



# **Evolutionary Gait-Optimization Using a Fitness Function Based on Proprioception**

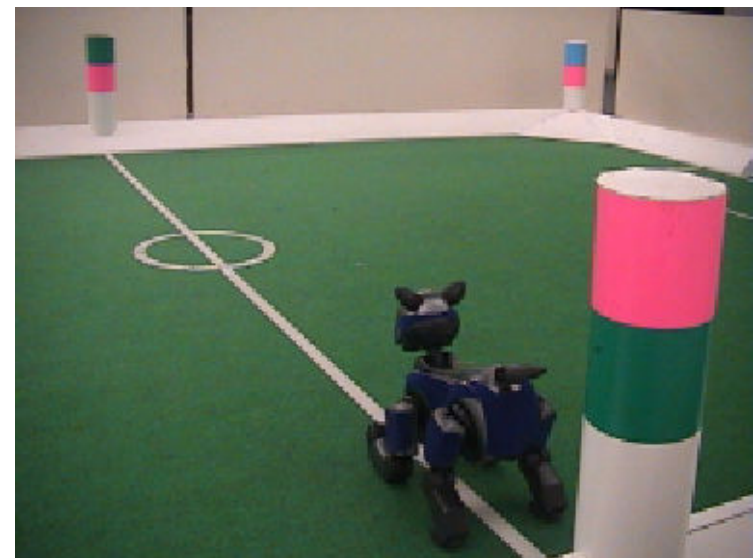
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## Introduction

- ⚽ Motivation
  - ⚽ Fast motion is important in RoboCup
  - ⚽ Good gait parameter sets are hard to find manually
  - ⚽ Therefore, robots should learn them automatically
- ⚽ Demands
  - ⚽ Easy to setup (maybe at RoboCup venue)
    - ⚽ *No ceiling camera*
  - ⚽ "Beacon to beacon" is limited
  - ⚽ Typical localization methods cannot be used
    - ⚽ *forward odometry not reliable under learning conditions*
    - ⚽ *re-localization required → slow*
- ⚽ Proprioception
  - ⚽ Real odometry
  - ⚽ Acceleration sensors



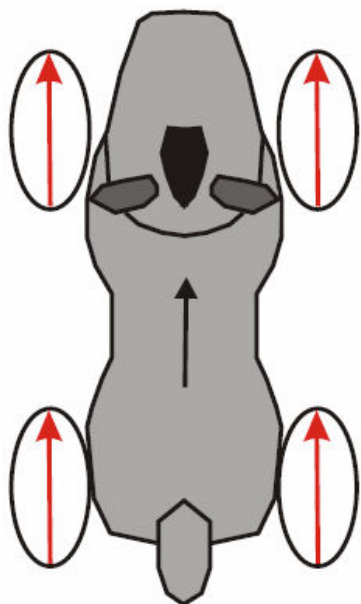
Kim and Uther, 2003



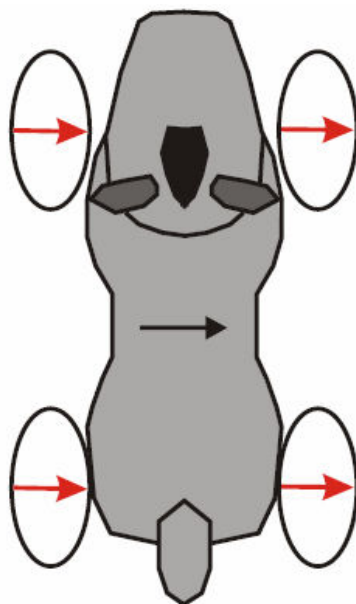
## Related Work

- G. Hornby, M. Fujita, S. Takamura, T. Yamamoto, and O. Hanagata. **Autonomous evolution of gaits with the Sony quadruped robot**. In *Proceedings of the Genetic and Evolutionary Computation Conference*, volume 2, Orlando, Florida, USA, 1999.
- G. Hornby, S. Takamura J. Yokono, O. Hanagata, T. Yamamoto, and M. Fujita. **Evolving robust gaits with Aibo**. In *IEEE International Conference on Robotics and Automation 2000 (ICRA-2000)*, 2000.
- M. S. Kim and W. Uther. **Automatic gait optimisation for quadruped robots**. In *Proceedings of the 2003 Australasian Conference on Robotics and Automation*, Brisbane, Australia, 2003.
- N. Kohl and P. Stone. **Policy gradient reinforcement learning for fast quadrupedal locomotion**. In *Proceedings of the IEEE International Conference on Robotics and Automation*, May 2004.
- M. J. Quinlan, S. K. Chalup, and R. H. Middleton. **Techniques for improving Vision and locomotion on the Sony Aibo robot**. In *Proceedings of the 2003 Australasian Conference on Robotics and Automation*, Brisbane, Australia, 2003.

