

Framsticks

genetic representations

comparison experiments

Maciej Komosinski and Adam Rotaru-Varga. Comparison of different genotype encodings for simulated 3D agents. *Artificial Life Journal*, 7(4):395-418, Fall 2001. [\[view pdf\]](#)

Goals

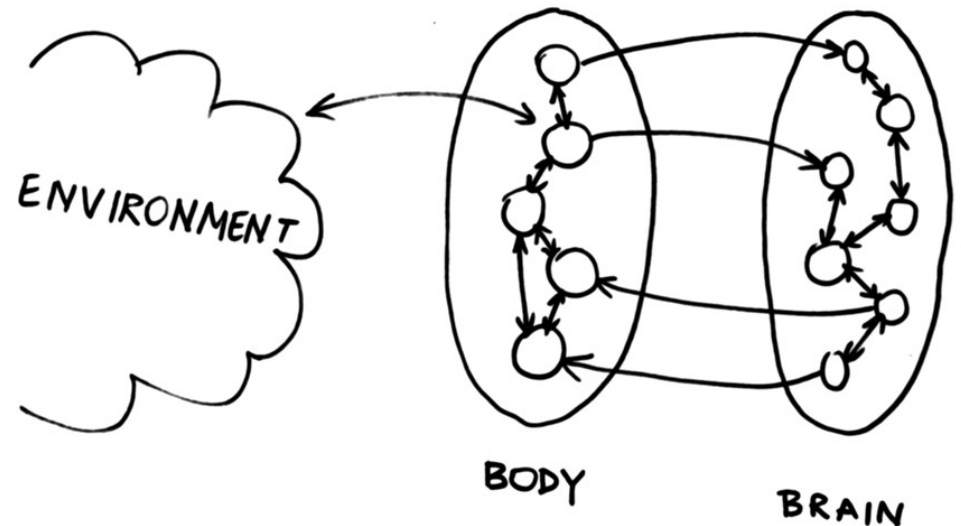
- co-evolve bodies and brains
- design various methods of description of body and brain
- study and compare the effectiveness of evolution using these methods
- in a single system.

Why to co-evolve brains and bodies?

- because it yields better results than with body separated from brain
- because it is natural
- embodiment: physical interactions (between body parts, signal processing) perform computations, a part of overall behavior
- brain and body strongly connected
 - evolution of body changes the cognitive space of the brain (e.g.: an eye placed on a limb, new senses)
 - evolution of brain changes usage of the body
- co-evolution: can cause change even in the absence of environmental change

What is the trouble?

- the ‘matching’ problem
 - parts of brain (neurons, nodes) must be connected to parts of body (sensors, actuators)
 - if matching is explicit, it can be disrupted by the change of either side, which can be catastrophic
- both are variable size
- crossover on complex representations



General problems in optimization of realistic autonomous agents

- Infinite search space
- Discrete-continuous space
- Hard to define neighborhood
- Solutions contain varying amount of information
- Hard to choose representation
- Very strong dependencies and connections between parts of a solution
- Evaluation function with many local optima
- Many non-feasible solutions and diverse constraints
- Non-determinism and complexity of evaluation
- Multi-criteria evaluation, complex definition of criteria, evaluation delayed to action
- Hard to estimate the time needed for evaluation and optimization

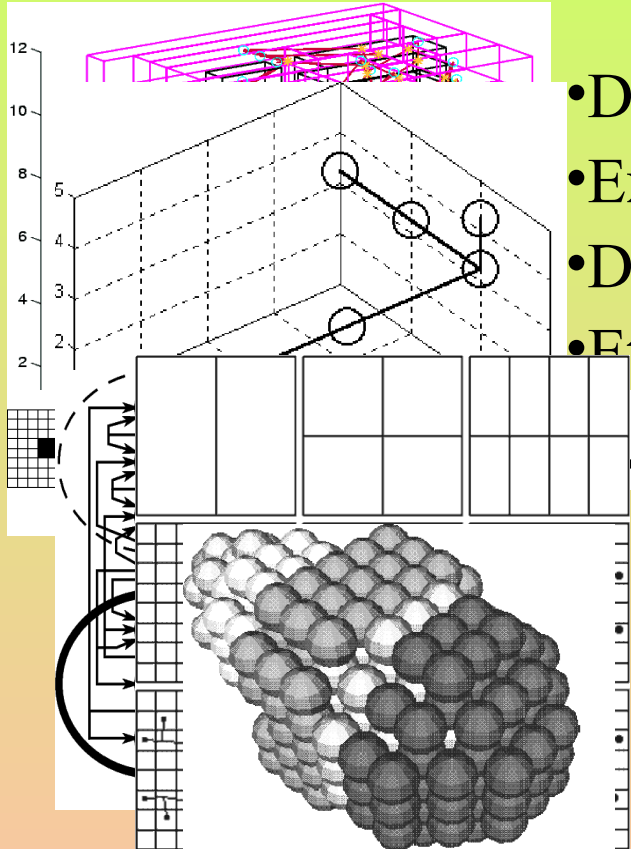
...the big problem is size and nature of the search space

