

# Neuroevolution

Elliot Greenlee and Jared M. Smith  
CS581 Algorithms, April 20, 2017



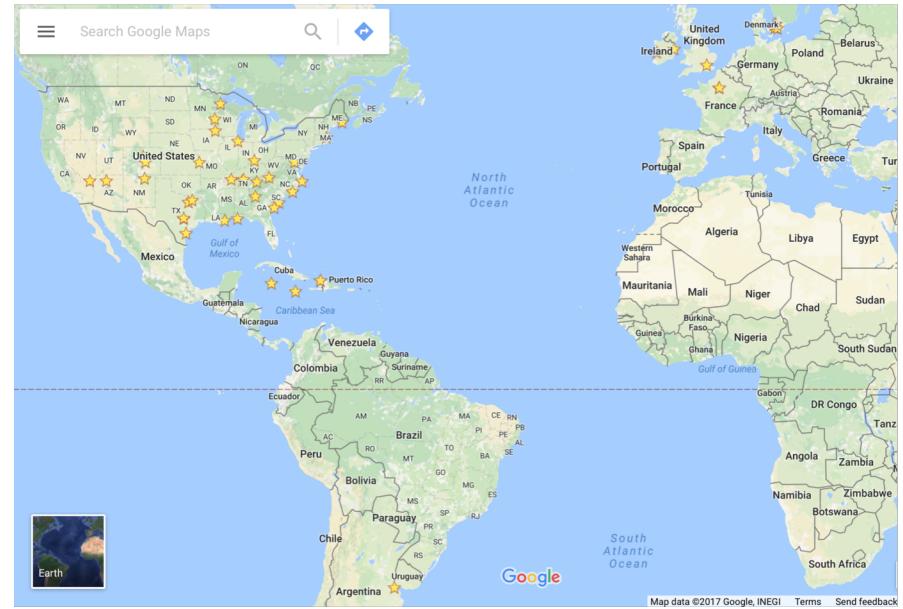
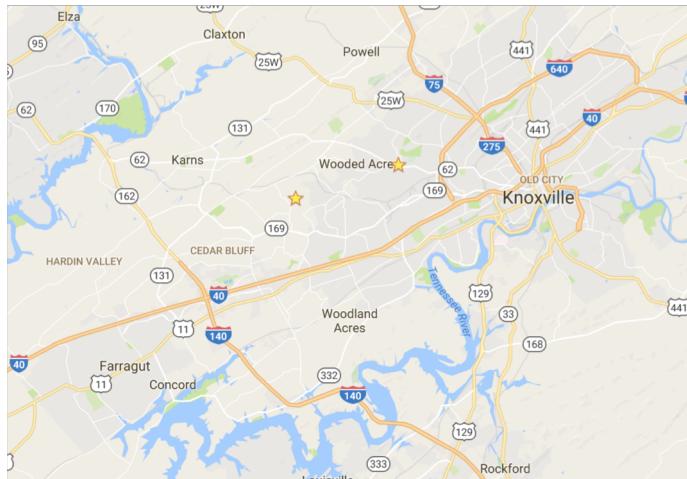
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# Test Questions

1. What is the typical training method for neural networks?
1. What three Darwinian processes are used in genetic algorithms?
1. What is one application of neuroevolution?

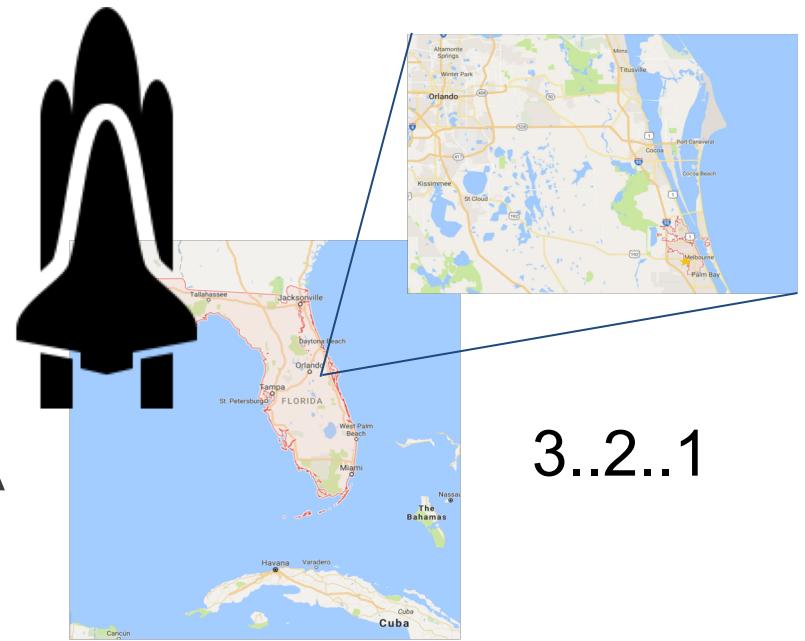
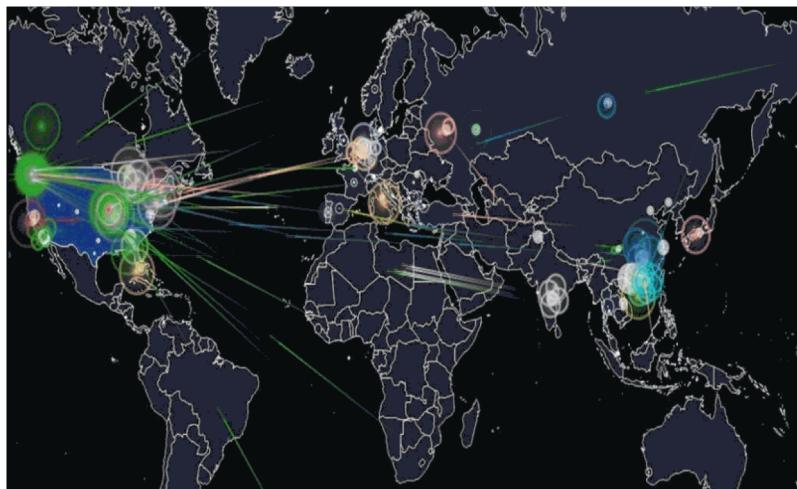
# About Elliot