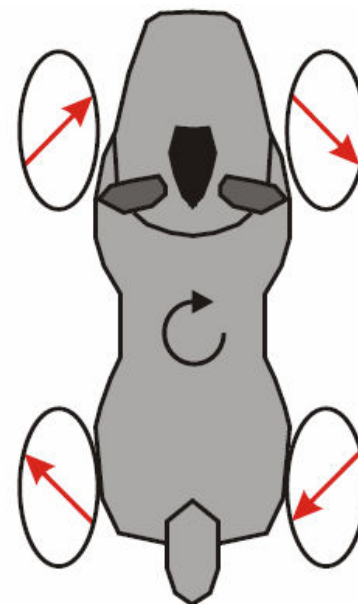
## Omni-Directional Walking



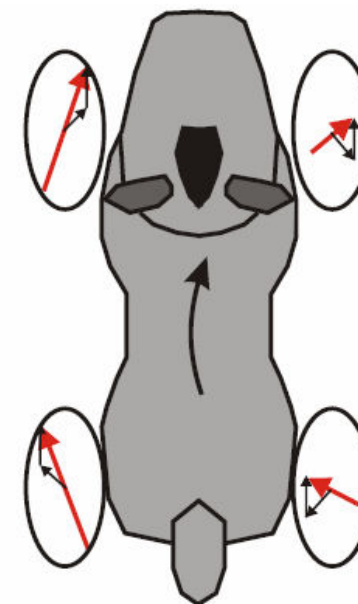
Forward



Sideward



Turning



Omni-directional

$$x^{left} = -x^r + r\theta^r$$

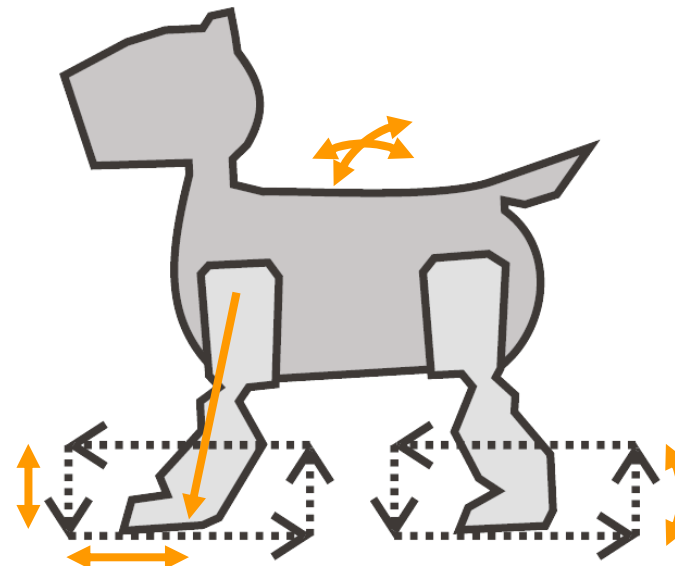
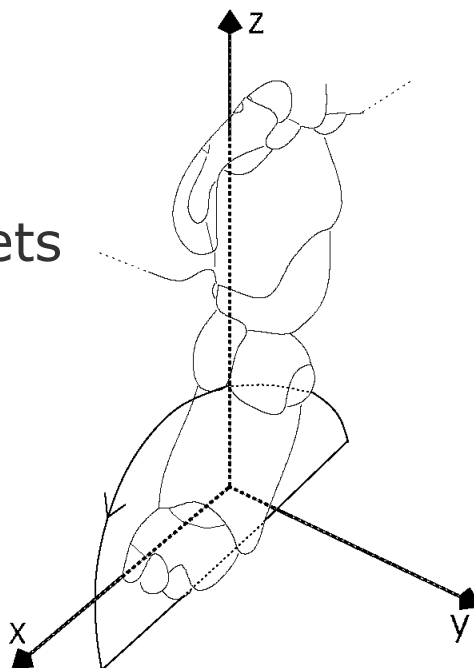
$$y^{left} = -y^r + r\theta^r$$