Genetics is important because each representation and its operators

- establishes different structure and order in the search space
- defines important information and ‘building blocks’ in another way
- is scalable in a different degree
- introduces different bias which leads to finding qualitatively different solutions
- imposes diverse local optima and displays various levels of robustness against being trapped into them
- can limit the space of valid solutions in a particular way
- has a specific degree of coherency, redundancy, easiness of interpretation, etc.

Current artificial genomes – very diverse

Simple (direct)



- Different base
- Explicit / implicit
- Different systems
- Etc...

No comparison possible!

What is the best representation?

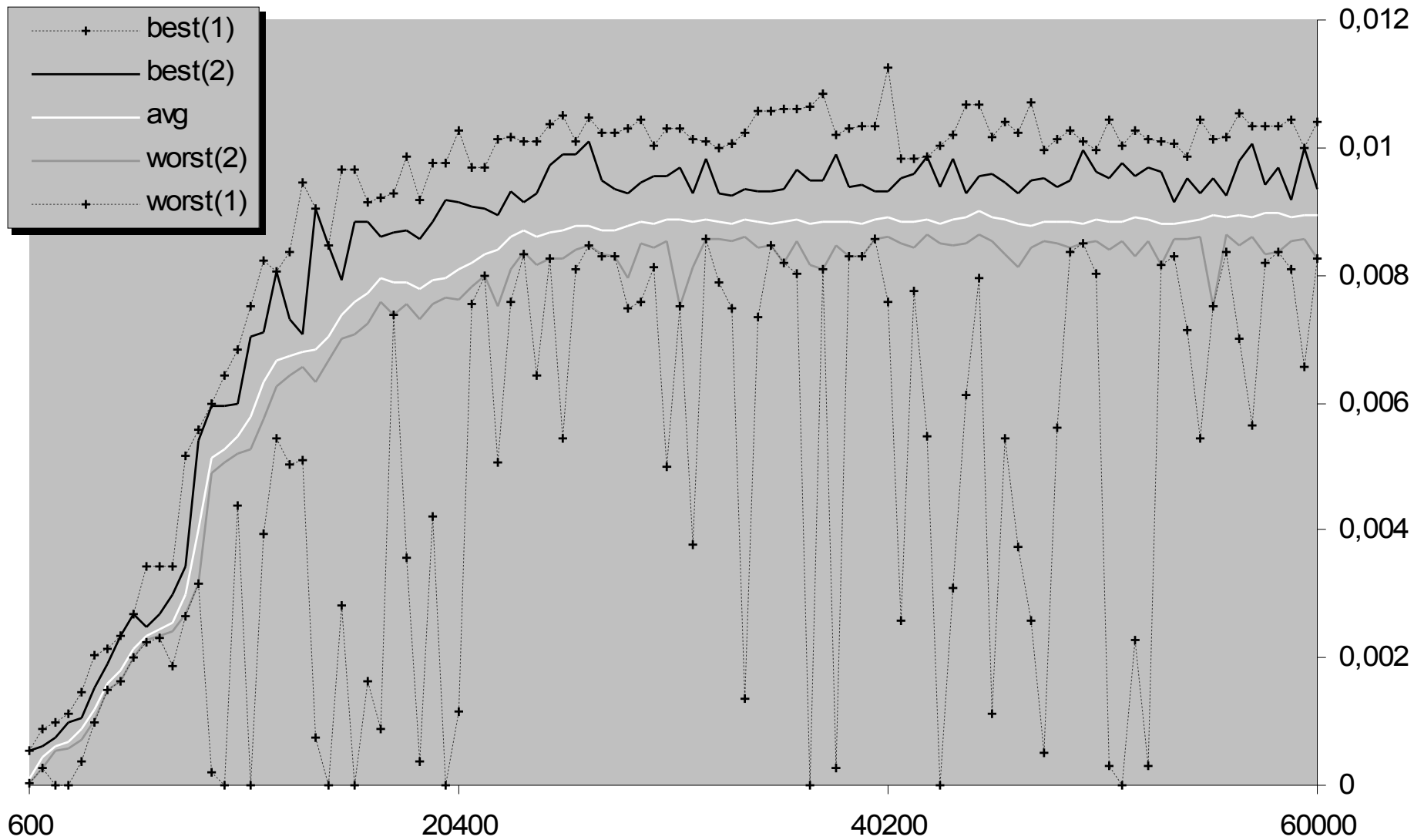
- Fitness values
- Nature of solutions
- Simplicity
- Understandability

