

# ORCA – Organic Robot Control Architecture

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# Overview

- **Motivation**
- **ORCA-Architecture**
- **Demonstrator OSCAR**
- **Adaptive Filters**
- **Health-Signal**
- **Learning**
- **SILKE-Approach**
- **Outlook**

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# Motivation

## Autonomous mobile robots in human environments

unstructured,  
dynamically changing  
environment

no explicit model of  
the environment

-> fault-tolerance,  
safety



no explicit fault  
model

complex control  
systems

-> engineering  
bottleneck

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## Organic Computing can help

# ORCA - Organic Robot Control Architecture

## Goals:

- Self-organization: adapt to malfunctions without a formal model
- Learning: online and in-situ under hard real-time constraints
- Safety: avoid critical system states at any time
- Goal-Directedness: stable improvement of behavior
- Low Cost: overhead should be as low as possible

Approach: **controlled emergence** on lower functional control levels

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# ORCA - Organic Robot Control Architecture

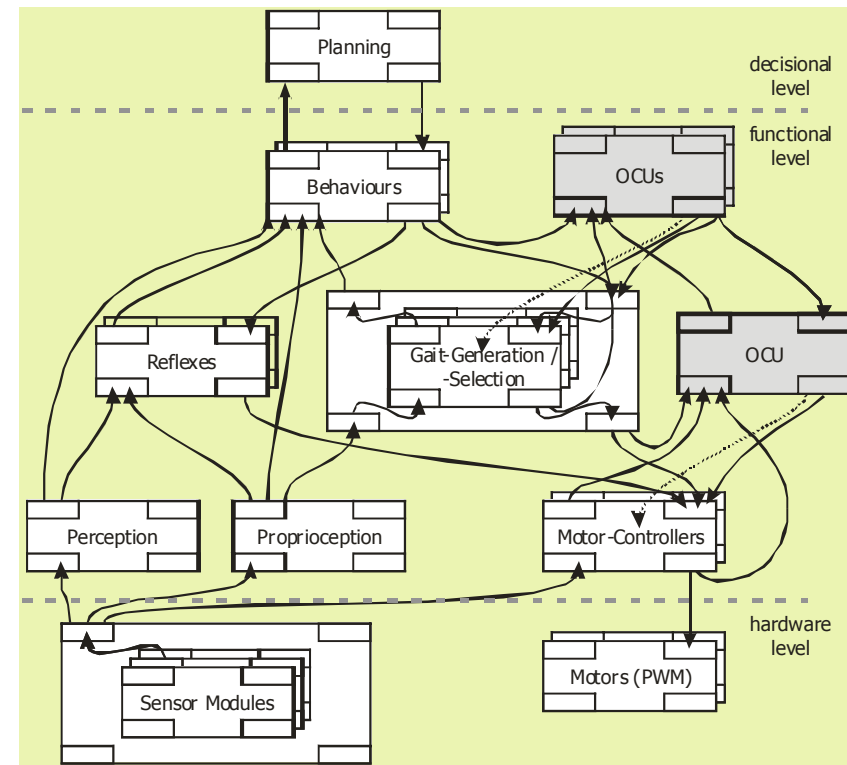
Hierarchic architecture

Components are selfdescribing

Within one level components combine each other autonomously (like p2p)

Components:

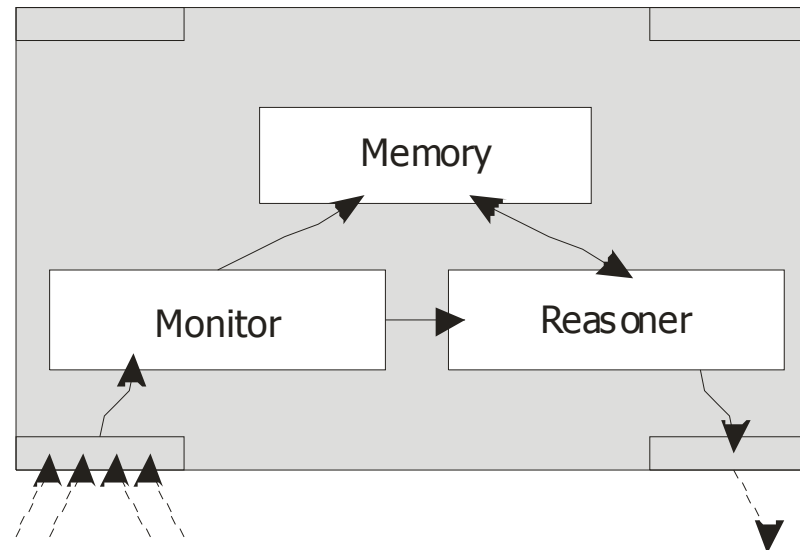
- BCU = Basic Control Unit
- OCU = Organic Control Unit



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# OCU



- Monitor: anomaly detection
- Memory: short term history
- Reasoner: hard real-time  
determination of a counteraction

# OSCAR

## Organic Self-Configuring and Adapting Robot

Hexapod with 18 DOF  
(Servos)

Ground contact Sensor

Control Computer:

- JControl/Smartdisplay
- Servo Driver Modul
- I2C-Bus
- PC (BlueTooth)

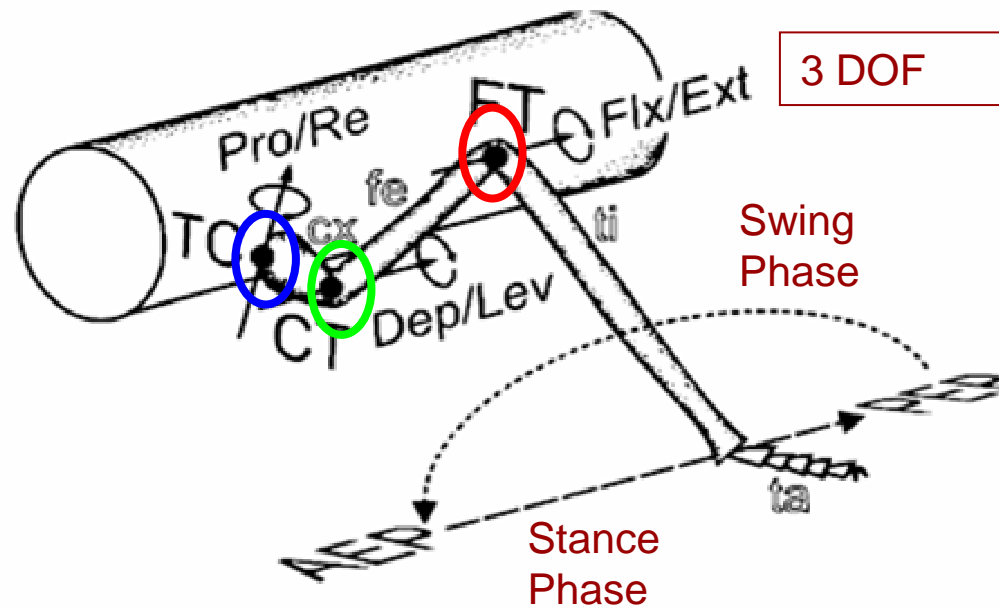
Programming Language:  
JAVA



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# Basic Approaches for Insect Walking



3 Joints:  $\alpha, \beta, \gamma$

AEP: Anterior Extreme Position  
PEP: Posterior Extreme Position

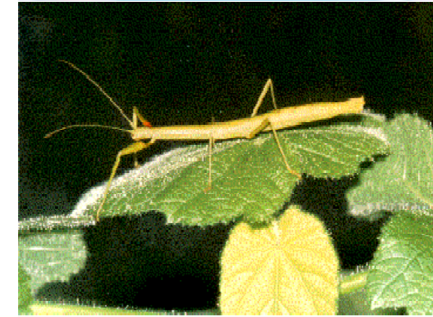
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# Distributed Walking Control in OSCAR

Insects produce **global** walking patterns with **local** coordination rules. [Cruse et. al. 90]

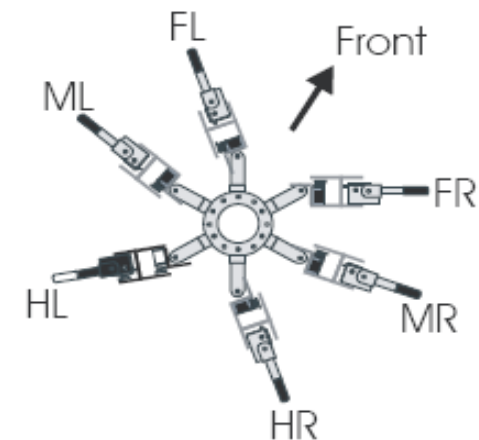


```
if (leg[i-1].groundContact() && (leg[i+1].groundContact())){  
    leg[i].swing();  
}  
else  
    leg[i].stance();  
}
```

So far only one single rule implemented with circular neighborhood relation for all legs.