$$x^{right} = -x^r - r\theta^r$$

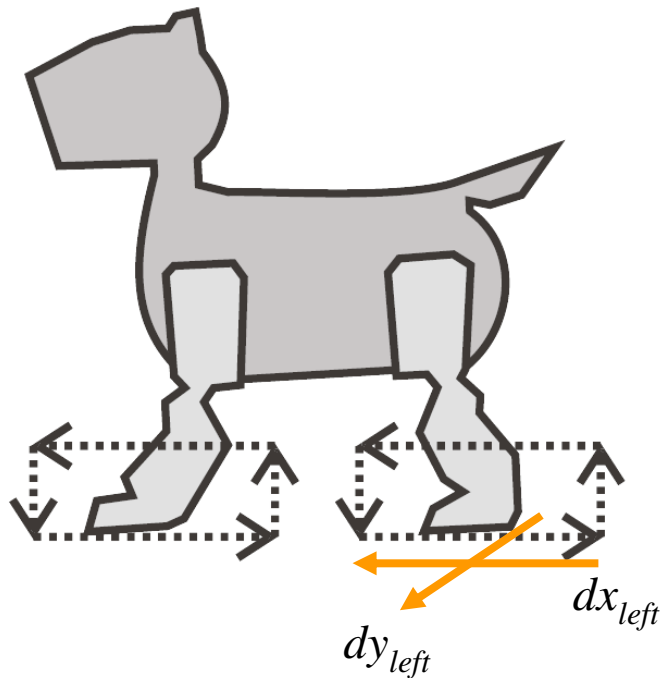
$$y^{right} = -y^r + r\theta^r$$

## Gait Parameters

- ⊙ Front/Rear locus
  - ⊙ x, y, z offsets
  - ⊙ step height
  - ⊙ tilt
  - ⊙ ground, lift, air, and lowering phases
- ⊙ Step
  - ⊙ size
  - ⊙ duration
- ⊙ Rear to Front offsets
  - ⊙ x speed ratio
  - ⊙ phase shift
- ⊙ Body shift
  - ⊙ x and y ratios
  - ⊙ phase offset



