## Winning Humanoid Soccer Robots of Team NimbRo: Mechatronics, Perception, Control, and Learning

#### **Sven Behnke**

Autonomous Intelligent Systems



## RoboCup German Open 2005





#### **Humanoid Soccer Final RoboCup 2005**





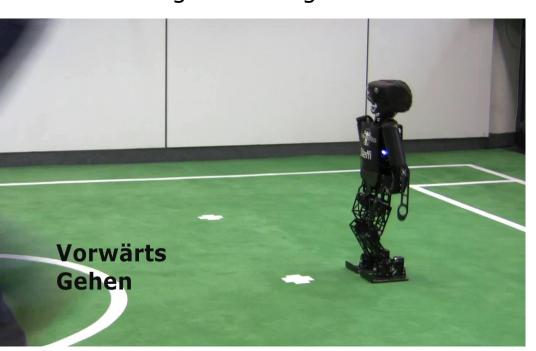
## RoboCup 2008 KidSize Final NimbRo vs. Team Osaka

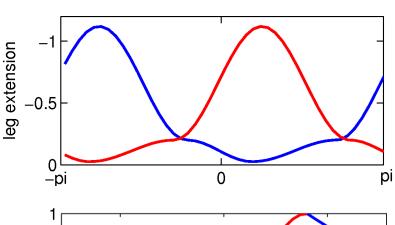


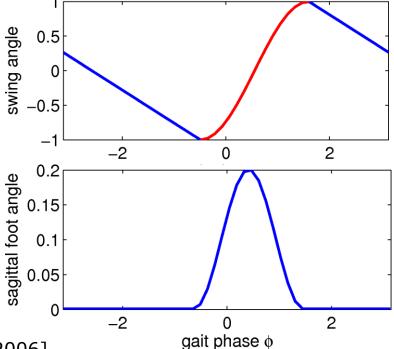


#### **Omnidirectional Walking**

- Continuously changing walking speeds: sagittal, lateral, yaw
- Key ingredients:
  - Rhythmic weight shifting
  - Leg shortening
  - Swing in walking direction









[Behnke: ICRA 2006]

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## RoboCup 2013 Final





## **Capture Step Framework**

Step parameters

**Balance Control** Motion Generator Velocity input:  $(x, \dot{x}, y, \dot{y})$ Motor targets Robot State Estimation

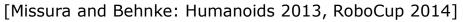


[Missura, Behnke: Humanoids 2013, RoboCup 2014]



## **Omnidirectional Capture Steps**

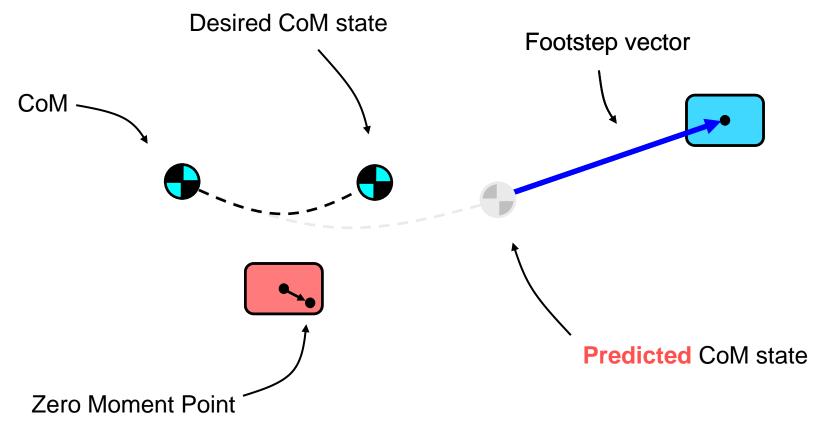






#### **Balance Control**

Adapt ZMP, timing, and foot placement



[Missura and Behnke: Humanoids 2013, RoboCup 2014]



#### **Dynaped with Small Feet**

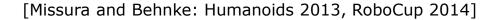


## Dynaped with Small Feet

August 2014, Bonn



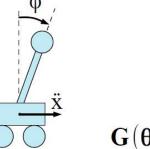




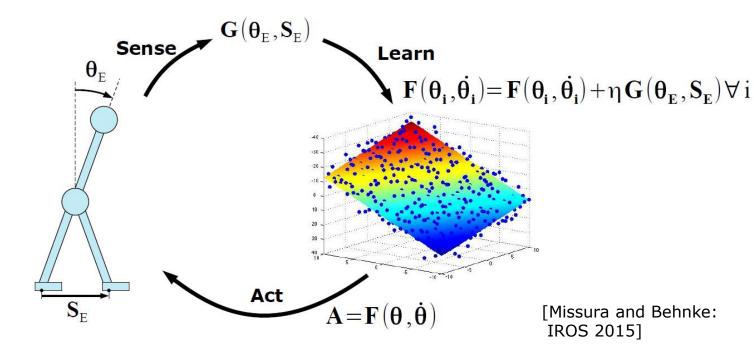


#### **Online Learning of Foot Placement**

- Function approximator for step size
- Online update based on tilt and step size error



$$\mathbf{G}(\mathbf{\theta}_{\mathbf{E}}, \mathbf{S}_{\mathbf{E}}) = \mathbf{\theta}_{\mathbf{E}} + \mathbf{p}_{1} \tanh(\mathbf{p}_{2} \mathbf{S}_{\mathbf{E}})$$





