Evolutionary robotics

Don't code, evolve!

What is evolution?

"Definition:

[1] The process by which different kinds of living organism are believed to have developed from earlier forms during the history of the earth.

[2] The gradual development of something"

- Based on Darwin's evolution theory
 - Reproduction is the key to life
 - Better fitted (adapted) individuals have better chance to reproduce (i.e. survive)
 - Successful phenotype* traits are reproduced, modified, combined

* phenotype - The set of observable characteristics of an individual resulting from the interaction of its genotype with the environment.

How to represent an individual?

- Each individual has its own genotype \rightarrow representation of the individual
 - Same as genes in people
- In programming this can be a lot of things
- The simplest problems might have individuals represented as a binary number, hard problems might have complex neural nets
- One of the simplest evolutionary problems is $Max1 \rightarrow evolve$ an individual with all 1s, starting with a random population
- Simple robot problems can be modeled by using neural nets which for example map the robot's input sensors to signals for its two wheels

Glossary

<u>Fitness</u>

"Score" of an individual. How good the individual is amongst the population. Survival of the fittest -> the higher fitness one has, the more likely their genes are to be passed on.

Fitness function

Function determining fitness of each individual in the population.

Recombination / Cross over

Altering the genes of multiple individuals by combining parts of their genotypes.

<u>Mutation</u>

Pseudo random altering of genes of a single individual.

<u>Selection</u>

Choosing which individuals continue to the next generation, which recombine etc.

Evolution

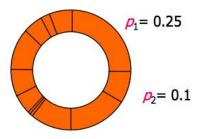
- Evolutionary algorithms are de facto population based stochastic search algorithms
 - We are non-deterministically looking for a sufficient individual in a given population
- Recombination and mutation create variability
 - Also prevent the evolution from getting stuck
- Selection leads the search in the right direction
 - Only the better traits are preserved

Simple genetic algorithm

- In time *t* = 0 generate a random initial population P(0) of n m-bit genes (individuals)
- From P(t) to P(t+1)
 - Compute fitness for each individual from the population
 - Repeat n/2
 - Select a pair x, y from the population
 - Cross over x, y with probability p_c
 - Mutate every bit of x and y with probability p_m
 - Insert x, y to P(t+1)
- Different settings and probabilities of course yield different results
- Multiple crossover, mutation and selection options

Selection

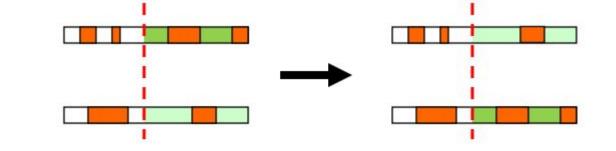
- Based on the individual's fitness
- Basic: Roulette wheel
 - Each individual occupies a slice of the roulette based on its fitness with regard to the sum of all the fitnesses
 - The roulette is spun n-times



- Other selection types: Tournament, rank based
- Possible elitism (the *m* best individuals are preserved)

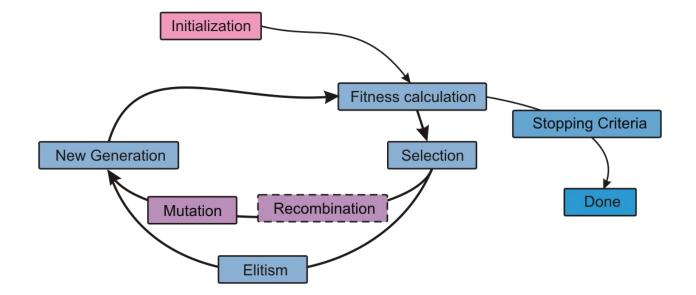
Cross over

• Single point



• Multi point

Evolution in a picture



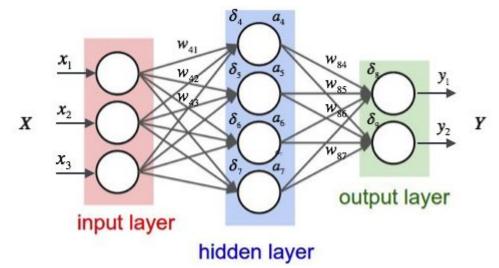
Great example - https://www.youtube.com/watch?v=GOFws_hhZs8&t=254s

Evolution parameters

- Each problem has its own representation
 - Have to choose proper cross-over, mutation and mainly fitness function
- Mutation regarding neural nets
 - Change in a weight of a synapse
 - Change of the type of the activation function
 - Adding/removing synapses
 - Adding/removing neurons
- Other important parameters
 - Population size
 - Probability of a cross-over
 - Probability of a mutation
 - Size of the elite group