Swing phase has constant length.

Duration of stance phase determines velocity.



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# Self-Organizing Gait Patterns in OSCAR



Slow Gait



Tetrapod Gait



Tripod Gait

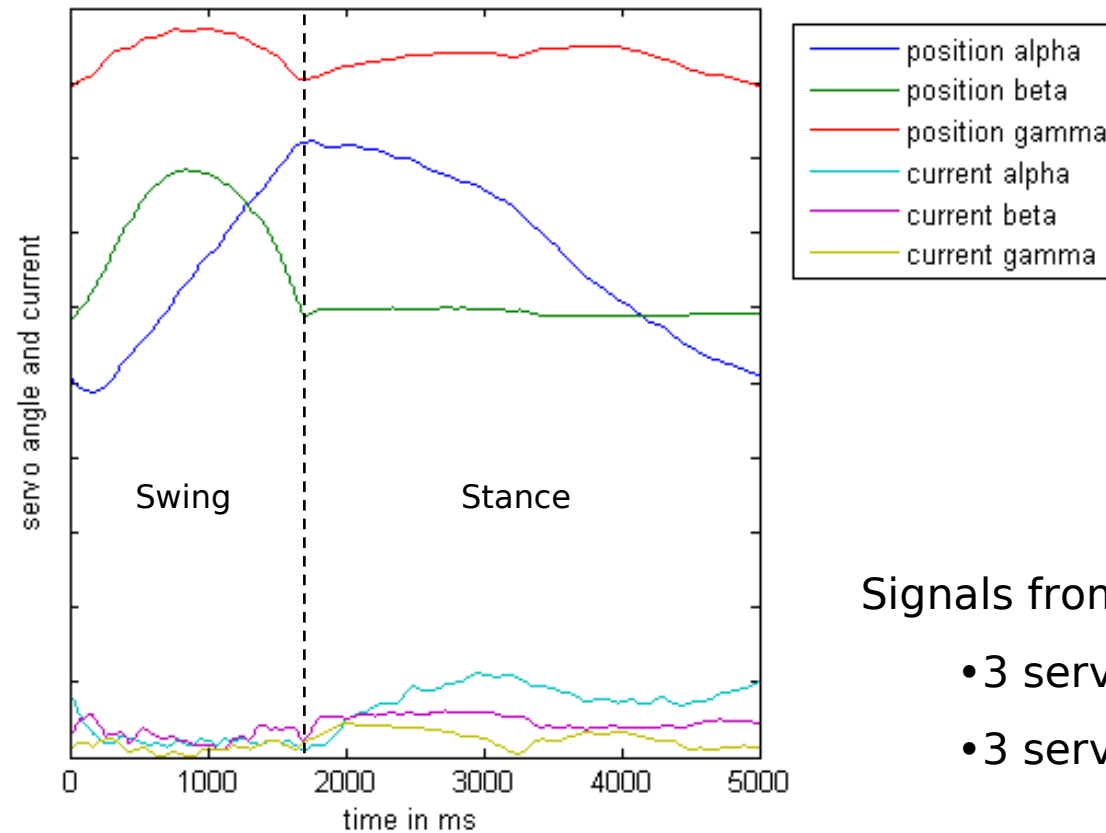


Emergent Gaits with  
Increasing Speed!

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# Monitoring of one Leg



Signals from one Leg:

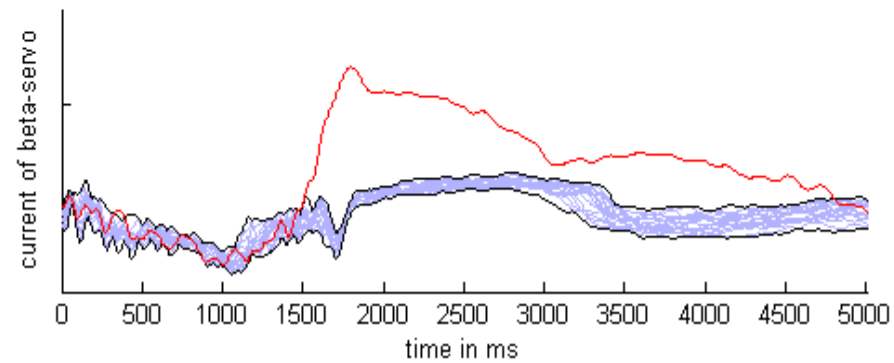
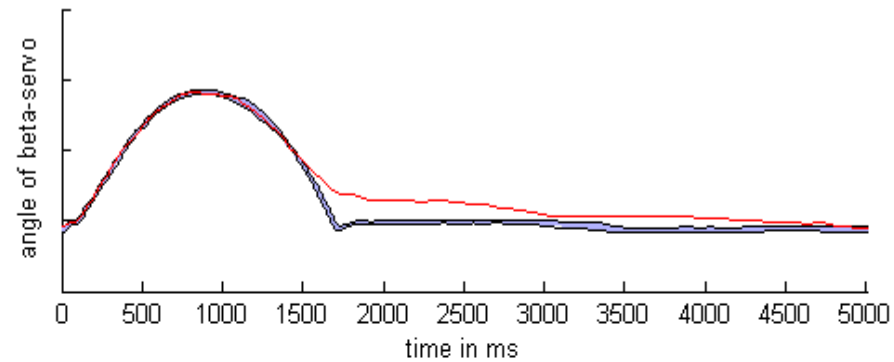
- 3 servo-positions ( $\alpha, \beta, \gamma$ )
- 3 servo-currents ( $\alpha, \beta, \gamma$ )

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# Monitoring – tolerance band

