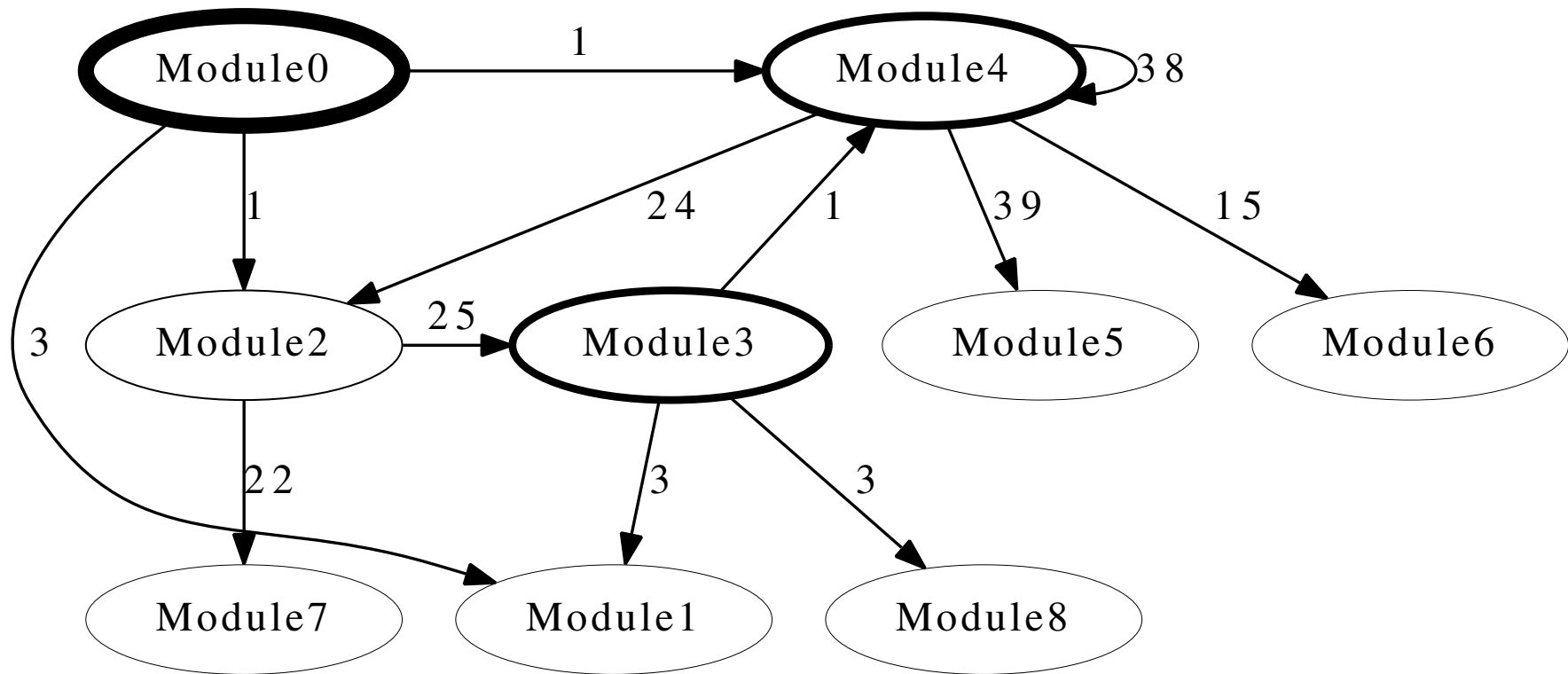


DSOAR Effort



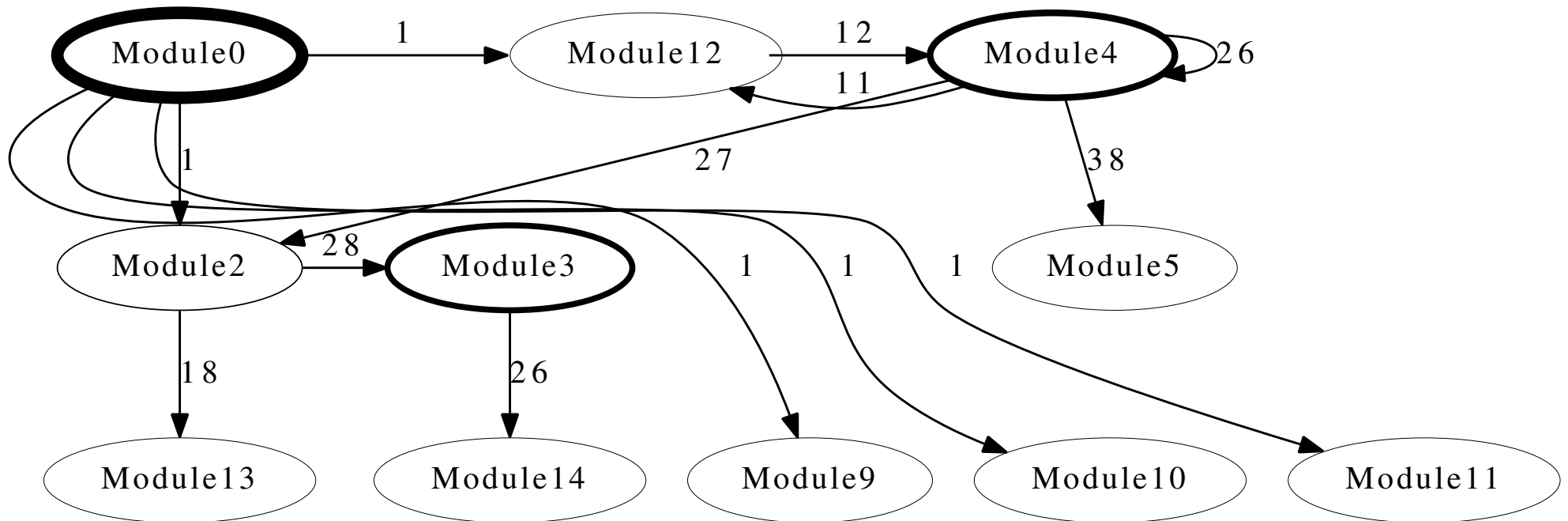
Evolved DSOAR

Architecture (in one environment)



Evolved DSOAR

Architecture (in another environment)



Conclusions

- Tags provide an effective mechanism for the evolution of modular programs that solve difficult problems
- Tags may provide or explain mechanisms that support the evolution of modularity in a range of other systems, both natural and artificial