Simple neural net

- Possible model for a robot
 - Output layer of 2 neurons represents two wheel signals
 - Input layer represents robot's sensors



Line follower evolution

for Arduino robot

Line follower problem

- Something like Hello World for new programmers
- The goal of the exercise is to develop a robot that can properly follow a black line on a white sheet of paper (with harder version which can have turns and signals etc.)
- Not a trivial task but a good one to introduce the robotics to the newcomers
- Programming this robot takes time and people often have to learn new technologies because of the robot's architecture

• Can lazy people solve it with evolution?

Goals

- implement a simulator for an
 Arduino robot used in the subject
 Introduction to Robotics (NAIL028)
- develop a line following robot using evolution
- somehow port to the real robot?



Motivation

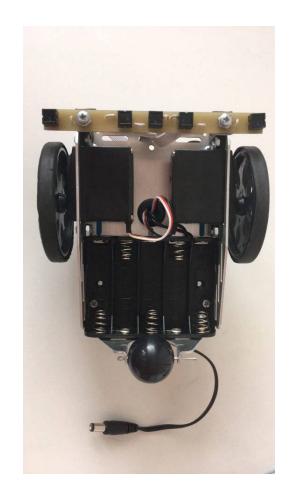


What we worked with

- Arduino Parallax Boe-Bot
- JavaScript/Webpack/Node.js
- Neataptic.js library
 - > Implemented NEAT algorithm in JavaScript

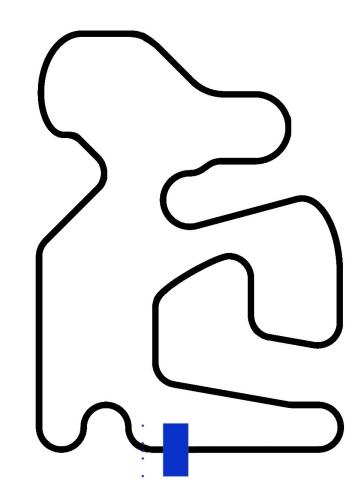
Robot

- 5 sensors on a ramp
 - digital input (0/1 ~ black/white)
 - > analog input (0–1000 ~ grayscale)
- 2 wheels
 - possible values: 1300–1700 (from full clockwise to full counterclockwise)
 - ➤ examples
 - (1300, 1700) \rightarrow robot moves forward at full speed
 - (1700, 1300) \rightarrow robot moves backward at full speed
 - (1700, 1700) \rightarrow robot rotates in place
 - (1500, 1500) \rightarrow robot does not move



Simulator

- custom made in Javascript
- tailored to the Arduino robot
- users can use any black and white image (with size specification) which is transformed to pixel representation and used in the simulator
- http://robot-simulator.herokuapp.com
- 1 metre in real life is represented by 1000 pixels in the simulator
 - > 1mm precision



Simulator

- adjustable parameters:
 - > track size in real world
 - starting position
 - interval of sensor detection
 - starting and max speed
 - > Arduino servo specifics (values range, which servo signal means stopping)
 - robot's initial rotation
 - wheel gauge (distance of the centers of the wheel)
 - sensors' positions in regard to the center of the wheel axis
 - sensor radius
 - controller of the robot (a neural net in our case)
- metres and seconds used as units

Robot controlling system

- neural net with 5 input neurons and 2 output neurons
 - however, the simulator can theoretically work with any controller capable of mapping a 5-dimensional input vector to a 2-dimensional output vector
- we used digital (=binary) inputs from the sensors (0/1 for black/white)
- outputs are expected in the interval (-1,1) and are subsequently mapped to (1300, 1700) by the simulator

Evolution

- **NEAT** used to evolve networks
 - JS library neataptic (<u>https://github.com/wagenaartje/neataptic</u>)
- starting with a random 2-layer network with 3 hidden neurons and sigmoid activation function on all nodes

Configurations

- > population: 100-200
- ➤ max generation: 200-500
- ➤ mutation rate: 0.8
- > experimented with various mutation strategies and fitness

Mutations

- possible mutations:
 - > modify weight of a synaptic connection
 - modify threshold of a neuron
 - > modify the neuron's activation function
 - > add a connection
 - > remove a connection
 - > add a neuron
 - remove a neuron