# Odometry



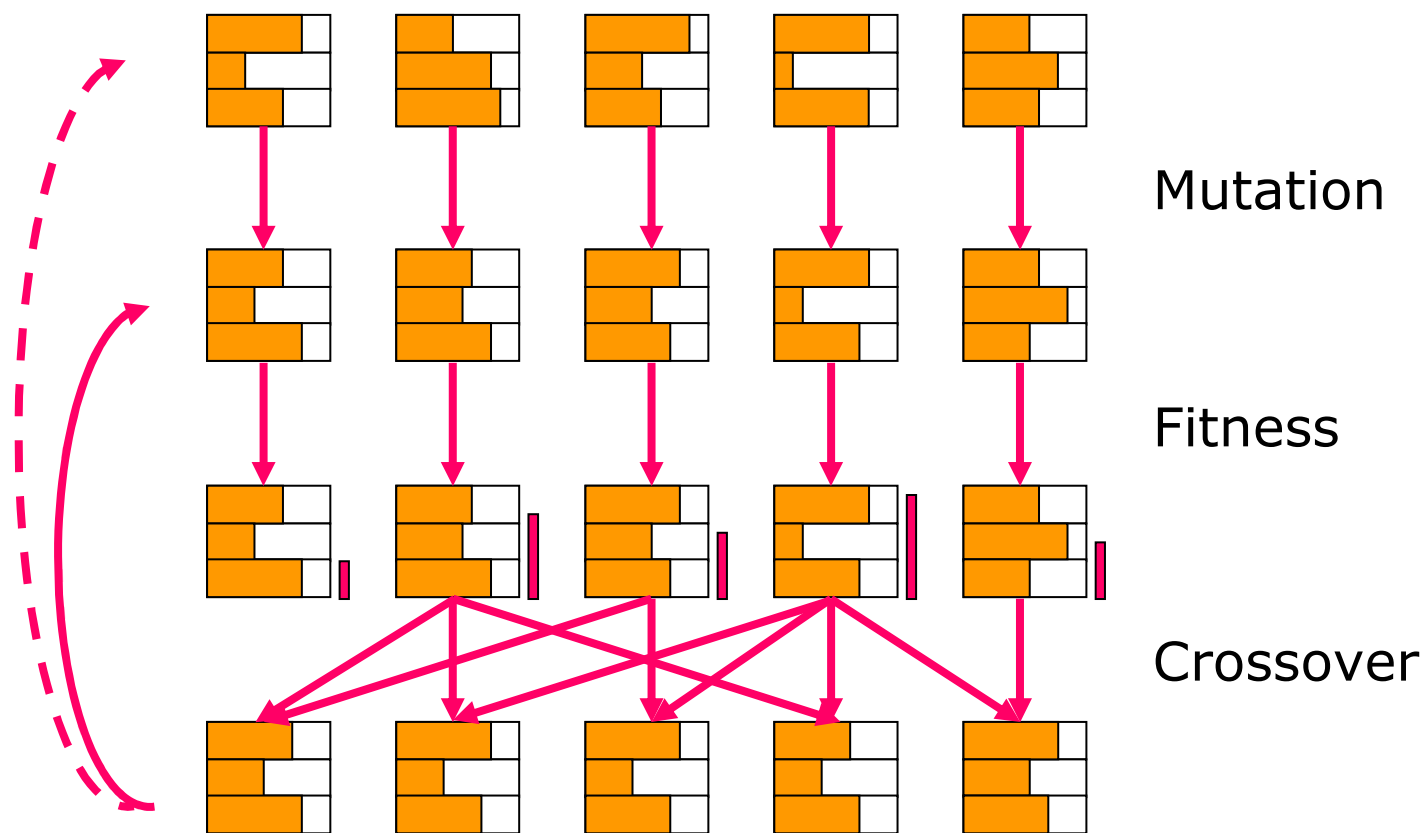
$$x_{t_a, t_b}^m = -\frac{dx_{t_a, t_b}^{right} + dx_{t_a, t_b}^{left}}{t_b - t_a}$$

$$y_{t_a, t_b}^m = -\frac{dy_{t_a, t_b}^{right} + dy_{t_a, t_b}^{left} + dx_{t_a, t_b}^{right} - dx_{t_a, t_b}^{left}}{t_b - t_a}$$

$$\theta_{t_a, t_b}^m = -\frac{dx_{t_a, t_b}^{right} - dx_{t_a, t_b}^{left}}{r(t_b - t_a)}$$



# Evolutionary Algorithm





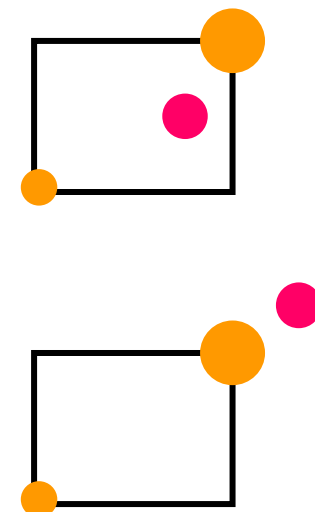
## Fitness Function

- ⚽ Procedure
  - ⚽ Gait parameter set of an individual is selected
  - ⚽ Robot walks for 5 seconds
  - ⚽ The first two seconds are ignored
- ⚽ Fitness is a weighted combination of
  - ⚽ the deviation between requested motion and measured motion,
  - ⚽ the standard deviation of the measurements of the three acceleration sensors in the robots body
- ⚽ Goal
  - ⚽ Move fast and smooth



## Crossover

- ⚽ Selecting individuals
  - ⚽ Each pair of individuals for crossover are drawn randomly based on their fitness
- ⚽ Combining individuals
  - ⚽ Interpolation
    - ⚽ *Random interpolation between parameters weighted by the fitness of the individuals*
  - ⚽ Extrapolation
    - ⚽ *Extrapolating the parameters of one individual away from the other*
    - ⚽ *Selection again randomly, weighted by the fitness*