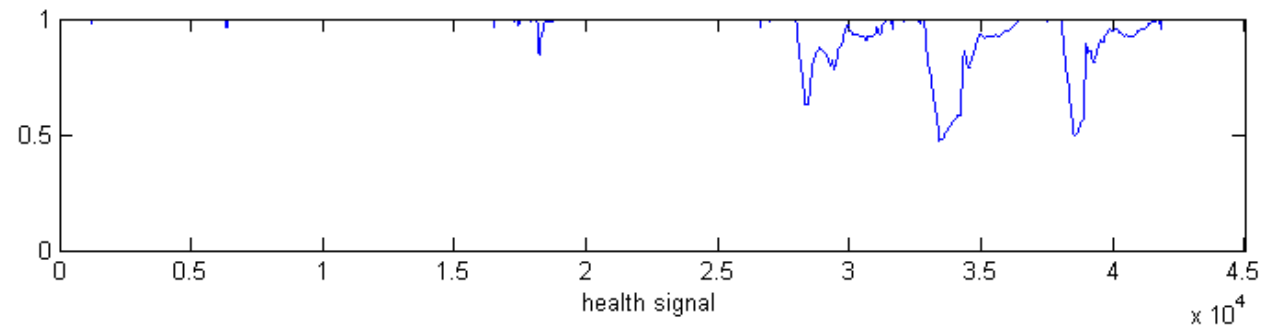
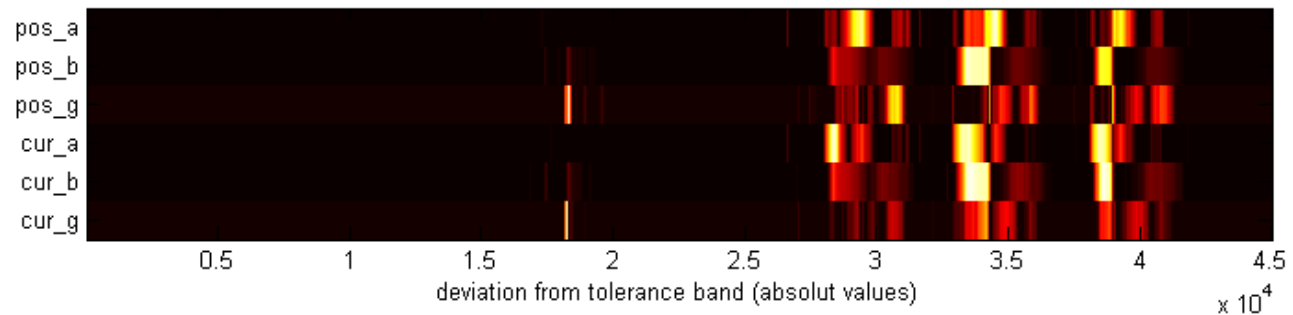
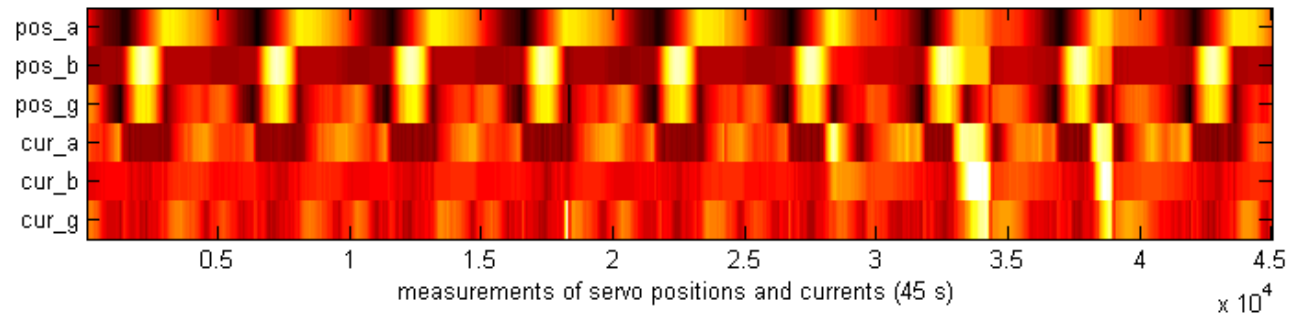
beta-servo:  
position and current



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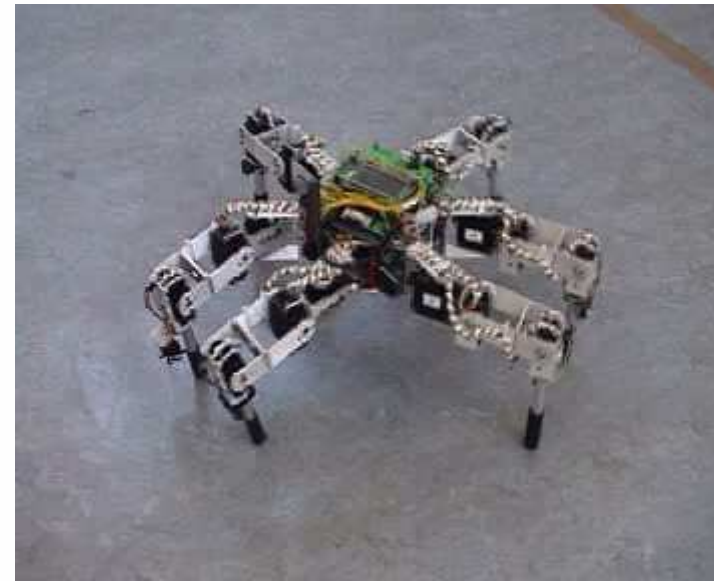
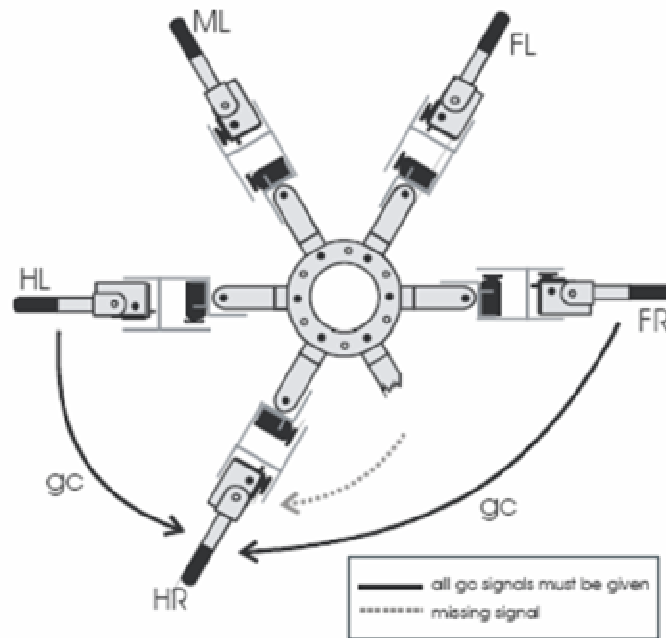
# Monitoring - Health-Signal



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# OSCAR Walking with Lost Middle Leg



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Same single rule, only  
neighborhood relation changes.

# Treatment of Local Component Faults in OSCAR by Means of Adaptive Filters

At FhG-IAIS, component faults in OSCAR and their effect on its trajectories were exhaustively simulated, using the robot simulator Gazebo.

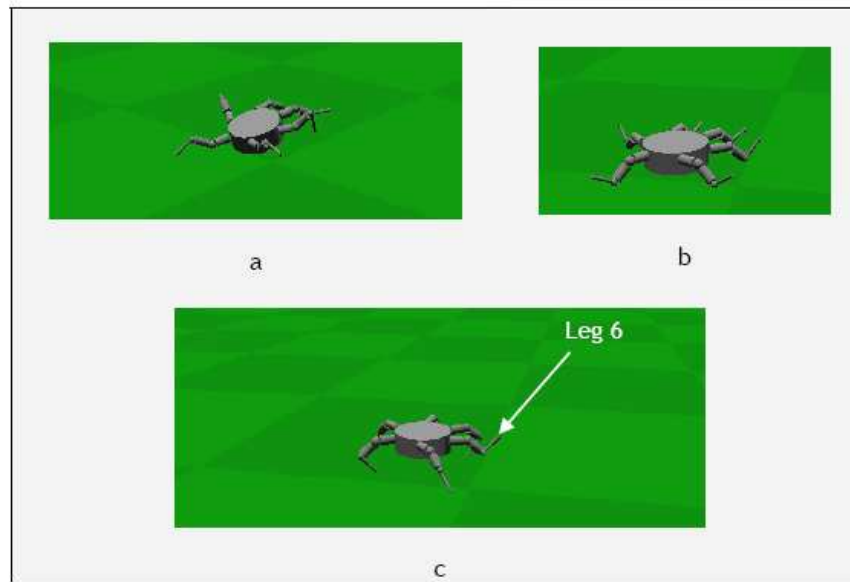
For compensating fault effects, the components as e.g. joints and servo motors are controlled by means of adaptive filter algorithms, i.e. output variables  $y$  are estimated by weighted sums of successively sensed values of a control variable  $x$ :

$$y(n) = \sum_{i=0}^{N-1} w_i(n) * x(n-i)$$

The weights  $w_i$  are trained to memorize the experience to compensate deviations of  $y$ .