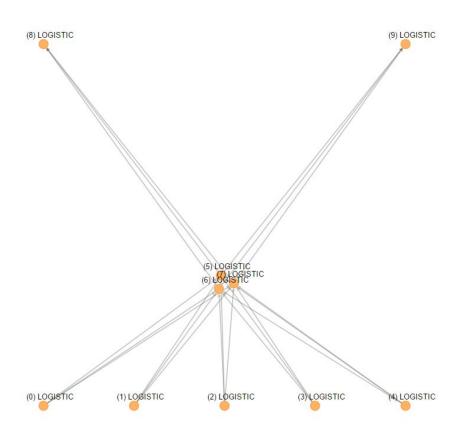
Results

First steps

- mutation: only change of synaptic weights and neural thresholds allowed (no structural mutations)
- fitness: travelled distance
- stopping condition: robot is out of track (all sensors see white)

Result:

Failure. Can't even make the first turn.

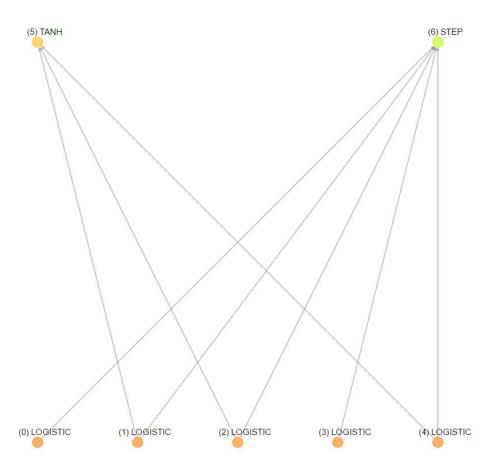


Partial success

- mutation: all allowed, but:
 - input neurons fixed on LOGISTIC (=sigmoid activation)
- fitness: travelled distance & middle sensor on the line
- stopping condition: robot is out of track (all sensors see white) <u>for a</u> <u>certain period of time</u>

Result:

Can navigate about a quarter of the track before failing.



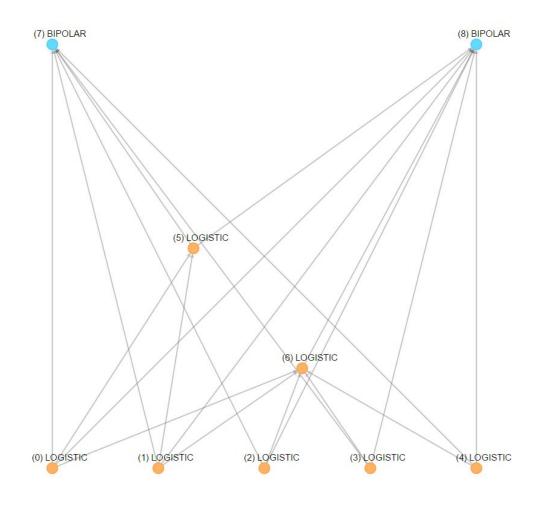
Video

Cheating robot

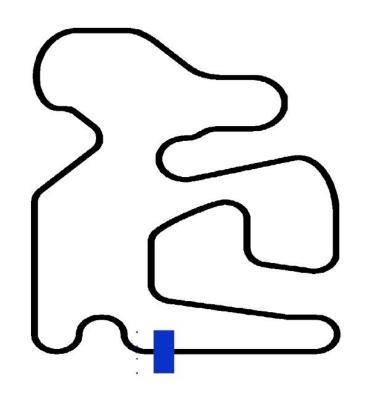
- mutation: all allowed, but:
 - input neurons fixed on LOGISTIC (=sigmoid activation)
- fitness: travelled distance & middle sensor on the line & speed
- stopping condition: robot is out of track (all sensors see white) for a certain period of time

Result:

Well...



Video



Looks like an image instead of a video?

> Object {left: 0.2, right: 0.2} > Object {left: -0.2, right: -0.2} > Object {left: 0.2, right: 0.2} > Object {left: -0.2, right: -0.2} > Object {left: 0.2, right: 0.2} > Object {left: -0.2, right: -0.2} > Object {left: 0.2, right: -0.2} > Object {left: -0.2, right: 0.2}

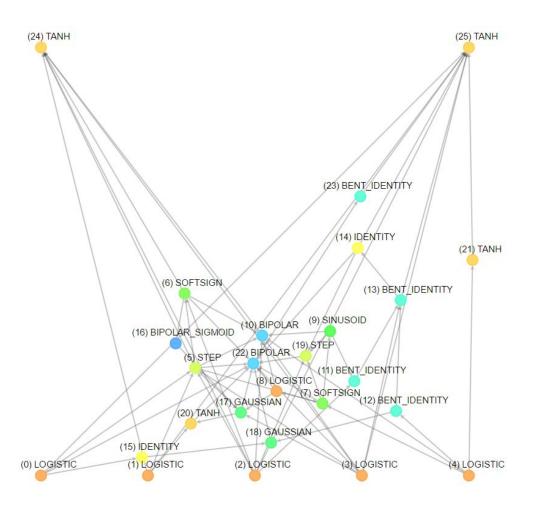
Every tick, the robot iterates between going full speed ahead and full speed in reverse.