Complex (biologically inspired)

Experiments

- 3 one-criterion tasks
 - Average height of agent center (maximize; NN turned off)
 - Average height of agent center (maximize; NN turned on)
 - Average velocity (maximize)
- 10 runs for each task and each genotype format (*simul*, *recur*, *devel*)
- 90 runs in total
- System main parameters
 - Steady-state
 - Population size: 200
 - Cloning probability: 20%
 - Crossing over prob.: 16%
 - Mutation prob.: 64%
 - Stabilization period

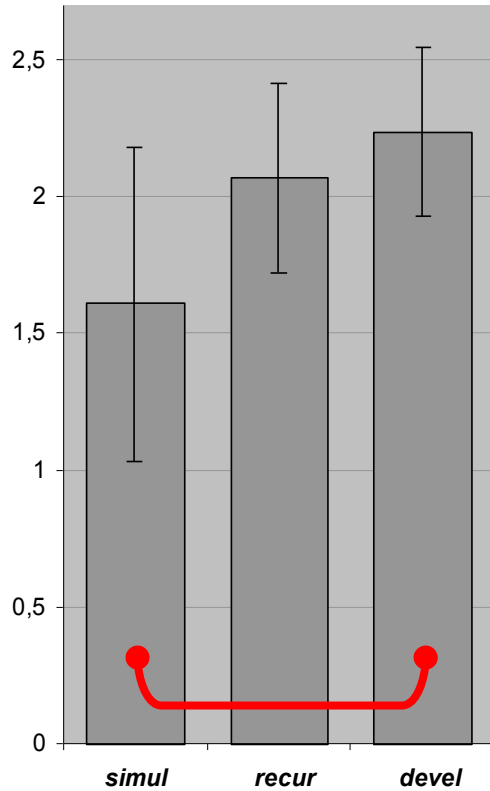
Non-deterministic evaluation



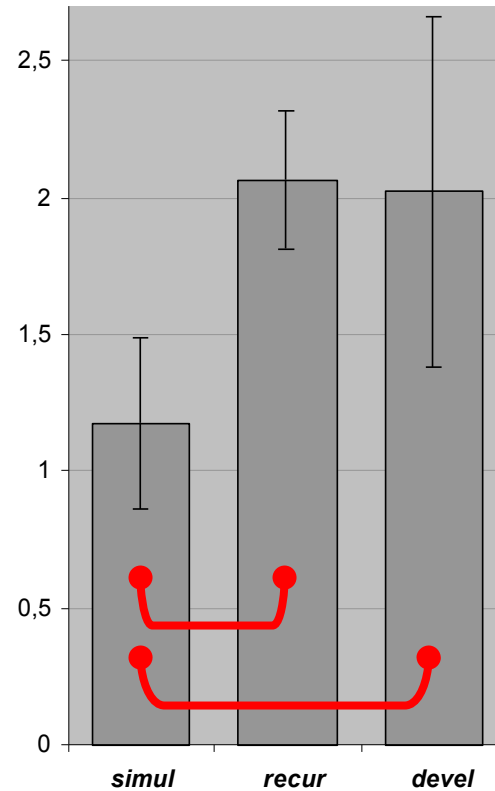
Results (Quantitative)

