

# Robot Walking with Genetic Algorithms

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# Outline

- Introduction
- Genetic algorithms
- Quadruped Robot
- Hexapod Robot
- Biped Robot
- Evaluation

# Why walking?

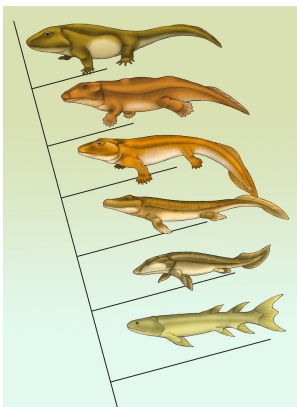
- Mobile robots
- Similar to humans and animals
- Can move forward on different surfaces
- Can climb

# Why walking?

- Mobile robots
- Similar to humans and animals
- Can move forward on different surfaces
- Can climb

Problem: Walking is complex

# Evolutionary Algorithms



Fishapod-Tetrapod-Tree-186821546

<http://www.deviantart.com/art/Fish->

- Biological inspired
- Survival of the fittest
- Reproduction and Mutation

# Genetic Algorithm

- Population based

# Genetic Algorithm

- Population based
- Solution as chromosomes

# Genetic Algorithm

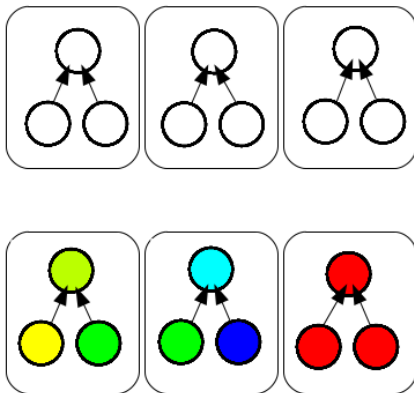
- Population based
- Solution as chromosomes
- Fitness function



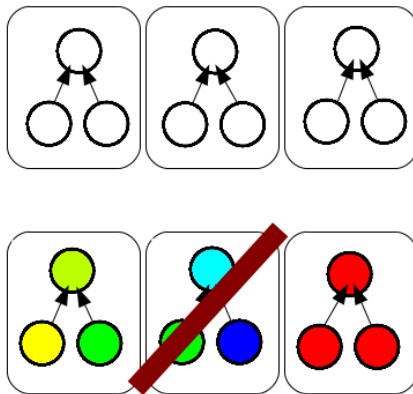
# Genetic Algorithm

- Population based
- Solution as chromosomes
- Fitness function
- Generating a new generation

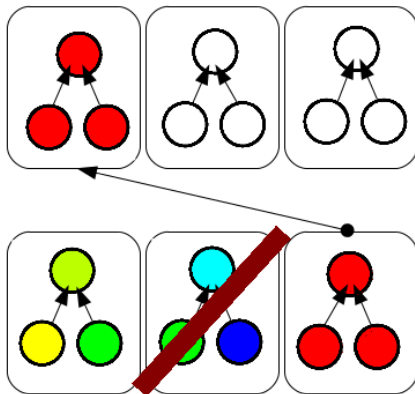
# Example: Finding Orange



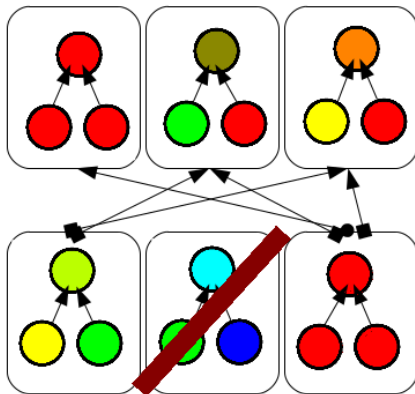
# Example: Finding Orange



# Example: Finding Orange



# Example: Finding Orange



# Quadruped Robot



<http://www.roboticstoday.com/robots/aibo-ers-7>

- 4 legs
- 2 DOF per leg

# Setup

- Code for the walking exists

# Setup

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- 235 mm/s with human testing



# Setup

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- 235 mm/s with human testing
- 4 Robots to test parallel and autonomous

# Setup

- Code for the walking exists
- 235 mm/s with human testing
- 4 Robots to test parallel and autonomous
- Some hours to learn

# Genetic Algorithm

- Population: 30
- Chromosomes: Parameters of the walking algorithm
- Fitness function: Distance moved forward

# Result

- Distance measurement sometimes failed

# Result

- Distance measurement sometimes failed
- Time was enough

# Result

- Distance measurement sometimes failed
- Time was enough
- 290 mm/s with GA optimization

# Result

- Distance measurement sometimes failed
- Time was enough
- 290 mm/s with GA optimization
- The robot got faster

# Hexapod Robot



<http://robot-kingdom.com/hexapod-robot-tutorials-where-everything-begins/>

- 6 legs
- 2 DOF per leg



# Setup

- Evolution of a neural network for walking control

# Setup

- Evolution of a neural network for walking control
- Separated in two steps

# Setup

- Evolution of a neural network for walking control
- Separated in two steps
- Step one: move a leg