The "winner"

- mutation: all allowed, but:
 - input neurons fixed on LOGISTIC (=sigmoid activation)
 - > output neurons fixed on TANH
- fitness: travelled distance & middle sensor on the line & speed
 - big penalization for going backwards
- stopping condition: robot is out of track (all sensors see white) for a certain period of time

Result:

Flawless on both the training track and a previously unseen, more complicated track.



Video



Is such a large network necessary?

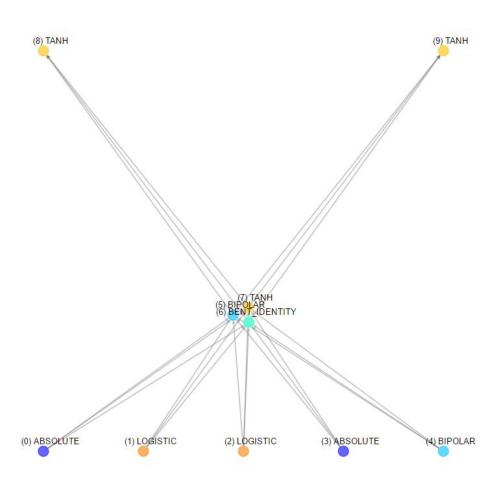
- penalization of large network in fitness function
 - > always degraded to networks with no hidden neurons
- fix the structure again, but use better fitness and stopping condition

Final experiment

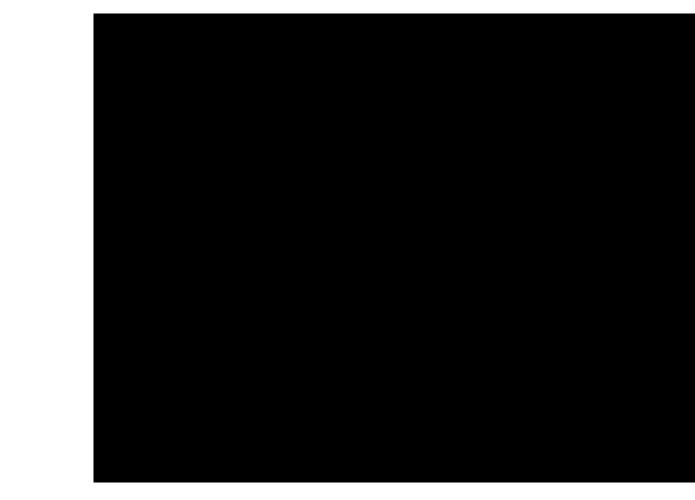
- mutation: only non-structural, but
 - output neurons fixed on TANH
 - > input neurons are not fixed anymore
- fitness: travelled distance & middle sensor on the line & speed
 - ➢ big penalization for going backwards
- stopping condition: robot is out of track (all sensors see white) for a certain period of time

Result:

Flawless on the training track, but problems on the previously unseen track.



Video



Intermezzo - what could be done better?

- simulator
 - > more parameters
 - friction, acceleration
- neuroevolution
 - better fitness function
 - > penalization for large networks
 - > train the neural net on multiple tracks at once
 - > or even better, dynamically generate path and train the neural net on it

Arduino implementation

- Uses C
 - > How to implement the neural net?
- Using Neataptic.js, we can get the list of nodes and edges and then do the computation ourselves
 - https://pastebin.com/xGHupN7C
- Error prone, takes too long
- Can it be done better?
 - ≻ YES!
 - ➤ <u>https://pastebin.com/7i2zXzaM</u>

Real world results

- ✤ We have the neural net implemented in C
- Simulator works
 - Fairy tale environments where nothing can go wrong
 No outer influences
 - No outer influences
 - No friction
 - Instant acceleration
- Would the robot controlling neural net work just by porting it to the real world robot?

Real world results

✤ Yes.



Thank you for your attention. Questions?