- Knoxville, Tennessee
- MS in Computer Science
- PhD in the future
- Rock climbing, cooking, travel
- Security research at ORNL
- Machine learning and computer vision GRA with Dr. Hairong Qi



# About Jared

- Melbourne, Florida
- HackUTK and VolHacks founder
- Research Staff and PI at ORNL
- Computer security and networking GRA with Dr. Max Schuchard
- Hiker, backpacker, industry conference teacher (7 so far)



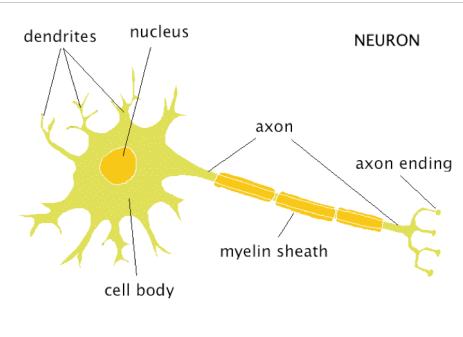
3..2..1



# Outline

- Overview
- History
- Algorithms
- Applications
- Implementations
- Open Issues
- Discussion

# Overview



Human brain



Neural networks



Self-improving neural networks

Tall or Short	Purple or White	Yellow or Green	Smooth or Wrinkled	Yellow or Green	Smooth or Wrinkled

Darwinian processes



Genetic algorithms



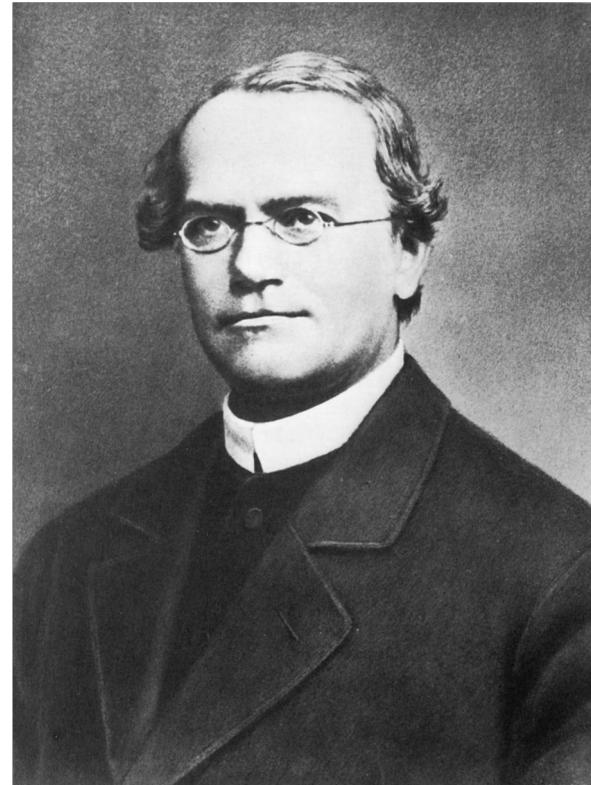
# Overview

- Applications include: robotics, video game AI's, and personal assistants



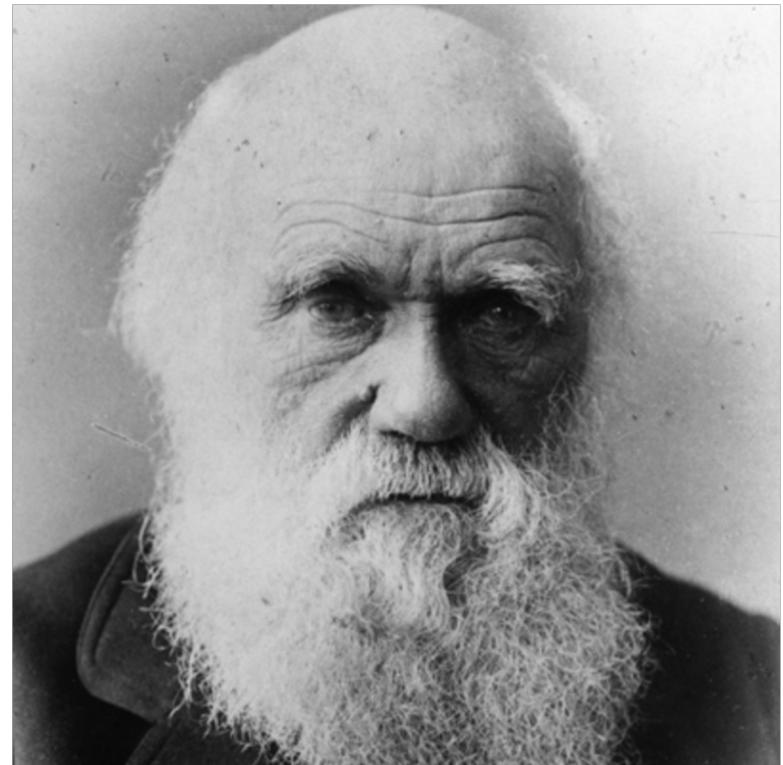
# History

- Gregor Mendel
  - Pioneer in genetics
  - Discovered inheritance
  - Used peas in his experiments
  - “Mendel’s Peas”



