

Genetic Algorithm (Part 3)

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Goal of today

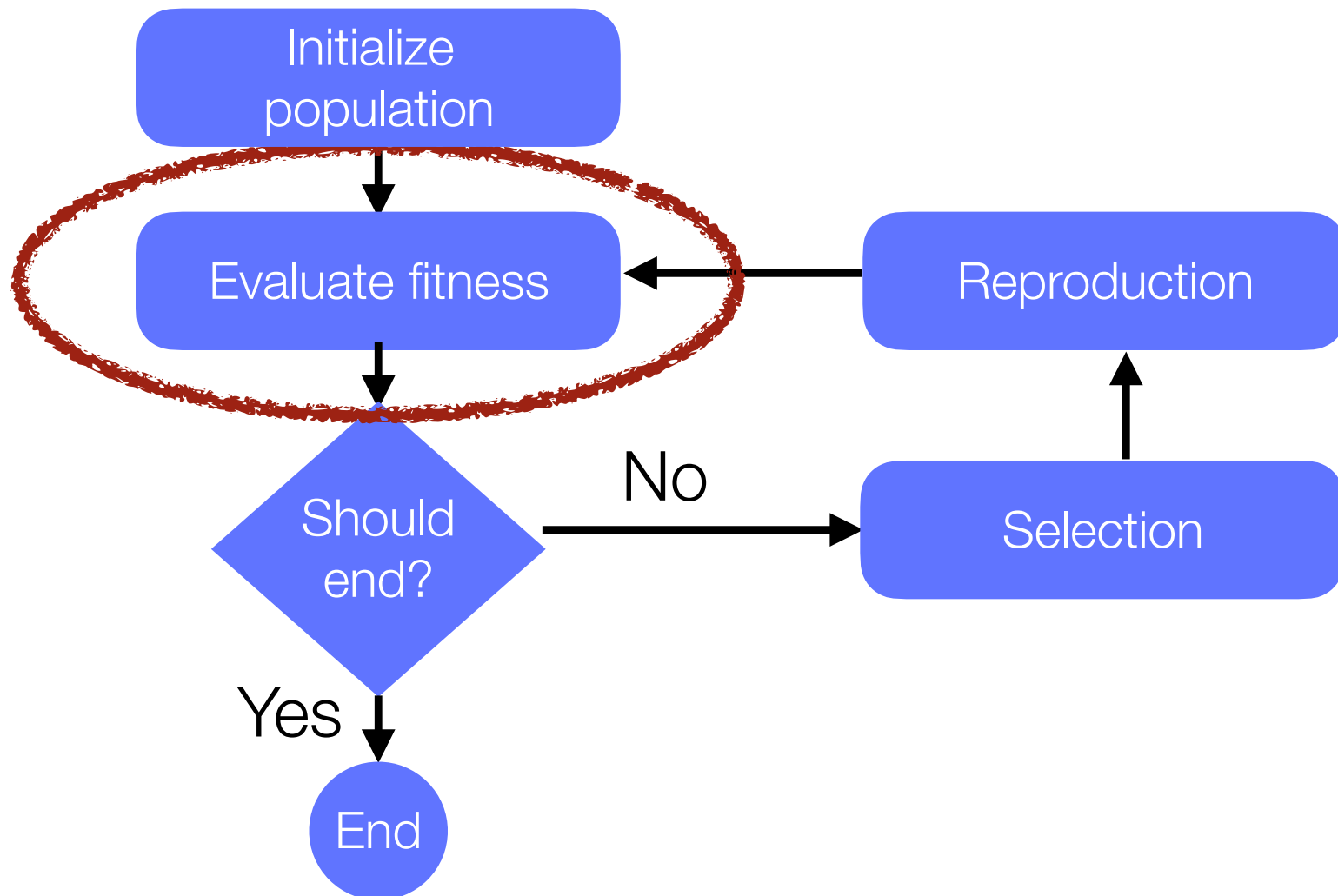
Cover advanced topics: *multi-objective optimization, Pareto front, elitism*

Provide complex examples involving *zoomorphic organisms*

zoomorphic | ,zōə'môrfik |
adjective

having or representing animal forms or gods of animal form: *pottery decorated with anthropomorphic and zoomorphic designs.*

Flow - chart of a genetic algorithm



Fitness function

GA searches for a solution, *without specifying an analytic way* of doing so

This is very appealing since we just need to say how to evaluate a possible solution instead of saying how to get it

However, a fitness function may be complex

It happens that often some constraints have to be specified:

E.g., a robot needs to find the exit *and* find the shortest path

E.g., numbers *cannot be* repeated when solving the knapsack problem

Minimization vs Maximization

For some problems, getting closer to the solution means decreasing a function. E.g.,

Distance between a robot to the exit

Error function

In such a case, the fitness function needs to be expressed as:

- objective

N - objective

$1 / (1 + \text{objective})$

Multi-objective optimization

How to enforce that the robot follows the shortest path?