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#### **Online Learning of Foot Placement**



[Missura and Behnke: IROS 2015]



## igus Humanoid Open Platform

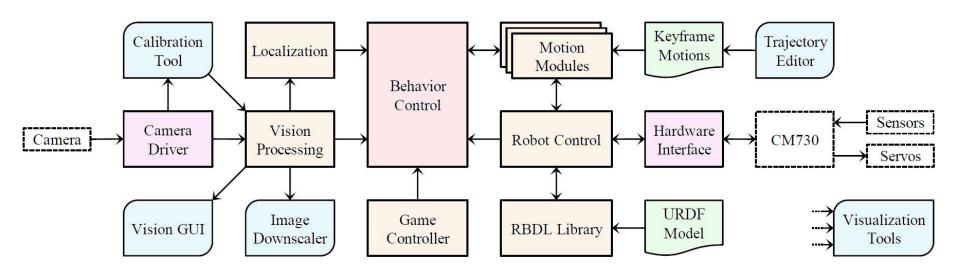
- 90 cm, 6.6 kg
- 3D printed structure
- 20 DoF
- Dual-core PC
- Wide-angle camera(s)
- IMU
- ROS-based software
- Hard- and software released: nimbro.net/OP

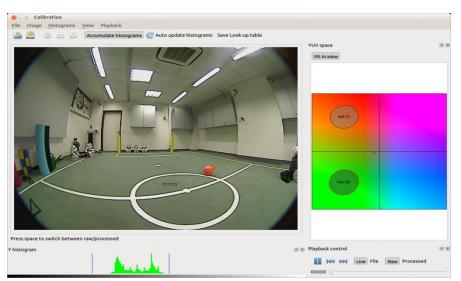
[Allgeuer et al. Humanoids 2015]





#### **ROS-based Software**







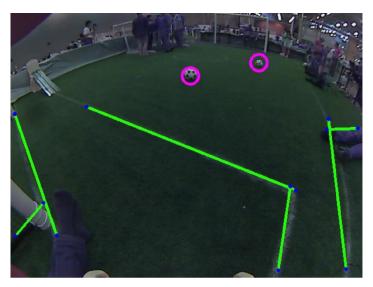


[Allgeuer et al. Humanoids 2015]