# History

- Charles Darwin
  - “Father of Evolution”
  - Natural selection
  - Studied ancestral relationships in the Gálapagos Islands



# History

- Marvin Minsky
  - Former Professor at MIT
  - “Father of Artificial Intelligence”
  - Came up with perceptrons, the foundation of artificial neural networks



# History

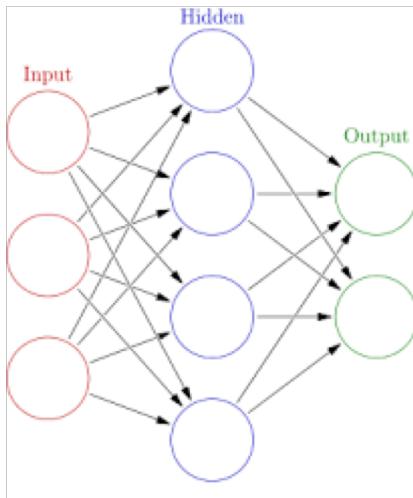
- Kenneth O. Stanley
  - Associate Professor at U. of Central Florida
  - “NeuroEvolution of Augmenting Topologies”
  - Developed original NEAT software in C++
  - Many other implementations



# Algorithms

## Neural networks

- layers
- weights
- activation function
- backpropagation



## Genetic Algorithms

- fitness
- crossover
- mutation
- selection

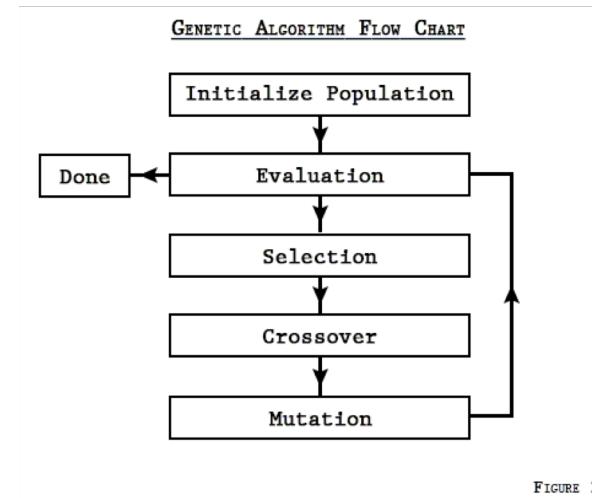


FIGURE 2

# Algorithms - Genome

## Node

- bias
  - response
  - activation
  - aggregation
- 
- distance()
  - mutate()
  - crossover()

## Connections

- weight
  - enabled
- 
- distance()
  - mutate()
  - crossover()

# Algorithms - Genetic Processes

- Initialize
- Copy
- Distance
  - node/connection structural similarity
  - individual gene distance metrics

## Mutate

- add node
- delete node
- add connection
- delete connection
- change connection genes
- change node genes

# Algorithms - Neural Networks

Create

1. For each layer:

    For each node in layer:

        For each connection:

Feedforward

1. For each node:

    For each forward link:

        Calculate

# Algorithms

For every generation:

1. Evaluate genome fitness
2. Stop if fitness threshold is reached
3. Do reproduction
4. Check for extinction
5. Speciate

# Algorithms - Genetic Processes

## Reproduction

1. Remove stagnated species
2. Compute new spawns
3. Find best from prior generation
4. Select two parents randomly
5. Cross parents
6. Mutate child

# Algorithms - Genetic Processes

## Crossover

1. For all similar connections, crossover
2. Add connections from more fit parent
3. For all similar nodes, crossover
4. Add nodes from more fit parent