Two components:

- distance from the exit
- length of the path



Multi-objective optimization

Constraints may be expressed in different ways:

Penalty for violation in the fitness function

Implicit via the *encoding* of the problem

Multi-objective fitness function

Implement a *repair capability* in case of an infeasible /
unsound individual

Multi-objective optimization

A fitness function F can have sub-fitness functions f_1, f_2, \dots, f_n

Do the f_i have the same priority?

Multi-objective optimization

If all the f_i have the very same priority, then we could have

$$F(x) = \min(f_1(x), f_2(x), \dots, f_n(x))$$

Multi-objective optimization

If all the f_i have the very same priority, then we could have

$$F(x) = \min(f_1(x), f_2(x), \dots, f_n(x))$$

Note that we have $F(x) \in \mathbb{R}$

$F(x)$ is a number, therefore comparison is trivial

Multi-objective optimization

Alternative definition of the fitness function exit to cope with multi-objective. The Pareto optimality is a definition

If we have n fitness functions, noted f_1, \dots, f_n

Then we can have $F(x) \in \mathbb{R}^n$, in that case

$$F(x_1) > F(x_2) \iff \forall i . f_i(x_1) \geq f_i(x_2) \wedge \exists i . f_i(x_1) > f_i(x_2)$$

Multi-objective optimization

Considering:

$$F(x_1) > F(x_2) \iff \forall i . f_i(x_1) \geq f_i(x_2) \wedge \exists i . f_i(x_1) > f_i(x_2)$$

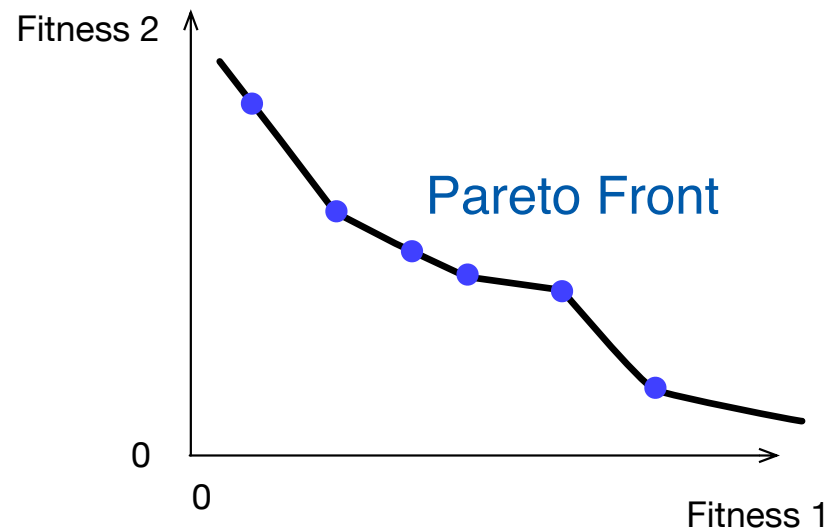
\forall indicates *for all*

\exists indicates *exists*

\wedge indicates *logical and*

Pareto front

Searching for solutions using Pareto optimality may produce a set of solutions that are non-dominated



Multi-objective optimization

The robot has two objectives:

- Finding the exit

- Having a short path

These two objectives have not the same priority

Finding the exit is more important

Multi-objective optimization

Our robot can have the fitness $F(x) \in \mathbb{R}^2$

x represents a path

$F(x) = (d, l)$ in which d is the distance from the exit and l is the number of steps to reach the exit

We therefore want to minimize both d and l , but we minimize l only if $d = 0$

Comparison is done as follow

If $d_1 < d_2$ then $F(x_1) > F(x_2)$

If $d_1 = d_2 \wedge l_1 < l_2$ then $F(x_1) > F(x_2)$

Else $F(x_1) < F(x_2)$

Bad luck?