Sven Behnke: Winning Humanoid Soccer Robots of Team NimbRo

#### Perception of the Game Situation

- Less relying on color
- Learned ball detection
- Goal detection



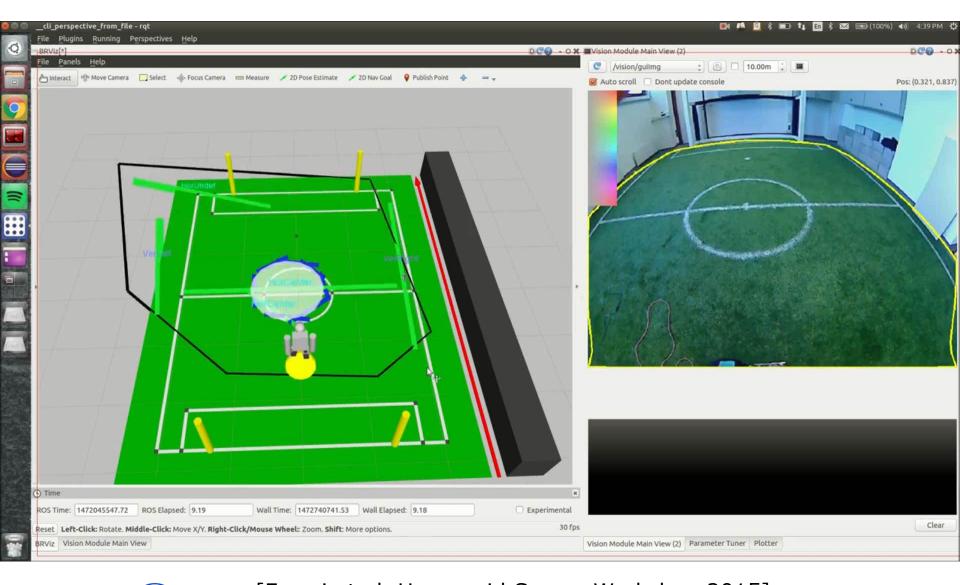






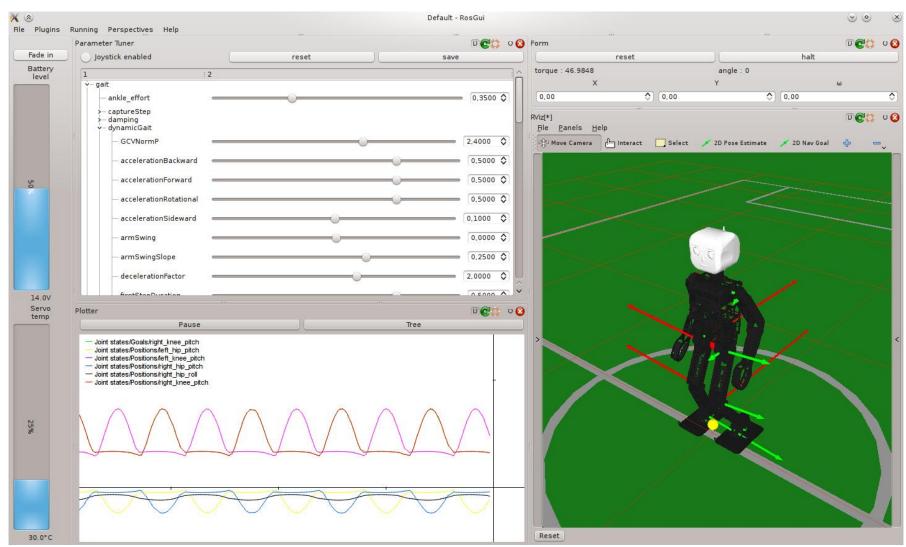
[Farazi et al. Humanoid Soccer Workshop 2015] Sven Behnke: Winning Humanoid Soccer Robots of Team NimbRo