# Algorithms - Genetic Processes

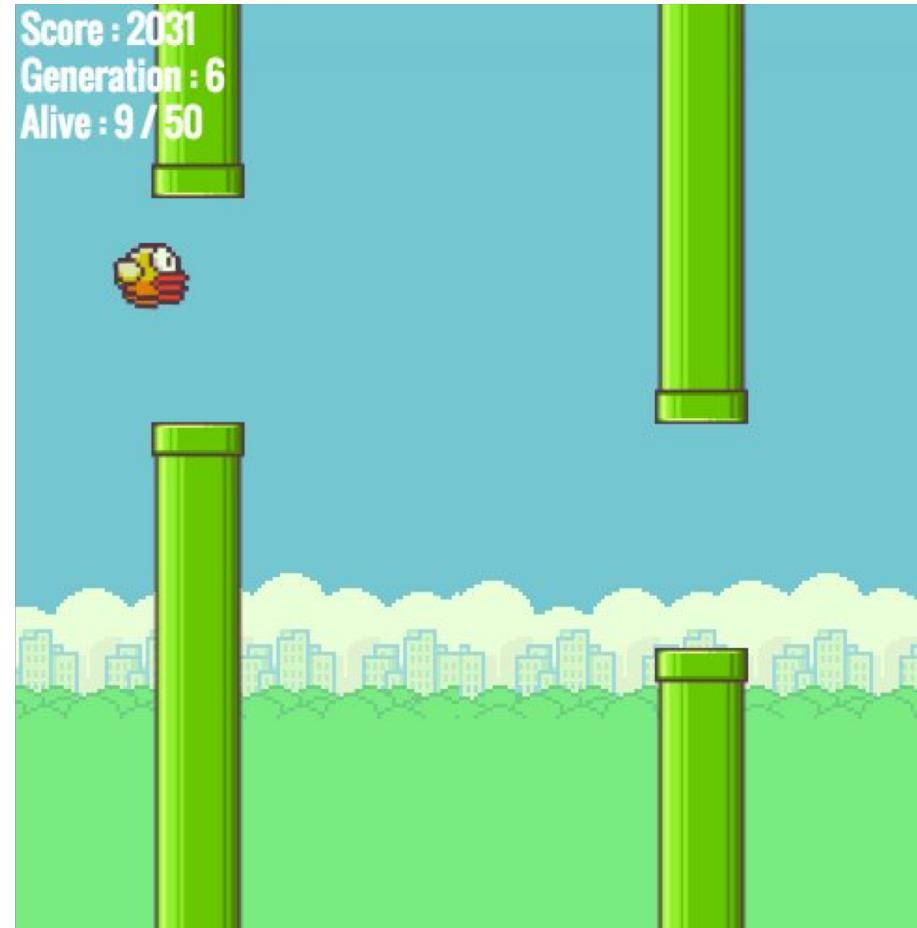
## Speciation

1. Compare all new samples to the prior species “representatives”
2. Find most similar new sample
3. Compare all new samples to the new species “representatives”
4. If similar enough, add. If different, new

## Stagnation

# Applications

- Robotics
- Artificial Life
- Video Games
- Substitute for reinforcement learning