An *exceptionally good individual* may be created in a particular generation

However, the characteristics of that individual may be lost if

- it is not selected to be parent

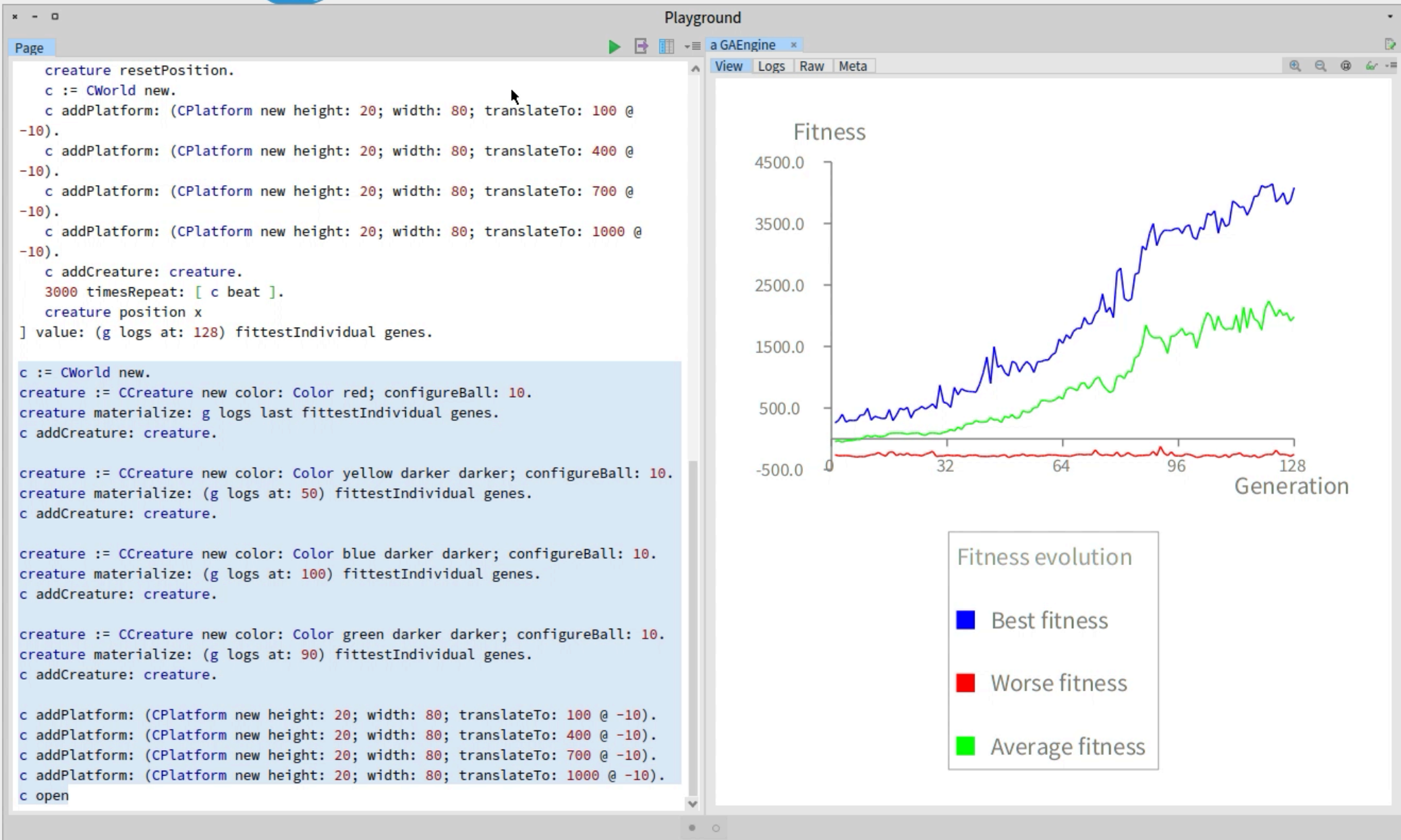
- its children are worse than the parent

Elitism

One common practice is to place in a new generation the best individual from the previous generation