# Setup

- Evolution of a neural network for walking control
- Separated in two steps
- Step one: move a leg
- Step two: walking with 6 legs

# Genetic Algorithm 1

- Chromosomes: Neural Network for one leg

# Genetic Algorithm 1

- Chromosomes: Neural Network for one leg
- Fitness function: Range of the oscillation

# Genetic Algorithm 1

- Chromosomes: Neural Network for one leg
- Fitness function: Range of the oscillation
- Stops when oscillation reaches the full length

# Genetic Algorithm 2

- Population: 10; 2 stay



# Genetic Algorithm 2

- Population: 10; 2 stay
- Chromosomes: Neural network for all legs

# Genetic Algorithm 2

- Population: 10; 2 stay
- Chromosomes: Neural network for all legs
- Fitness function: distance from starting point in the body axis

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- Population: 10; 2 stay
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- Crossover rate 0.1

# Genetic Algorithm 2

- Population: 10; 2 stay
- Chromosomes: Neural network for all legs
- Fitness function: distance from starting point in the body axis
- Crossover rate 0.1
- Mutation rate: 0.04

# Result

- Learned to move their legs

# Result

- Learned to move their legs
- Robots walked backwards

# Result

- Learned to move their legs
- Robots walked backwards
- Similar to living insects

# Result

- Learned to move their legs
- Robots walked backwards
- Similar to living insects
- Wave-gait vs tripod gait



# Biped Robot



- 2 legs
- 6 DOF per leg

<http://www.engadget.com/2006/11/28/kondo-adds-pivot-to-khr-1hv-biped-robot-kit/>

# Experiment

- Truncated Fourier Series and polynomial equations used for motor control

Joint name	Support Leg (Left Leg)	Swing leg (Right leg)
Left Hip Yaw	Sinusoid	Sinusoid
Left Hip Roll	Sinusoid	Sinusoid
Left Hip Pitch	Polynomial	Sinusoid
Left Knee Pitch	Polynomial	Sinusoid
Left Ankle Pitch	Polynomial	Sinusoid
Left Ankle Roll	Sinusoid	Sinusoid

Evaluating GA and PSO Evolutionary Algorithms for Humanoid Walk Pattern Planning

# Experiment

- Truncated Fourier Series and polynomial equations used for motor control
- Sinusoid equation the swing leg

Joint name	Support Leg (Left Leg)	Swing leg (Right leg)
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# Experiment

- Truncated Fourier Series and polynomial equations used for motor control
- Sinusoid equation the swing leg
- partly polynomial equations for the support leg

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Evaluating GA and PSO Evolutionary Algorithms for Humanoid Walk Pattern Planning

# Experiment

- Truncated Fourier Series and polynomial equations used for motor control
- Sinusoid equation the swing leg
- partly polynomial equations for the support leg
- Learning of the equations for every motor

Joint name	Support Leg (Left Leg)	Swing leg (Right leg)
Left Hip Yaw	Sinusoid	Sinusoid
Left Hip Roll	Sinusoid	Sinusoid
Left Hip Pitch	Polynomial	Sinusoid
Left Knee Pitch	Polynomial	Sinusoid
Left Ankle Pitch	Polynomial	Sinusoid
Left Ankle Roll	Sinusoid	Sinusoid

Evaluating GA and PSO Evolutionary Algorithms for Humanoid Walk Pattern Planning

# Genetic Algorithm

- Population: 20
- Chromosomes: Coefficients of the equations
- Fitness function: Walking without falling

<b>Model of implementation</b>	<b>Selection</b>	<b>Crossover</b>	<b>Mutation</b>
GA-1	Random	Random	10%
GA-2	Roulette wheel	Random	10%
GA-3	Roulette wheel	All pitches and all rolls	10%

Evaluating GA and PSO Evolutionary Algorithms for Humanoid Walk Pattern Planning

# Result

- Coefficients are learned
- Each motor has specific equation
- robots learned to walk

Algorithm	#chromosome	#Generation	Best Fitness (average 5 iteration)	Iterations to Converge
GA-1	20	20	0.51376	-
<b>GA-2</b>	<b>20</b>	<b>20</b>	<b>0.58862</b>	<b>20</b>
GA-3	20	20	0.50434	-

Evaluating GA and PSO Evolutionary Algorithms for Humanoid Walk Pattern Planning

# Evaluation

- Optimization algorithm



# Evaluation

- Optimization algorithm
- Finds a good solution

# Evaluation

- Optimization algorithm
- Finds a good solution
- PSO (particle swarm optimization) usually faster

Thank you for your attention!

# References

- Chernova, Sonia, and Manuela Veloso. "An evolutionary approach to gait learning for four-legged robots." Intelligent Robots and Systems, 2004.(IROS 2004). Proceedings. 2004 IEEE/RSJ International Conference on. Vol. 3. IEEE, 2004.
- Lewis, M. Anthony, Andrew H. Fagg, and Alan Solidum. "Genetic programming approach to the construction of a neural network for control of a walking robot." Robotics and Automation, 1992. Proceedings., 1992 IEEE International Conference on. IEEE, 1992.
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