## Experiments

- ⚽ Experiments 1 (ERS-210) and 2 (ERS-7)
  - ⚽ Robot is moving from goal to goal
  - ⚽ It is turned manually
  - ⚽ 5 seconds per individual (first 2 seconds ignored)
  - ⚽ Best individual is selected as basis for a new population after a restart
  - ⚽ Experiments of about 3 hours
- ⚽ Experiment 2 (ERS-7)
  - ⚽ Same as experiment 1, but robot is walking backward
- ⚽ Experiment 3 (ERS-7)
  - ⚽ Robot is following the ball
  - ⚽ Ball is moved manually
  - ⚽ Took about 4 hours





## Resulting Gaits – ERS-210



Forward: 311 mm/s

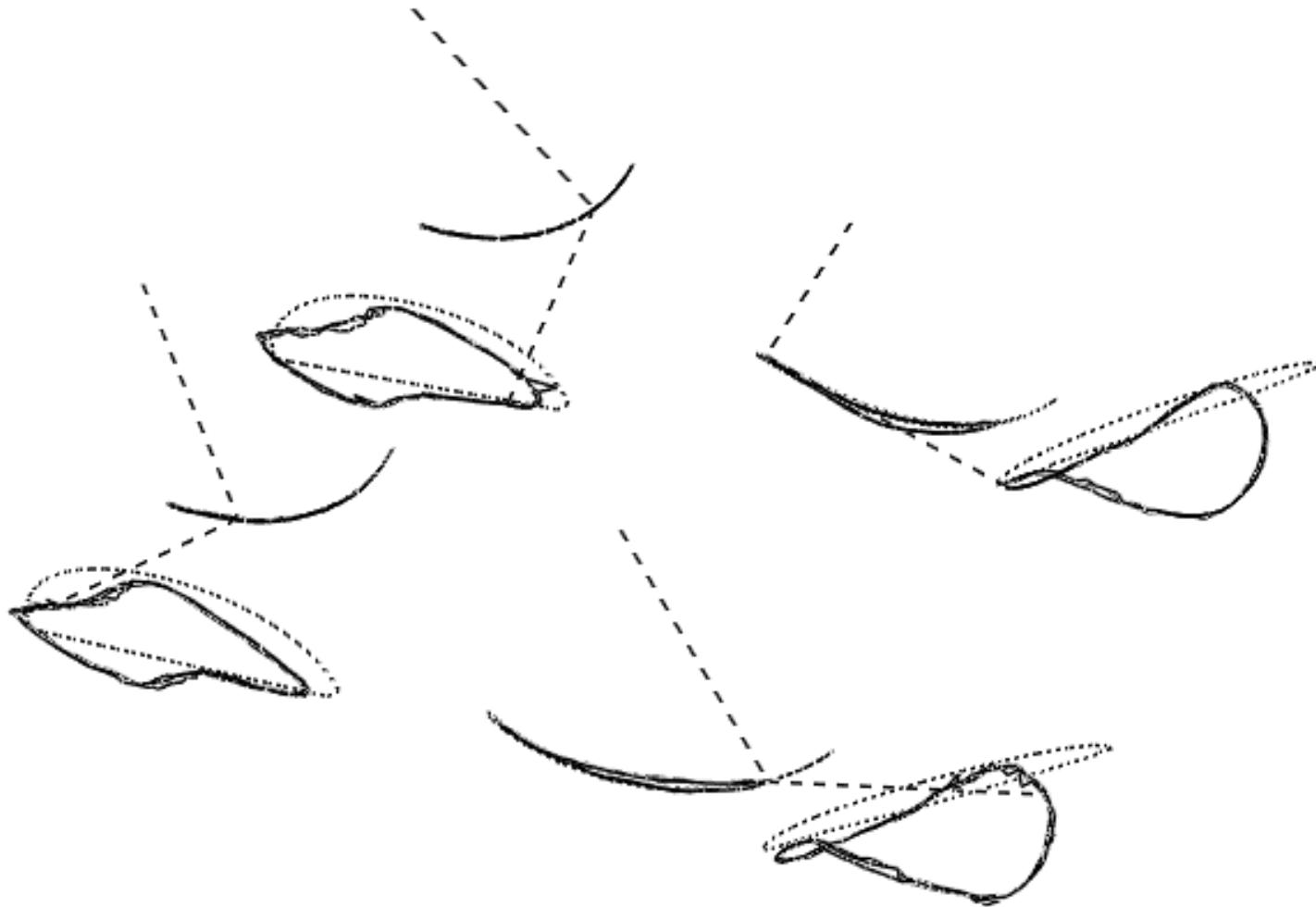


## Resulting Gait Parameters

Parameter	Initial Value	Best Value	Parameter	Initial Value	Best Value
Front locus			Rear locus		
z offset	76.85	58.97	z offset	108.72	111.54
y offset	78.1	67.24	y offset	76.95	58.09
x offset	55.88	63.26	x offset	-45.216	-35.96
step height	5.0	6.14	step height	24.0	27.27
tilt	-0.25	-0.48	tilt	0.05	0.12
ground phase	0.5	0.64	ground phase	0.5	0.34
lift phase	0.06	0.04	lift phase	0.06	0.19
lowering phase	0.06	0.12	lowering phase	0.06	-0.02
Step			Body shift		
size	76.0	89.77	x ratio	0	1.12
duration	0.64	0.512	y ratio	0	-0.10
Rear to Front offsets			phase offset	0	0.35
x speed ratio	1.1	1.0			
phase shift	0	-0.012			