The fitness curve never goes down

Each generation represents a progress if any

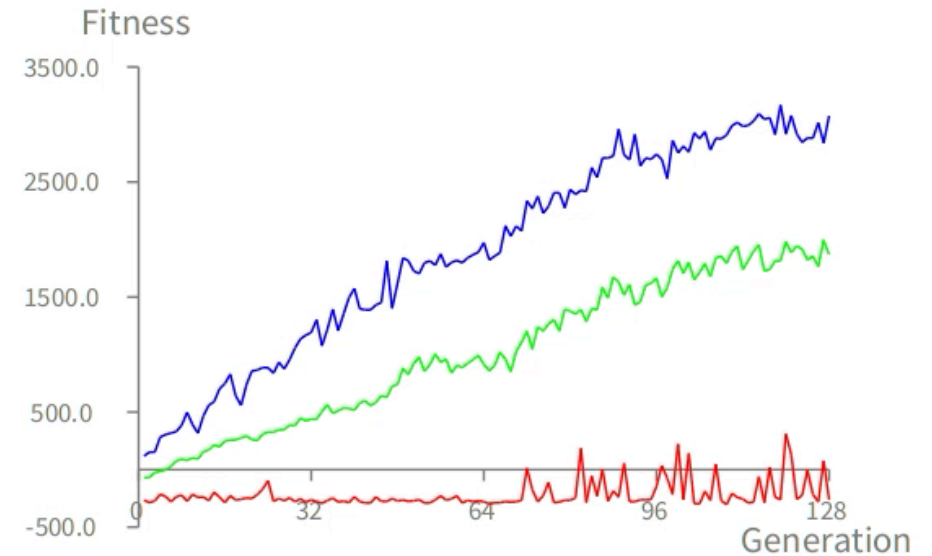


```

g selection: (tournamentSelection new).
g mutationRate: 0.02.
g endForMaxNumberOfGeneration: 128.
g populationSize: 100.
g numberOfGenes: numberOfMuscles * 5.
g createGeneBlock: [ :r :index | mg valueForIndex: index ].
g fitnessBlock: [ :genes |
    creature := CCreature new configureBall: numberOfNodes.
    creature materialize: genes.
    creature resetPosition.
    c := CWorld new.
    c addCreature: creature.
    1 to: 25 by: 3 do: [ :x |
        c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x *
100 @ -10).
        c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x *
100 + 50 @ -30).
        c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x *
100 + 100 @ -50).
        c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x *
100 + 150 @ -70).
    ].
    c addCreature: creature.
    3000 timesRepeat: [ c beat ].
    creature position x
].
g run.

```

```
creature := C Creature new configureBall: 10.
creature materialize: g result.
c := CWorld new.
1 to: 25 by: 3 do: [ :x |
    c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x * 100
@ -10).
    c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x * 100
+ 50 @ -30).
    c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x * 100
+ 100 @ -50).
    c addPlatform: (CPlatform new height: 20; width: 80; translateTo: x * 100
+ 150 @ -70).
].
c addCreature: creature.
c open
```