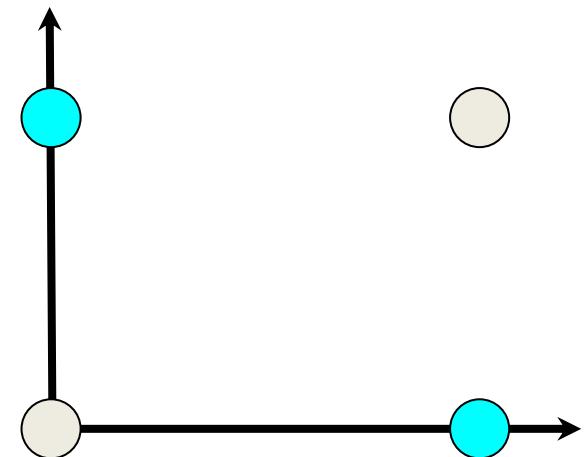
# Implementations

- XOR
- Single Pole Balancing
- Lunar Lander

# XOR

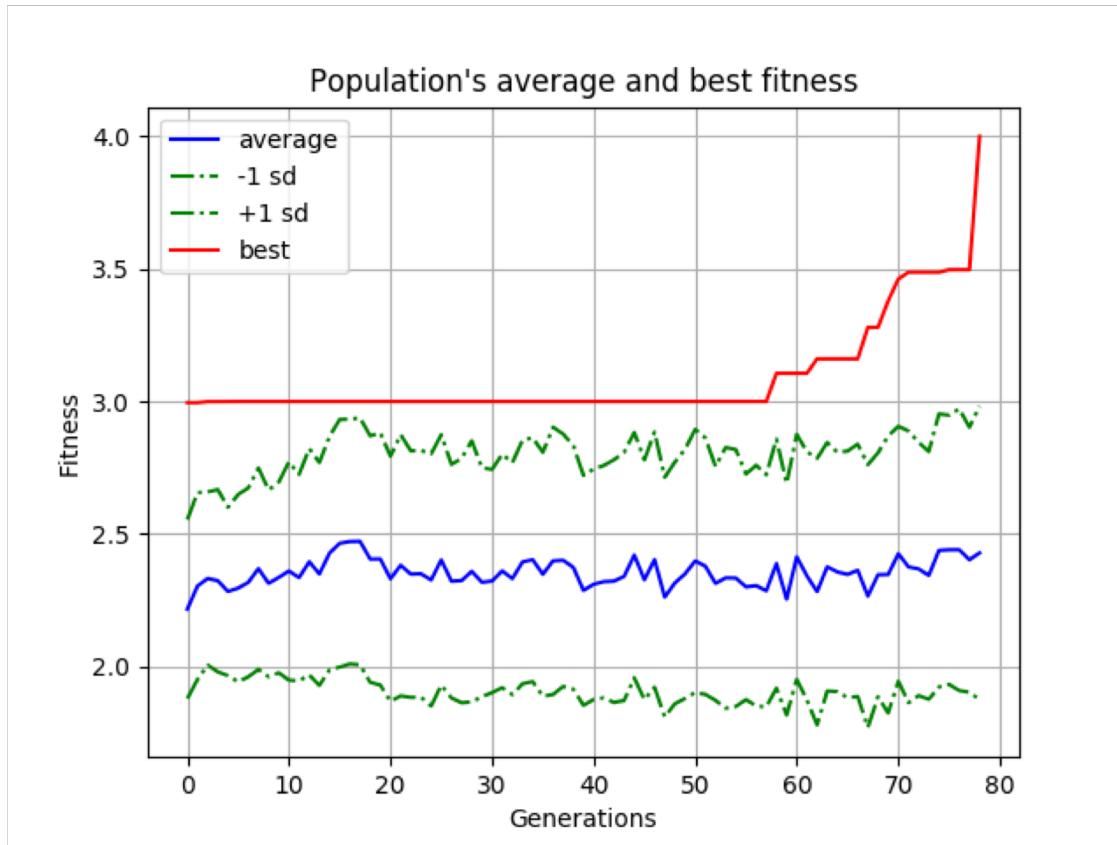
- Not linearly separable

Input 1	Input 2	Output
0	0	0
0	1	1
1	0	1
1	1	0



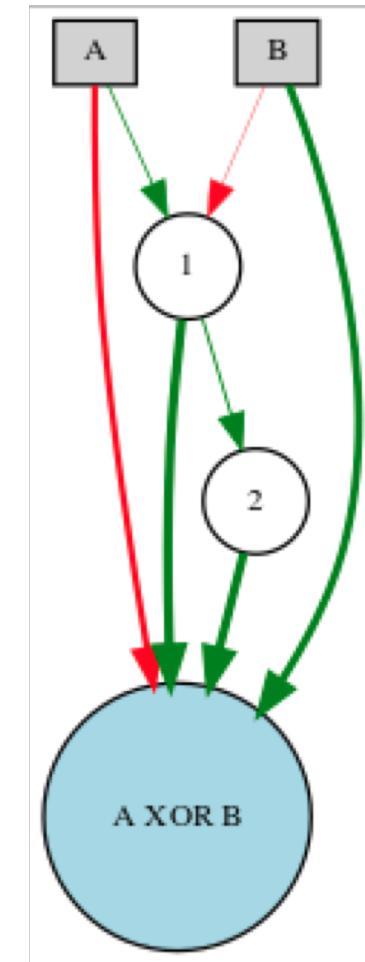
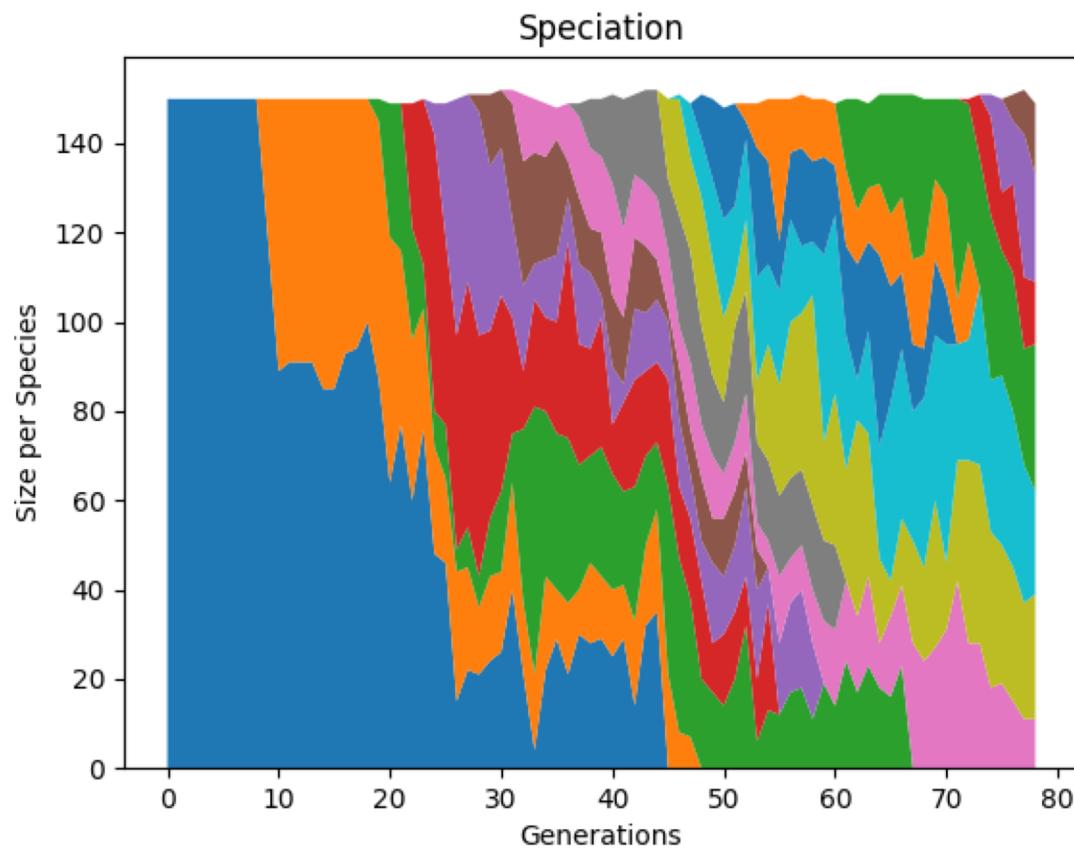
# XOR