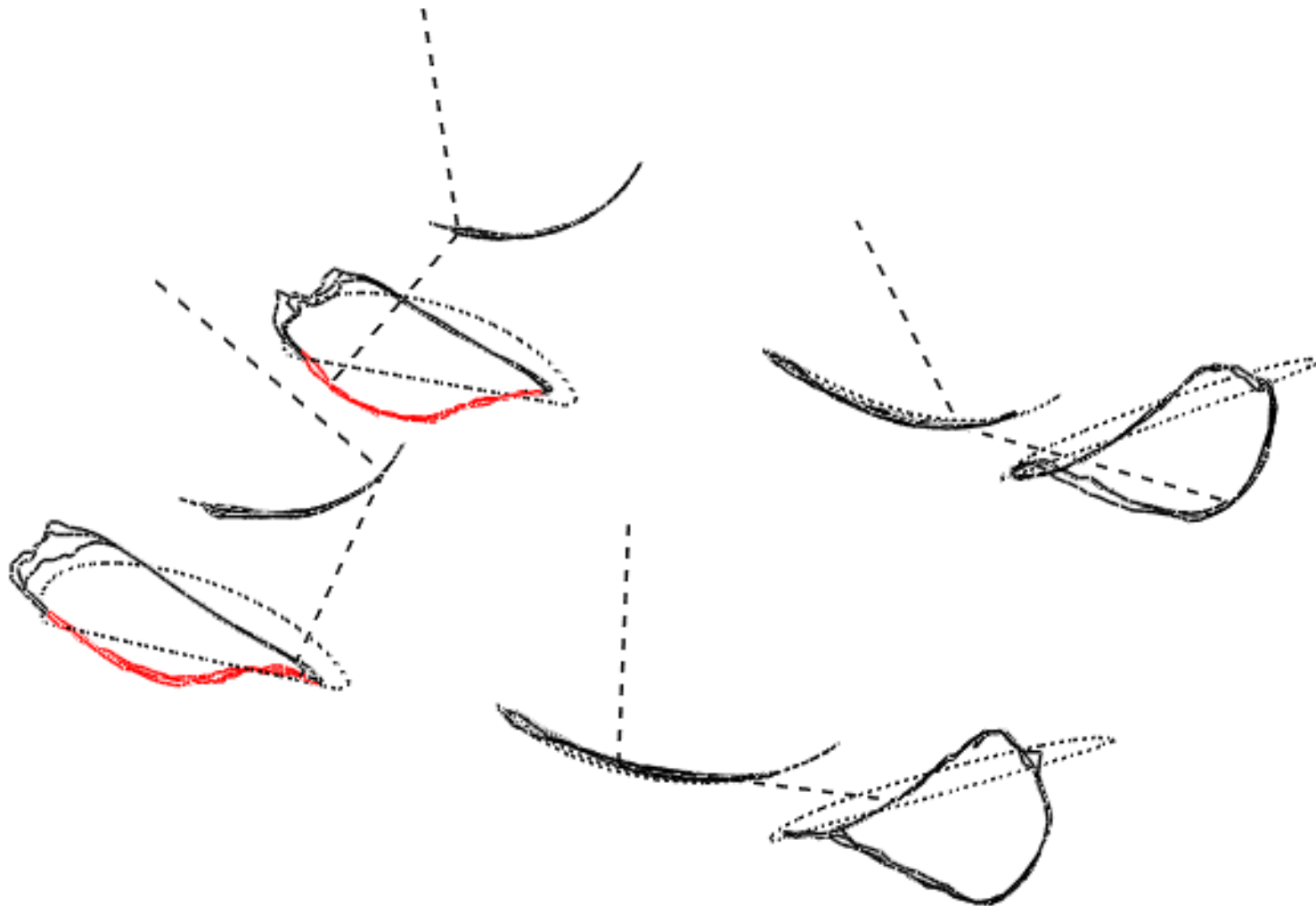


## Target Joint Angles and Actual Joint Angles without Ground Contact



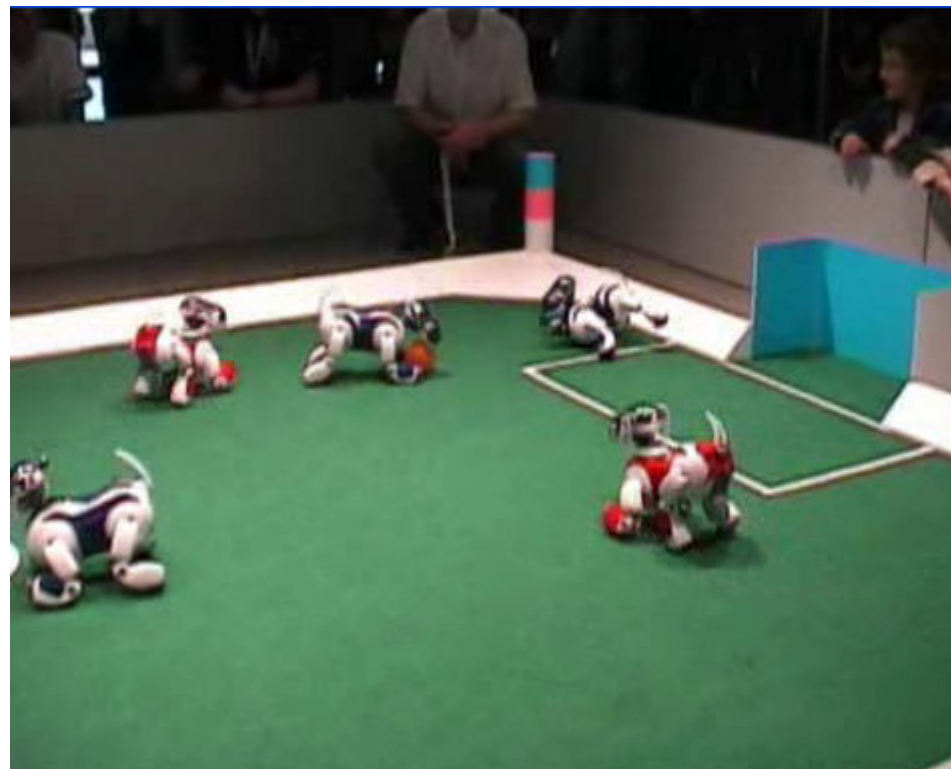


## Target Joint Angles and Actual Joint Angles with Ground Contact





## Resulting Gaits – ERS-7



Forward: 400 mm/s  
Backward: 296 mm/s  
Ball following: 331 mm/s, 195°/s



## Conclusions and Future Work

- ⚽ Conclusions
  - ⚽ Gait-Optimization
  - ⚽ Fitness Function based on odometry and vibrations
  - ⚽ Drawing of individuals
  - ⚽ Interpolation/extrapolation scheme for crossover
- ⚽ Future work
  - ⚽ Learning different parameter sets in parallel
  - ⚽ Interpolate between these sets even during learning







# Questions?