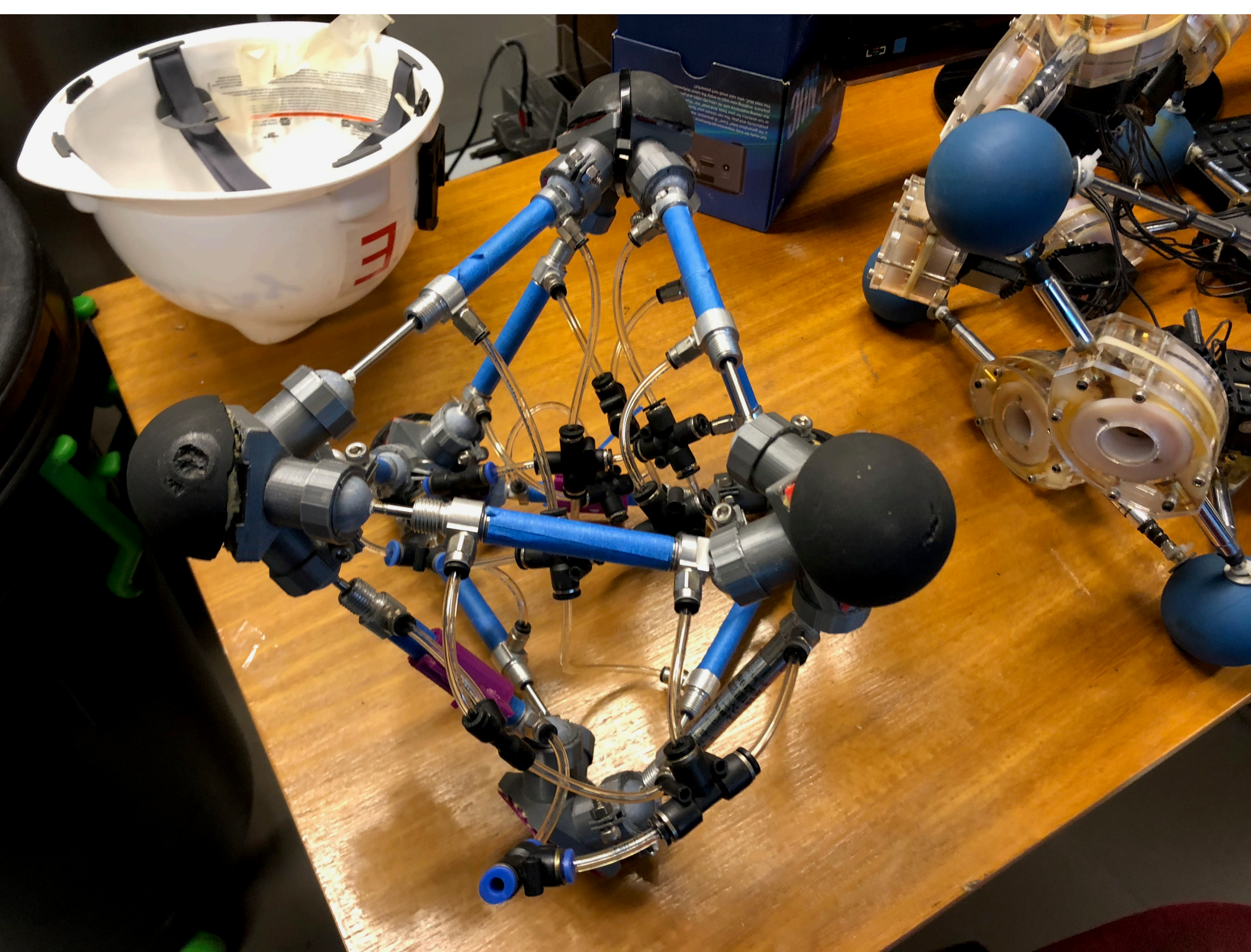


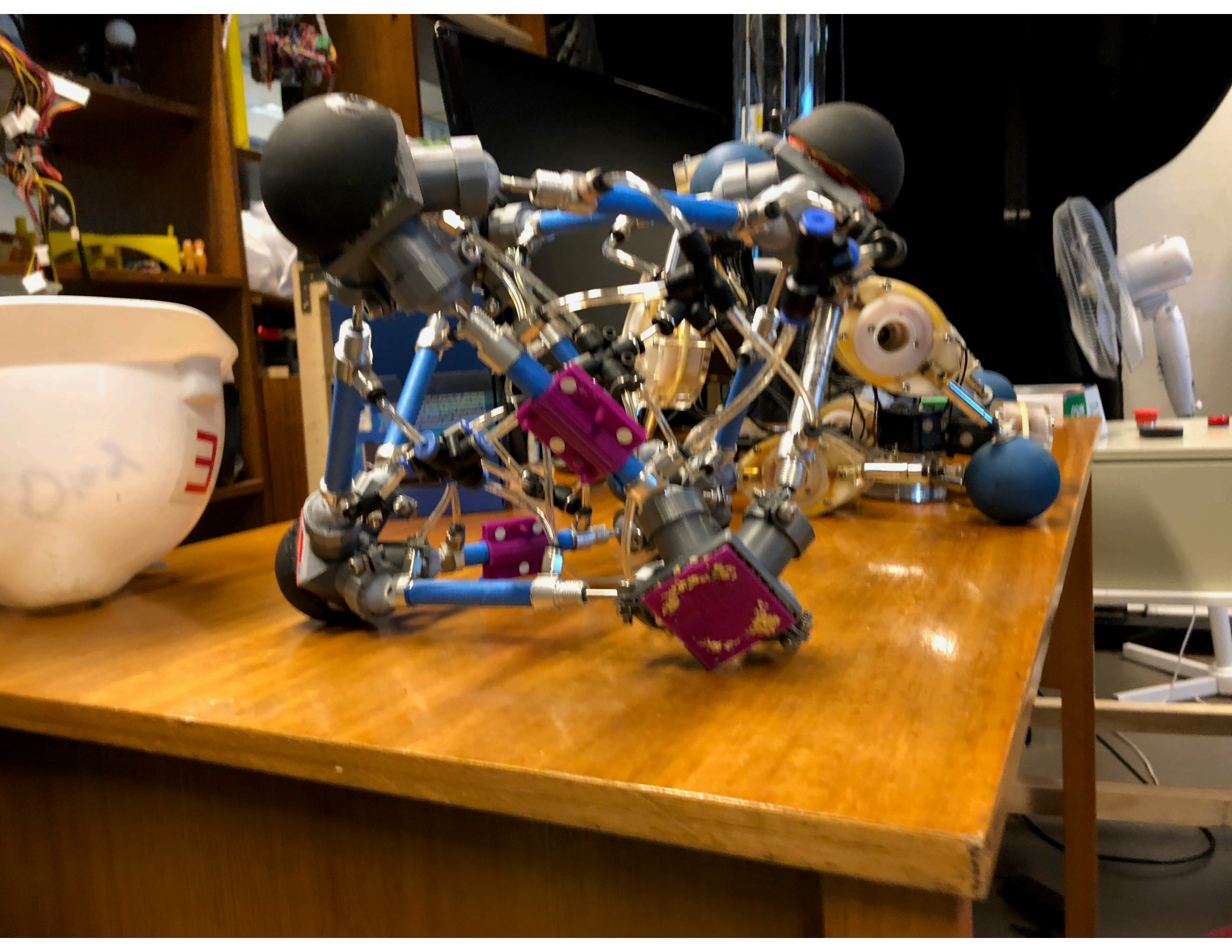
Fitness evolution

■ Best fitness

- Worse fitness

■ Average fitness





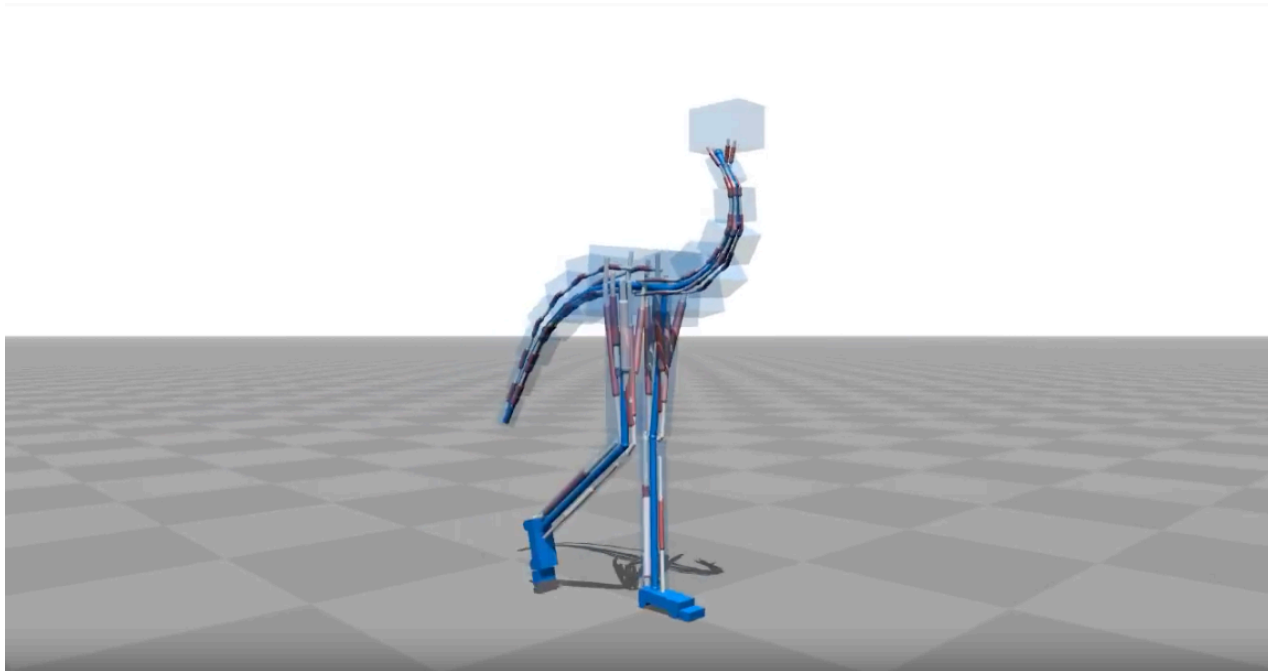
Evolving Virtual Creatures With Genetic Algorithms

<https://www.youtube.com/watch?v=bBt0imn77Zg>



Flexible Muscle-Based Locomotion for Bipedal Creatures

<https://www.youtube.com/watch?v=pgaEE27nsQw>





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