Genetics experiment

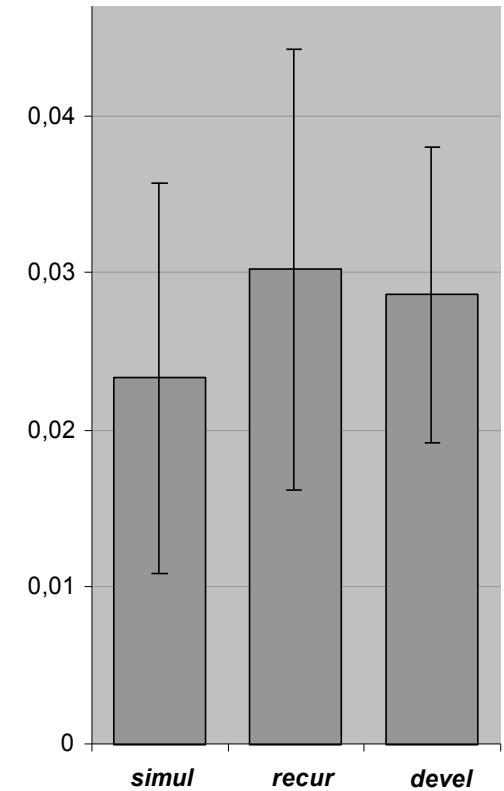
Height passive



Height active



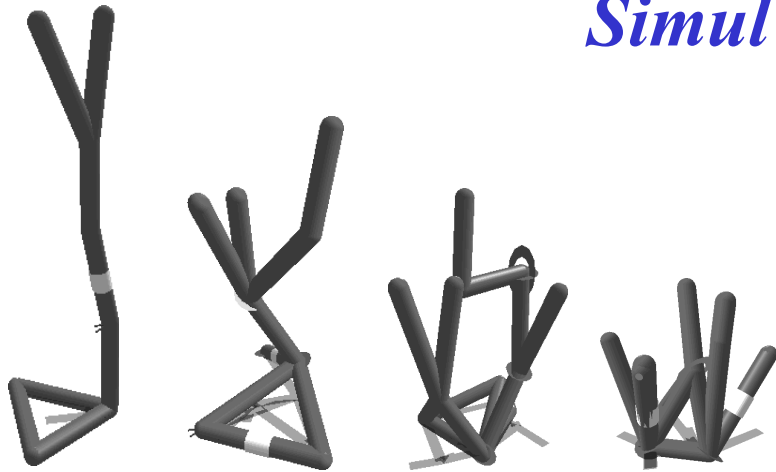
Speed



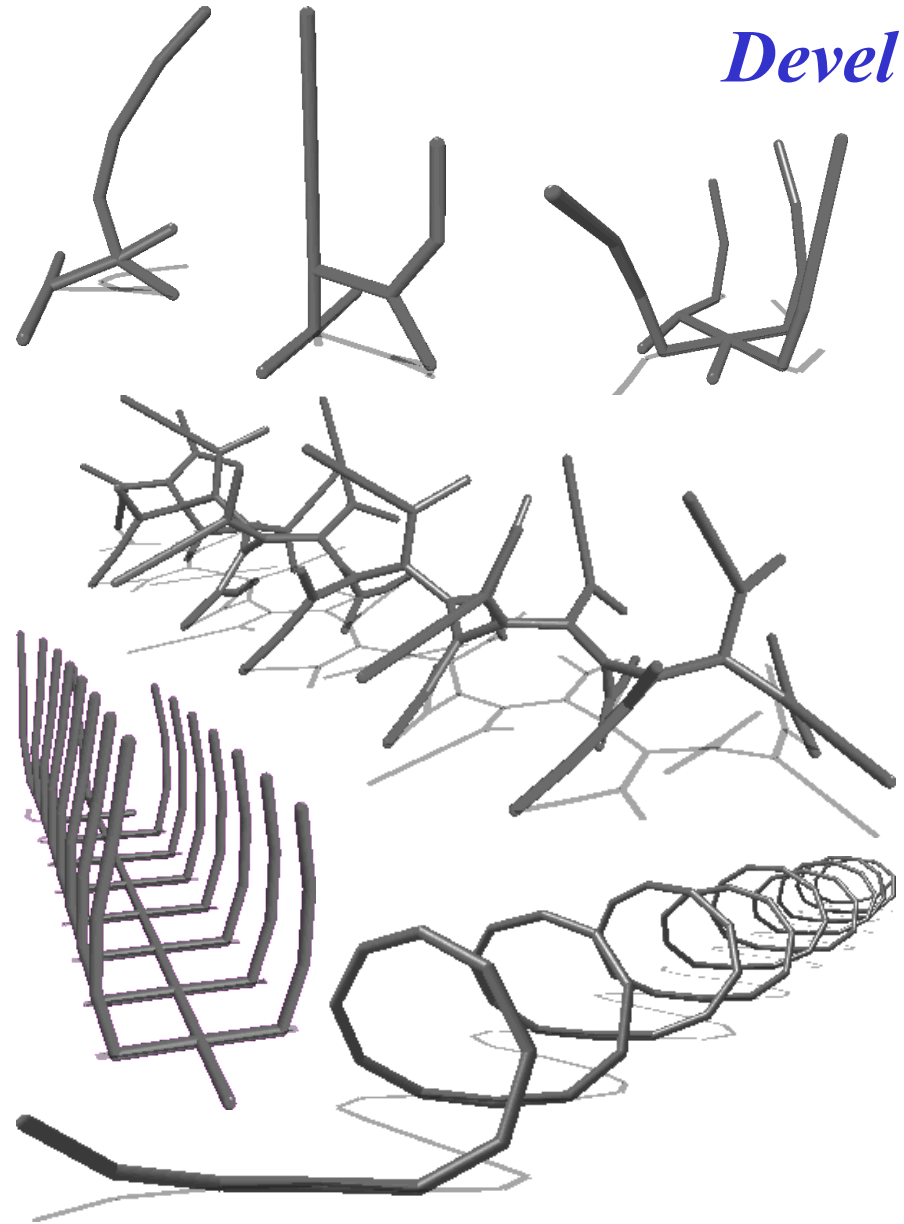
Results (Qualitative; height passive)