- Average # of generations over 5 runs: **102.4**



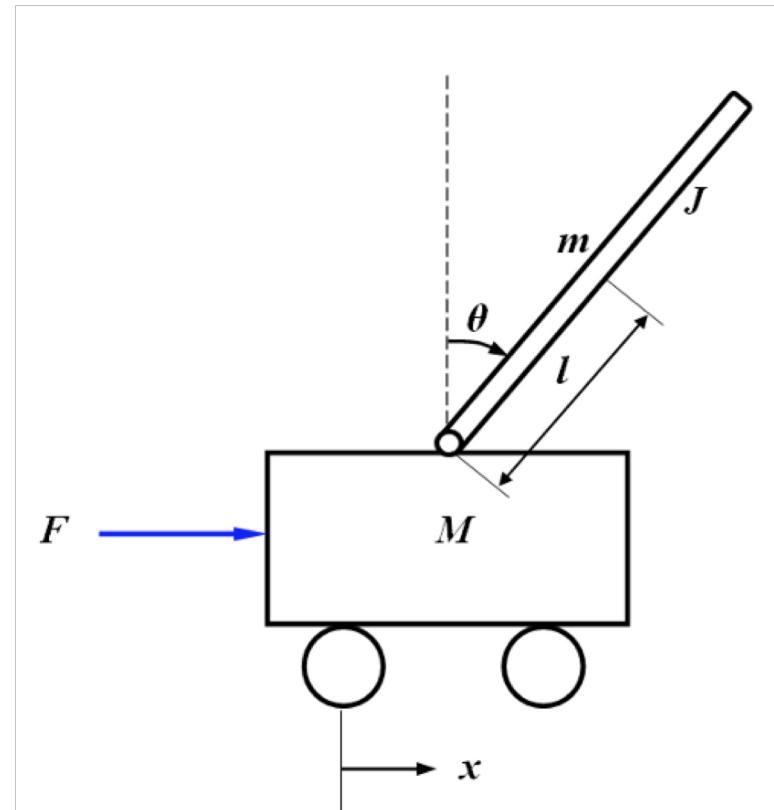
78  
Generation  
Example

# XOR



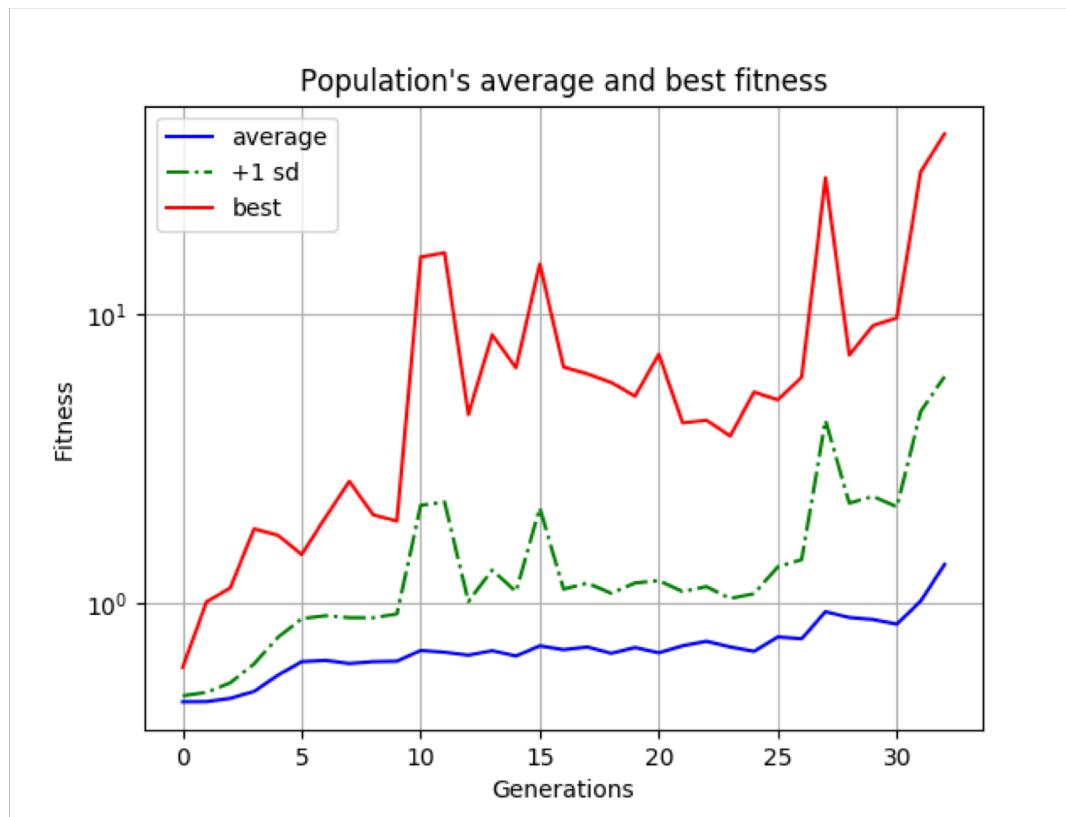
# Pole Balancing

- Balance a single pole on a cart
- Can move horizontally ( $x$ ) and a pole that can change angle w.r.t. the cart ( $\Theta$ )