The algorithms dealing with the fault effects inside a roboter leg were integrated to form a homogeneous Leg Fault Treatment Layer, i.e. this layer can autonomously compensate the different fault effects inside a robot leg.

# Compensation of Fault Effect



OSCAR getting healed after suffering from dislocated joints during motion

- (a) OSCAR falling down while walking.
- (b) Beta-joints healed. Controller will now heal gamma-joints.
- (c) OSCAR's pose after correction. Note that all legs are properly oriented except leg 6. This is because gamma-joint of this leg was already off the ground and therefore the controller skipped healing it.



# 'Health-Signal'

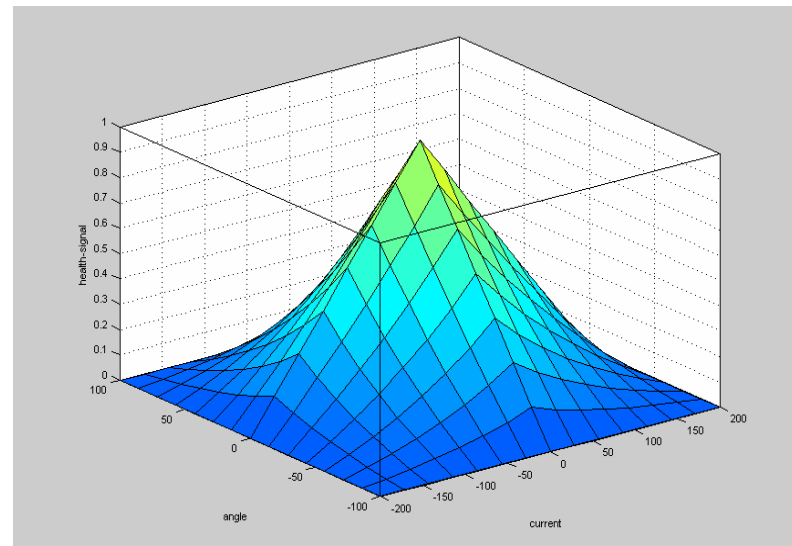
Specific aim:

- rate anomalies, e.g. irregular sensor or behaviour patterns
- at any system level (signals, modules, system itself)
- without a formal model
- potentially capable of learning

Approach:

- fusion and abstraction of data, states and module activities
- gradual severity index:           1 = completely normal  
  0 = totally abnormal behavior

-> (neuro-) **fuzzy-system** to generate the health-signals



Example:  
Health-signal  
versus current  
and angle errors

# Demonstrator CARL

(Creeping Autonomous Robot for Learning demonstrations)

Motivation:

- generalization of investigations on OSCAR
- simpler, but scalable testbed for
  - architectural issues
  - health-signal
  - emergent behavior
  - online learning
  - self-organization

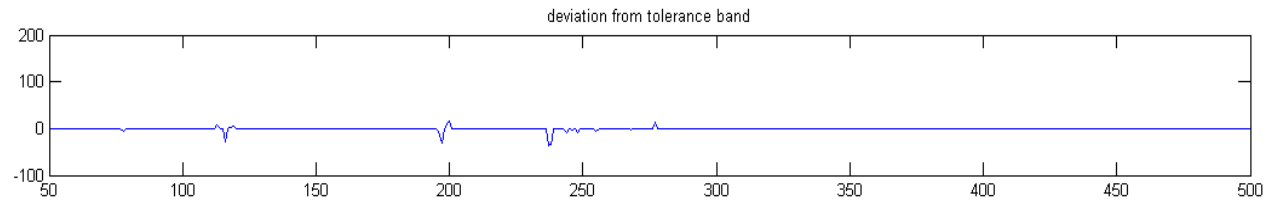
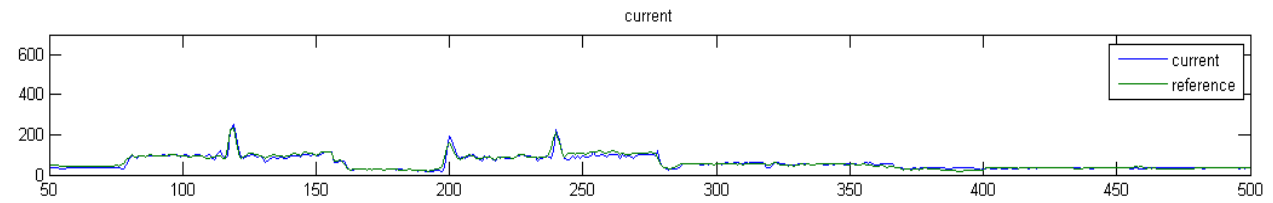


Current state:

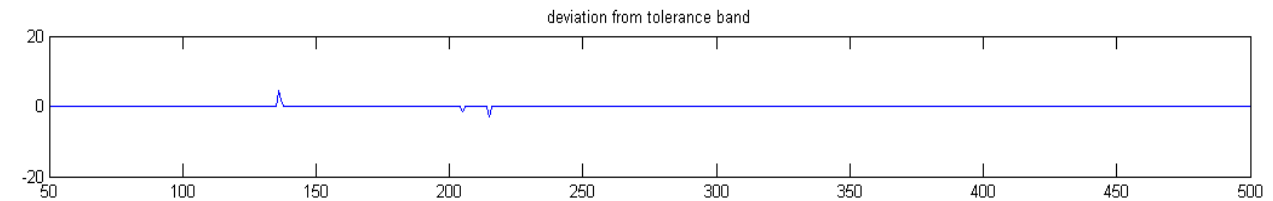
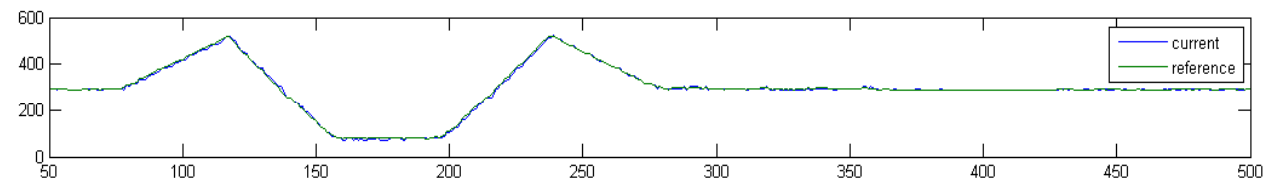
- health-signal
- emergent global behavior out of local rules
- first steps of learning

# Results

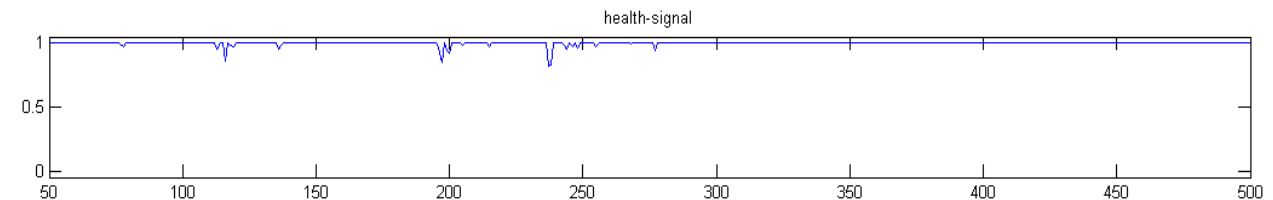
Motor  
Current:



Joint  
Angle:

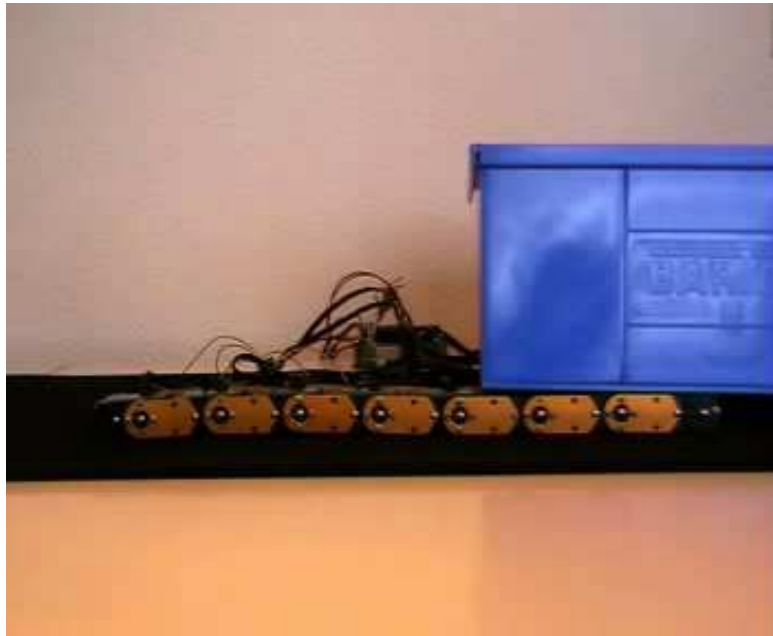


Health-  
Signal:

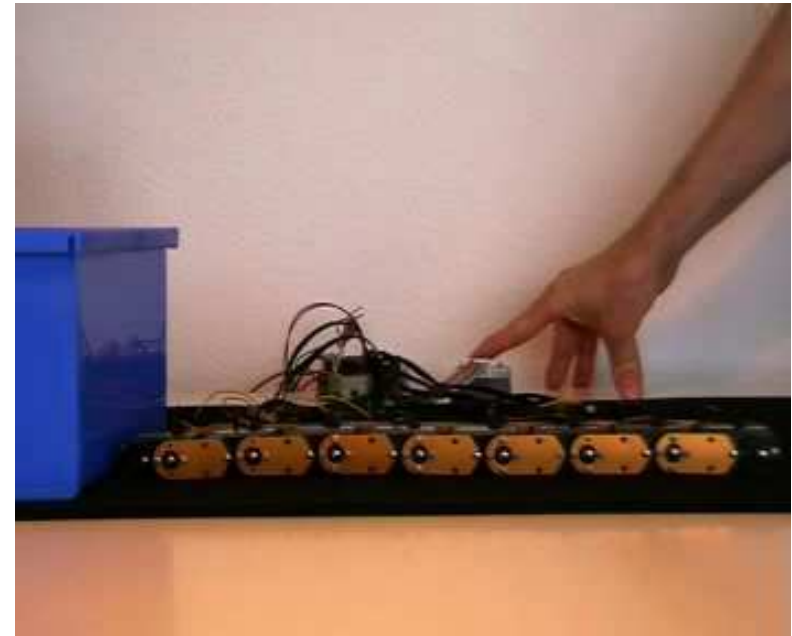


Normal behavior of CARL

# Test Scenarios



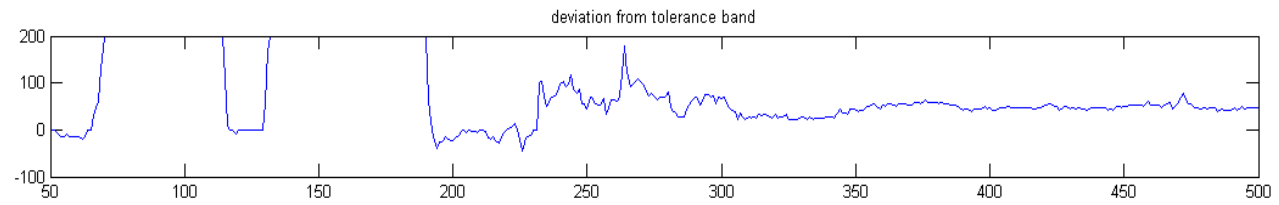
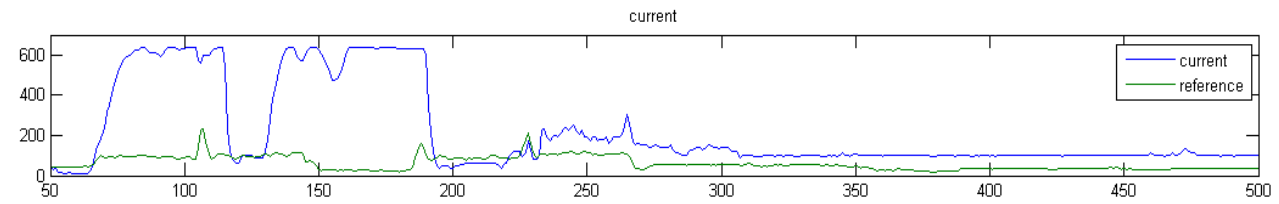
Blocked upward movement



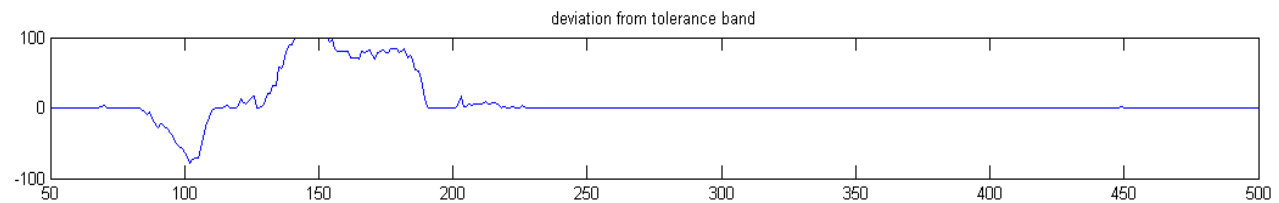
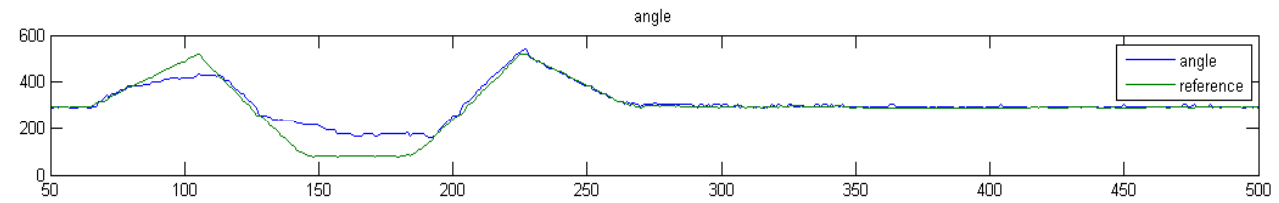
Blocked forward movement

# Results

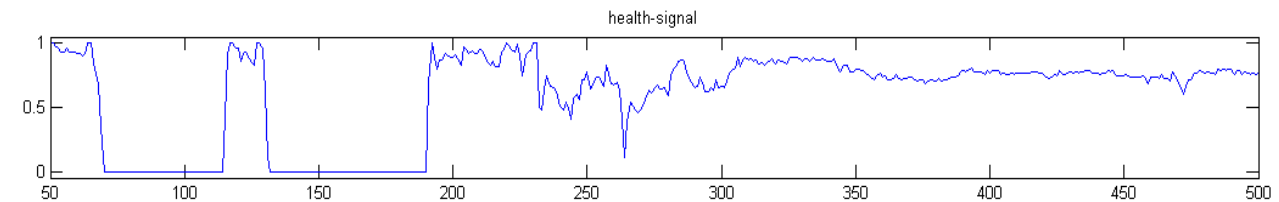
Motor  
Current:



Joint  
Angle:

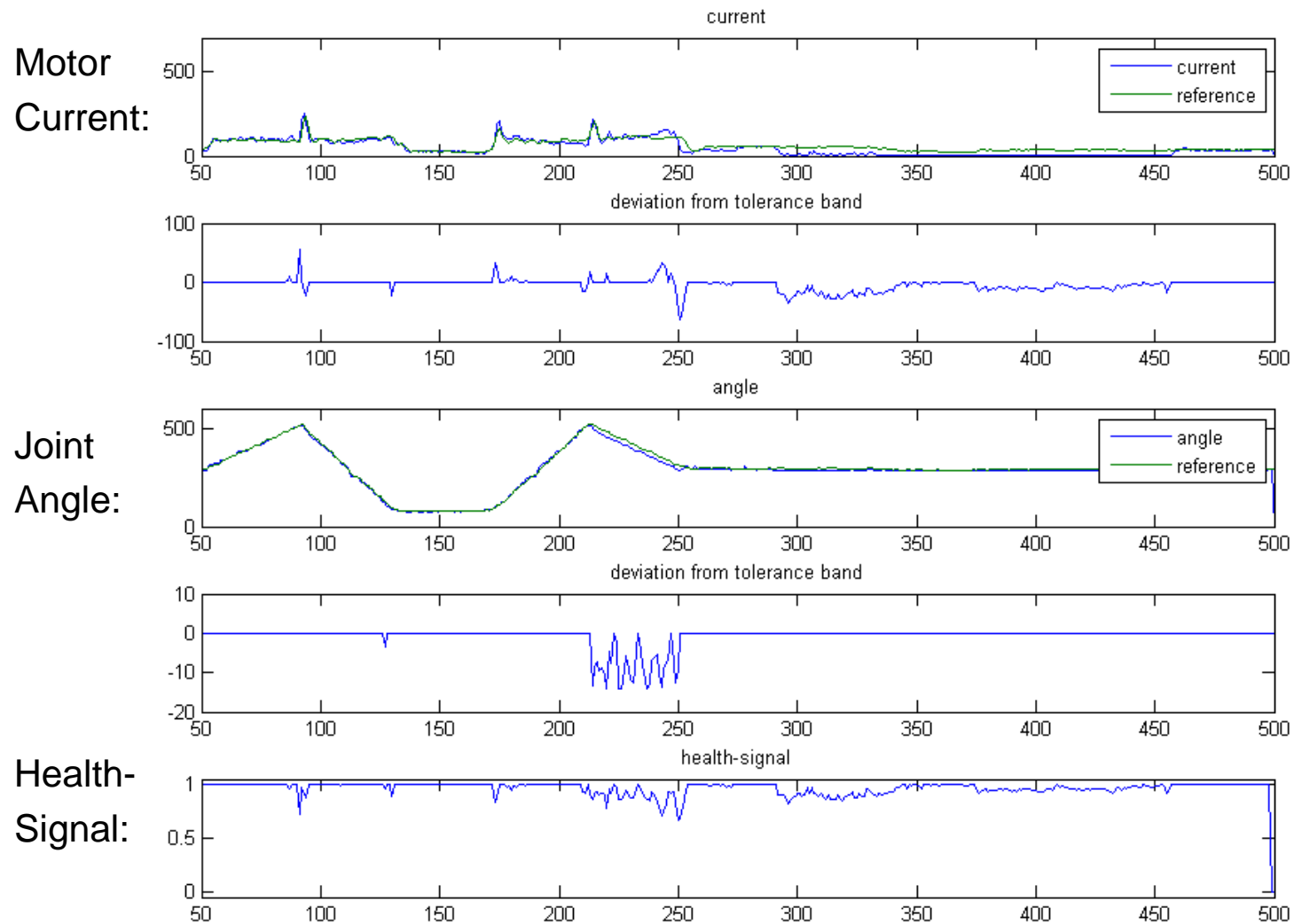


Health-  
Signal:



Blocked upward movement

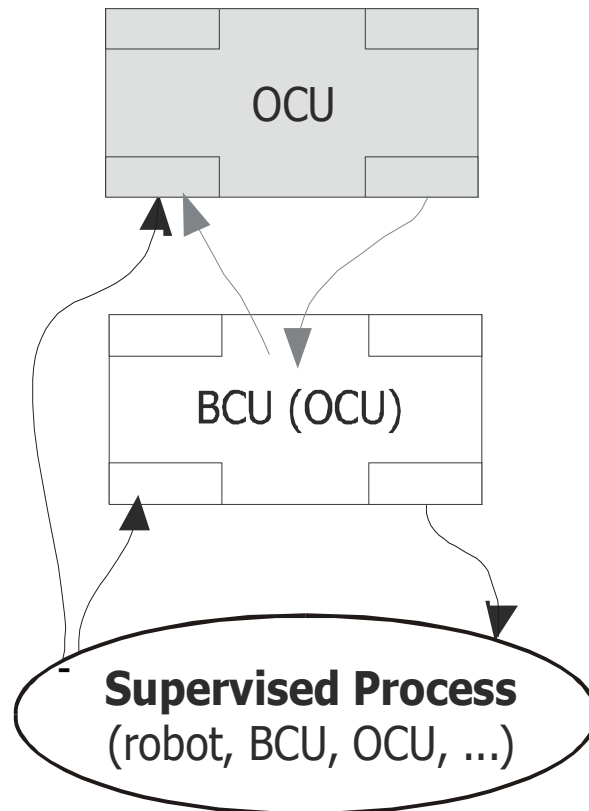
# Results



Blocked forward movement

# Learning-Basis

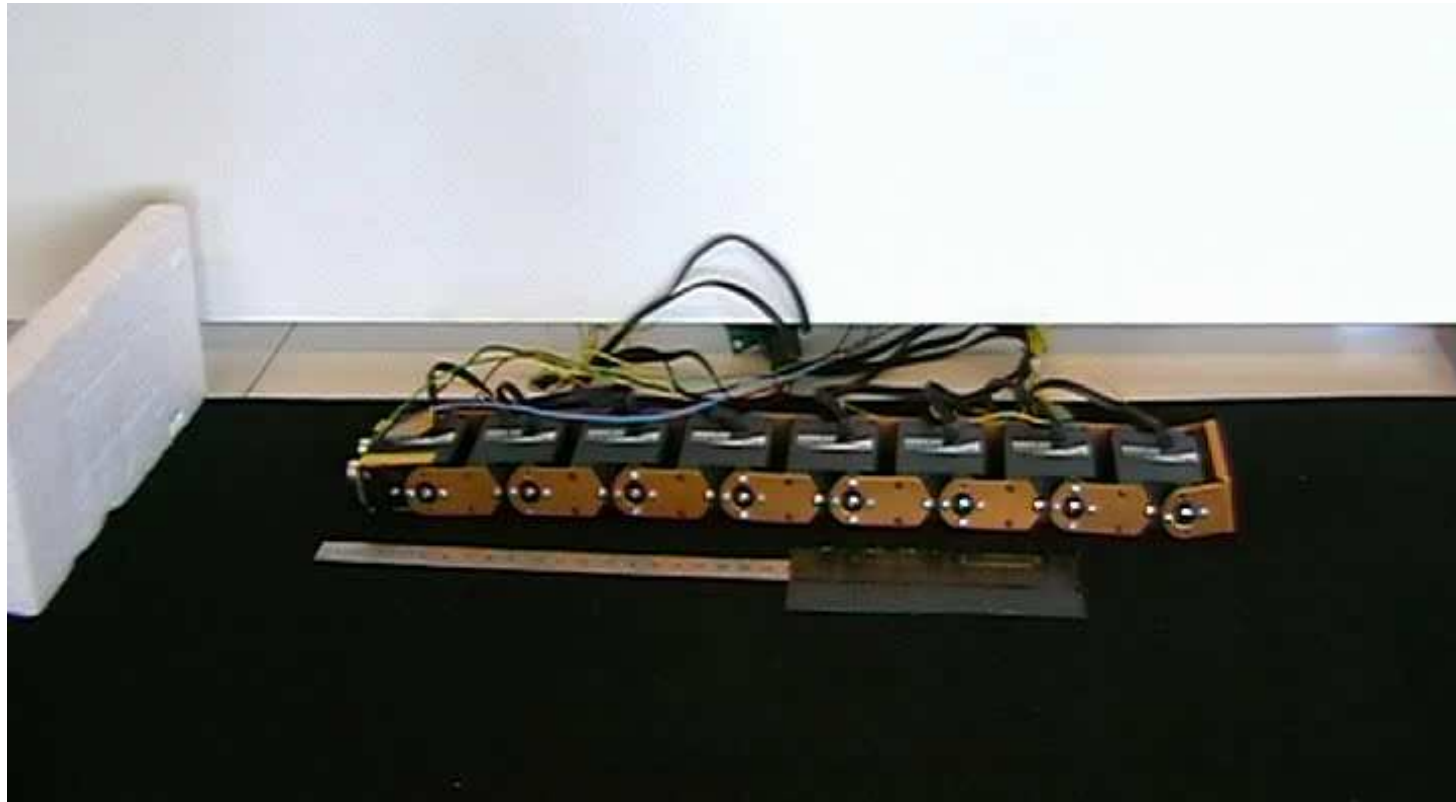
Basic approach:



**guarded online-learning**

- online and in-situ
- under hard real-time constraints
- hybrid crisp-fuzzy systems as learning paradigm
  - > partitioning of the input space
- avoidance of safety-critical states at any time
  - > safety warranty by tag-mechanism

# First Learning Experiment



Aim: optimizing the step width  
Sensory input: change in distance to block  
Tuning parameter: joint angle of wave

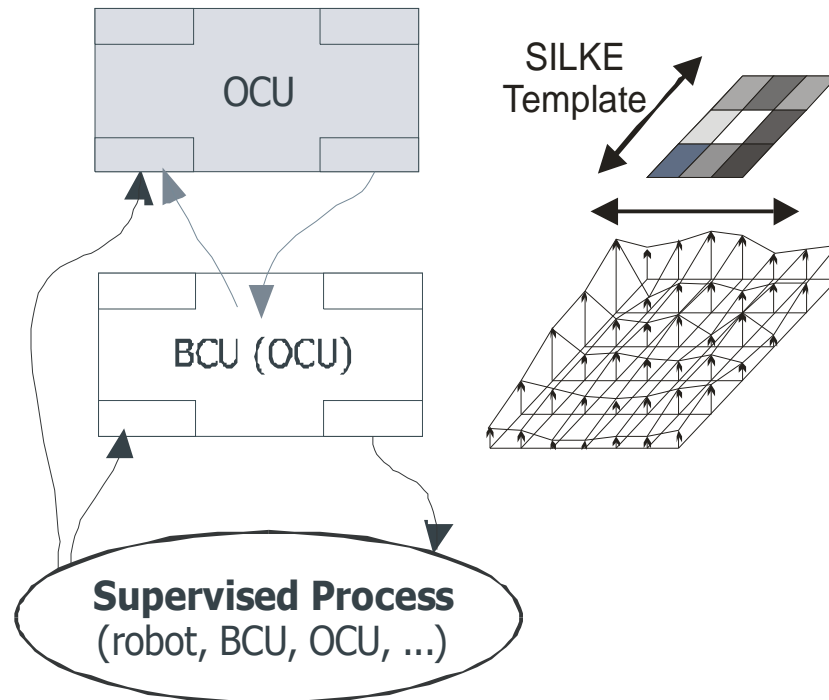


# SILKE-Approach

(System to Immunize Learning Knowledge-based Elements)

Aim: (meta-level) control of learning dynamics

- avoid harsh control actions
- stabilize convergence
- avoid uninitialized learning situations



Approach:

- locally operating template
- operation on 'cell-level'
- complementary to
  - safety-rules
  - OCUs

Operators:

- delimiting the steepness
- averaging
- ...

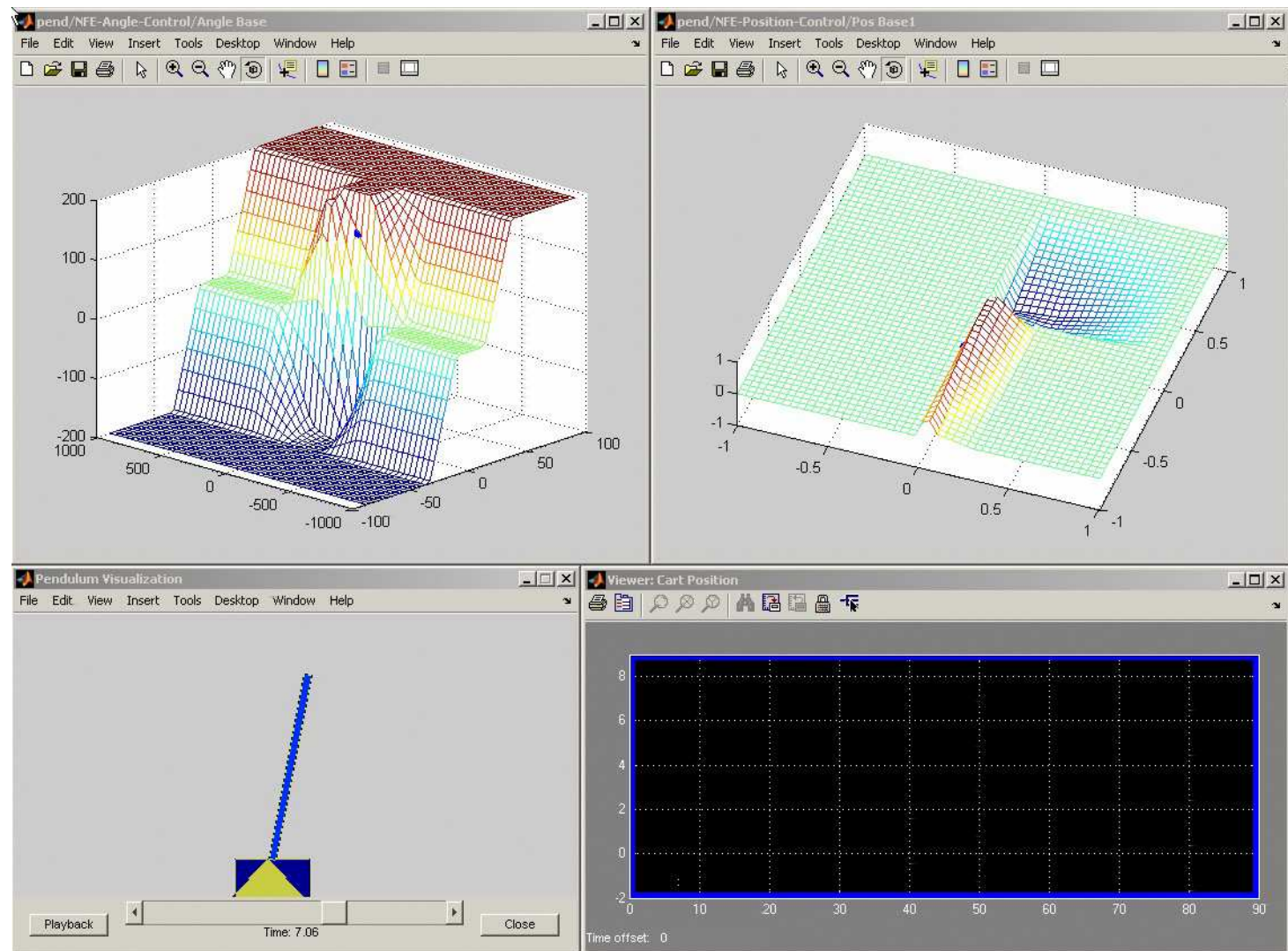
# Background

- Class of application:
- interacting learning systems
  - learning control of potentially instable, 'unsafe' systems