Genetics experiment

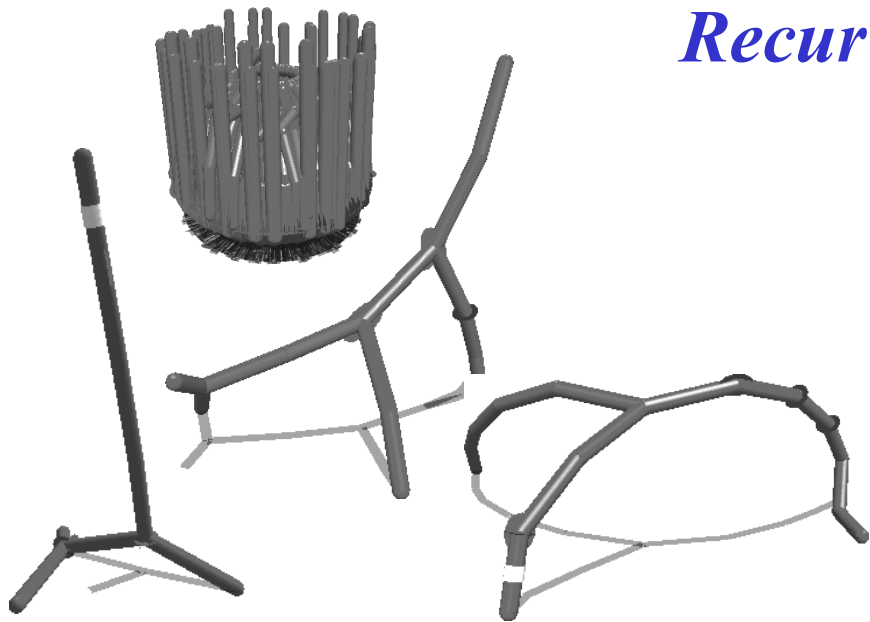
Simul



Devel



Recur



Conclusions

- *Simul* representation with full abilities of expressing agents was the worst one
- The limitation of the search space by higher level representations has not deteriorate results, but has improved them
- The most advanced *devel* encoding was not significantly better than *recur*
- Each higher-level representation introduces a specific bias and new quality (characteristics) into solutions
- For all representations, the best individuals were successful in terms of fitness value. It was difficult or impossible to construct better agents by hand, mainly because of high time costs
- It may be sometimes worthwhile to introduce advanced mechanisms into a representation, in order to obtain different nature of solutions, even when they are not improved in terms of fitness

Conclusions, cont.

- punctuated equilibria
- convergence
- exploitation of simulator imperfections
- redundancy, randomness
- many strong (implicit) dependencies inside agents