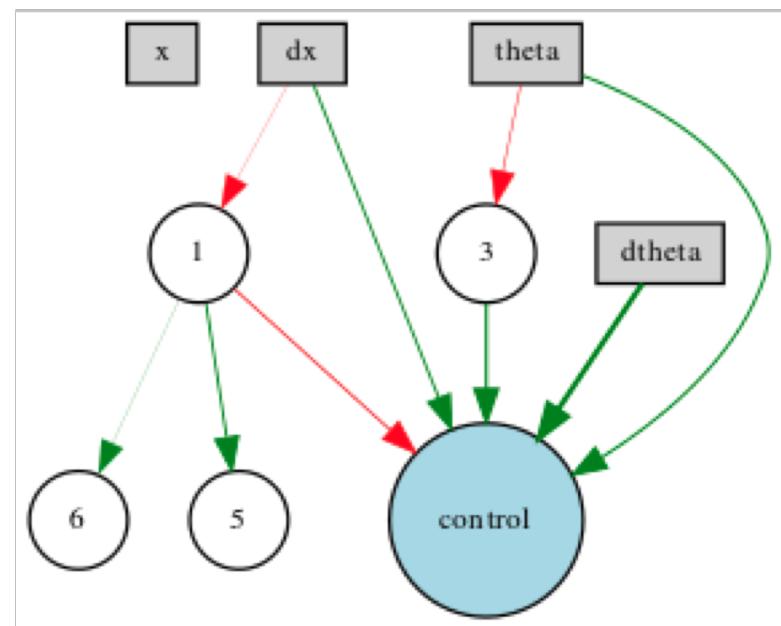
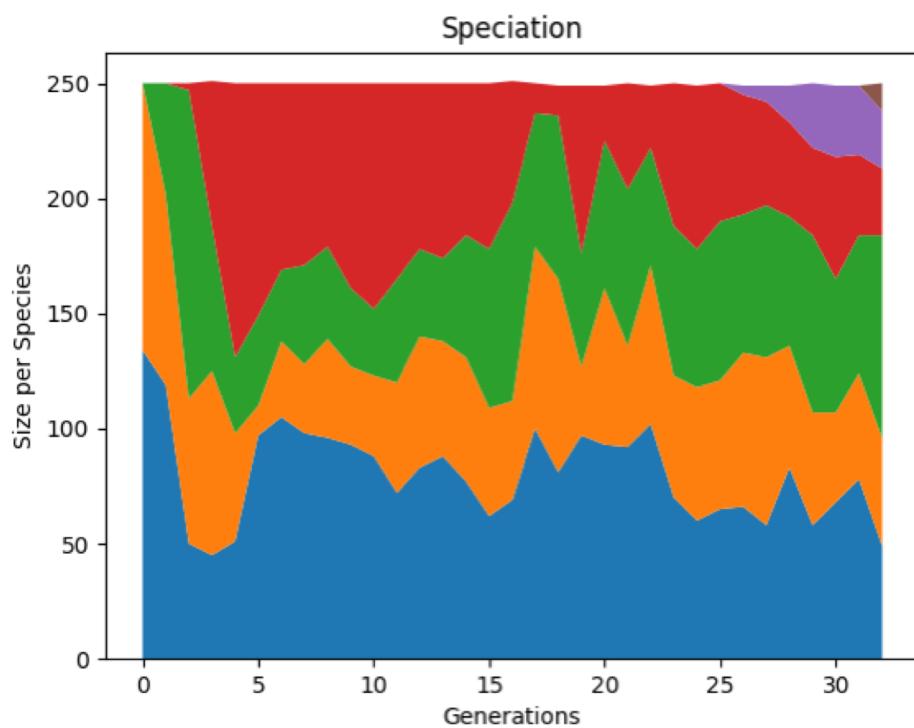
# Pole Balancing