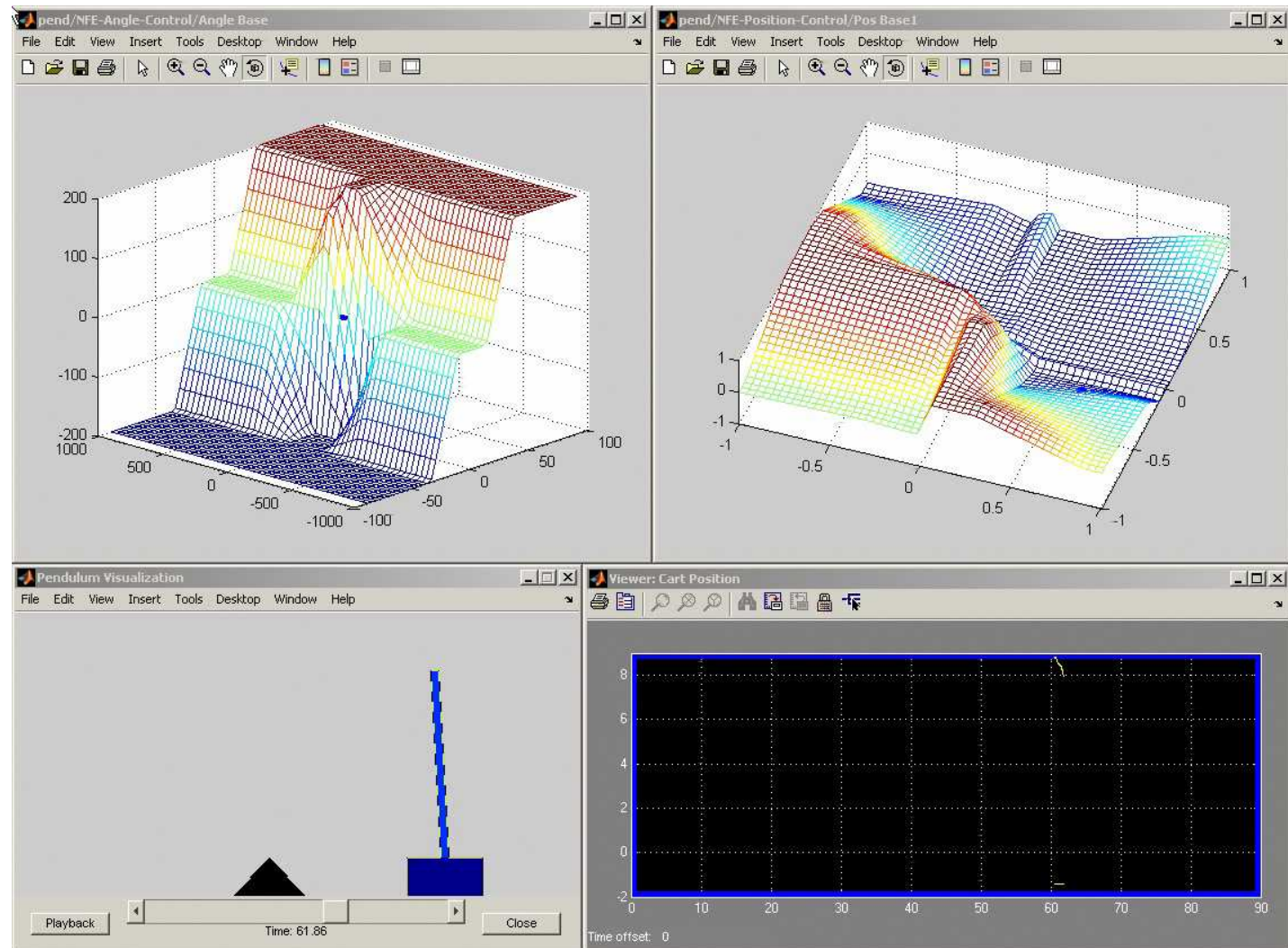
Concrete investigation: interacting learning balance and position controllers  
(in simulations)

# Learning Behavior without SILKE



Interacting learning balance and position controllers

# Learning Behavior with SILKE



Interacting learning balance and position controllers

# Outlook

- ORCA:
- architectural issues (e.g. separate health-signal hierarchy?)
  - self-organizing components
- Health-signal:
- learning of (more diverse) reference behaviors
  - abstraction and fusion of health-signals
- Learning:
- investigations on OSCAR / CARL and in faulty situations
  - control of interacting learning systems
- SILKE:
- investigations under real-life conditions (OSCAR / CARL)
  - development of new template operators
- OSCAR / CARL:
- additional sensors
  - complex environment & fault scenarios
  - simulation results verified on real machines

# Publications

1. Brockmann, W.; Maehle, E.; Mösch, F.; *Organic Fault-Tolerant Control Architecture for Robotic Applications*. 4th IARP/IEEE-RAS/EURON Workshop on Dependable Robots in Human Environments, Nagoya, Japan, Juni 16-18, 2005, Nagoya University/Japan 2005
2. Großpietsch, K.-E.; *Adaptive Filters for the Dependable Control of Autonomous Robot Systems*. Proc. DPDNS 05 Workshop, Denver 2005
3. Brockmann, W.; Großpietsch, K.-E.; Maehle, E.; Mösch, F.; *ORCA – Eine Organic Computing-Architektur für Fehlertoleranz in autonomen mobilen Robotern*. Mitteilungen der GI/ITG-Fachgruppe Fehlertolerierende Rechensysteme, Nr. 33, März 2006
4. El Sayed Auf, A.; Mösch, F.; Litza, M.; *How the Six-Legged Walking Machine OSCAR Handle Leg Amputation*, To appear at 9th Int. Conf. Simulation of Adaptive Behaviour – SAB'06, Sept 2006.
5. Großpietsch, K.-E.; Silayeva, T.A.; *Organic Computing - A New Paradigm for Achieving Self-Organized Dependable Behaviour in Complex IT Systems*. To appear in: Proc. IDIMT 2006 Workshop, Ceske Budejovice (Czech Republik), September 13-15, 2006
6. Raman Agarwal: *Intelligent Algorithms for Controlling an Autonomous Robot system*. Master Thesis, Fraunhofer IAIS, Sept 2006
7. Brockmann, W.; Großpietsch, K.-E.; Kleinlützum, K.; Maehle, E.; Meyer, D.M.; Mösch, F.: *Concept for a Fault-Tolerant Control Architecture for CLAWAR Machines*. To appear: 9. Int. Conf. Climbing and Walking Robots and the Support Technologies for Mobile Machines, Brussels, Belgium, Sept. 2006
8. Mösch, F.; Litza, M.; El Sayed Auf, A.; Maehle, E.; Großpietsch, K.-E.; Brockmann, W.: *ORCA – Towards an Organic Robotic Control*. To Appear: IWSOS - International Workshop on Self-Organizing Systems, Sept. 2006
9. Jakimovski B.; Litza, M.; Mösch, F.; El Sayed Auf, A.; *Development of an Organic Computing Architecture for Robot Control*. To Appear: GI-Workshop on Organic Computing – Status and Outlook, Dresden, Oct. 2006
10. Brockmann, W.; Meyer, D.M.; *Ein immunsystem-inspirierter Ansatz zum Überwachen des Lernens in Neuro-Fuzzy-Systemen*. To appear: 16. Workshop „Computational Intelligence“, Bommerholz, Nov. 2006