- Average # of generations over 5 runs: **28**



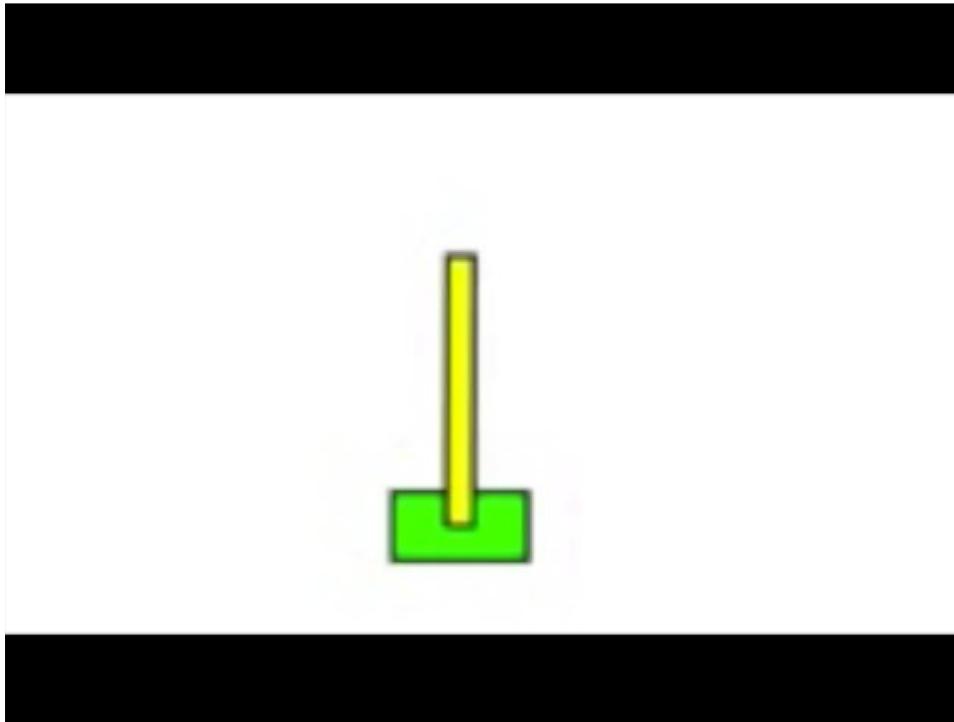
32  
Generation  
Example

# Pole Balancing



# Pole Balancing

0, 5, 10, 15, 18 generation videos



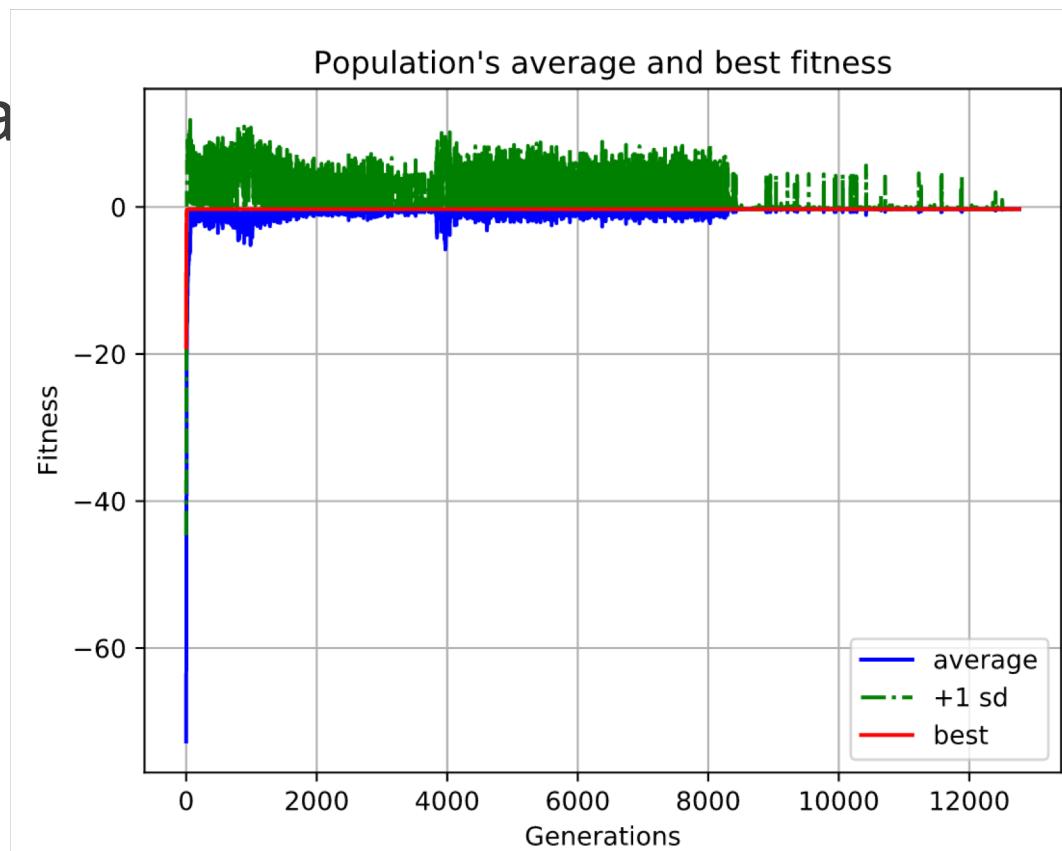
# Lunar Lander

- Game in which the goal is to land a spacecraft on flat ground
- Spacecraft has the ability to shoot particles to move the vehicle, which represents small rocket boosters that adjust movement



# Lunar Lander

- Limitations of algorithm
  - Initialization
  - Local Minima



# Lunar Lander

Demo!

# Open Issues

- Integrating with other types of learning
  - Q-Learning, TD-Learning
- Developing better evolutionary methods
  - Competitive and cooperative co-evolution
- Building “General AI”
  - Games, personal assistants, etc. can all benefit from better learning provided by neuroevolution and genetic programming

# References

- Kenneth O. Stanley and Risto Miikkulainen. 2002. Evolving neural networks through augmenting topologies. *Evol. Comput.* 10, 2 (June 2002), 99-127. DOI=<http://dx.doi.org/10.1162/106365602320169811>
- Michael O'Neill, Leonardo Vanneschi, Steven Gustafson, and Wolfgang Banzhaf. 2010. Open issues in genetic programming. *Genetic Programming and Evolvable Machines* 11, 3-4 (September 2010), 339-363. DOI=<http://dx.doi.org/10.1007/s10710-010-9113-2>
- NEAT-Python, <https://github.com/CodeReclaimers/neat-python>
- OpenAI Gym, <https://github.com/openai/gym>

# Discussion

- How does Neuroevolution compare to typical reinforcement learning?
- What future applications do you see Neuroevolution solving?
- Could you implement biologically-inspired neural networks on hardware? Would it help?

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