#### Localization



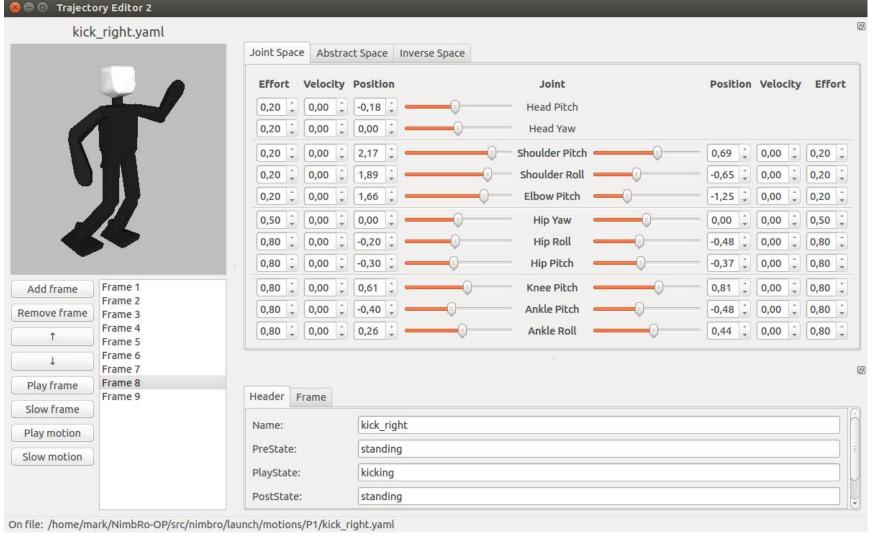


#### **Rviz 3D Visualization**





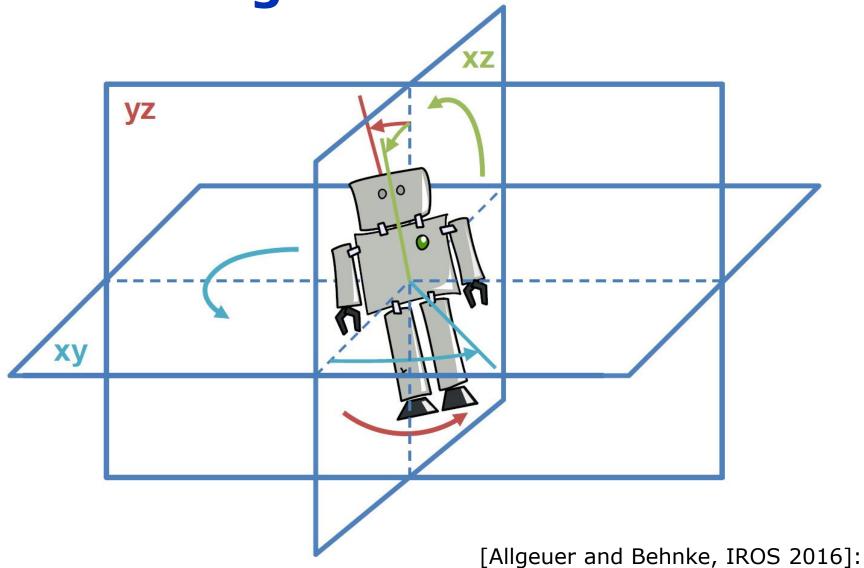
## **Trajectory Editor**





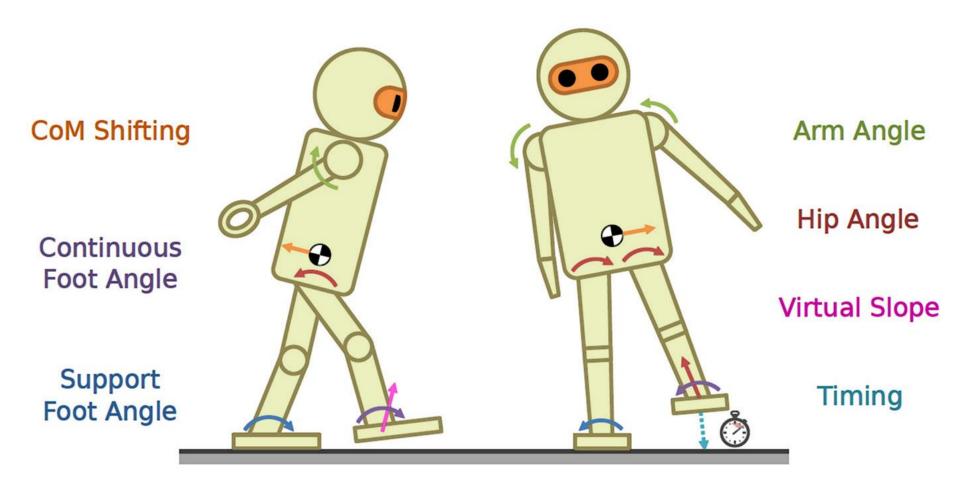
[Allgeuer et al. Humanoid Soccer Workshop 2013]

**Fused Angles** 





#### **Feedback Mechanisms**



[Allgeuer and Behnke: Humanoids 2016]



#### **PD Feedback**



[Allgeuer and Behnke: Humanoids 2016]



## **Landing Motion Backwards**





## **Landing Motion Forwards**





## **Getting Up**



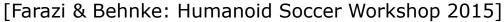
[Allgeuer et al. Humanoids 2015]



Behnke: Humanoid Robots – From Playing Soccer to Rescue Operations

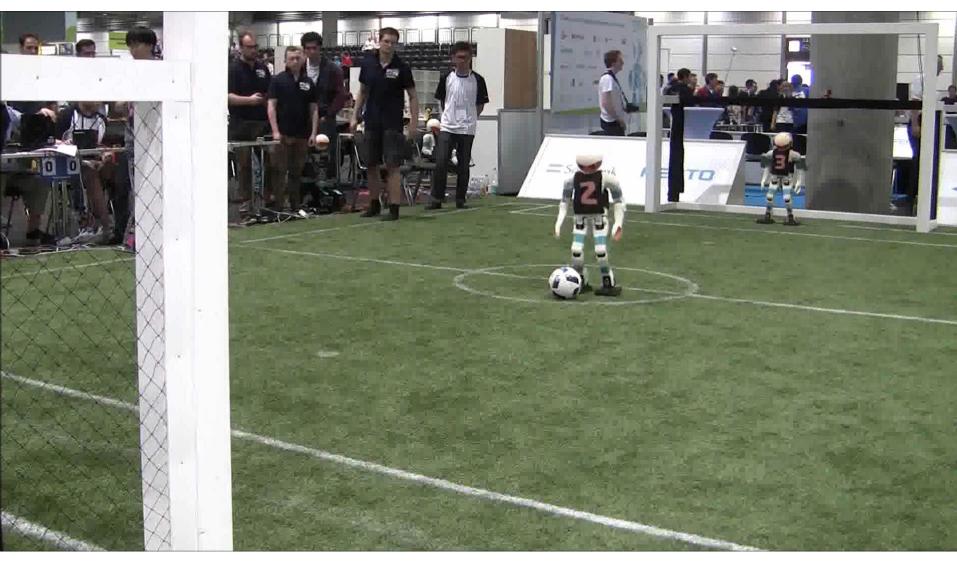
## **Visual Perception**

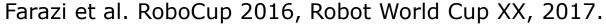






#### RoboCup 2016 TeenSize Final







Sven Behnke: Winning Humanoid Soccer Robots of Team NimbRo

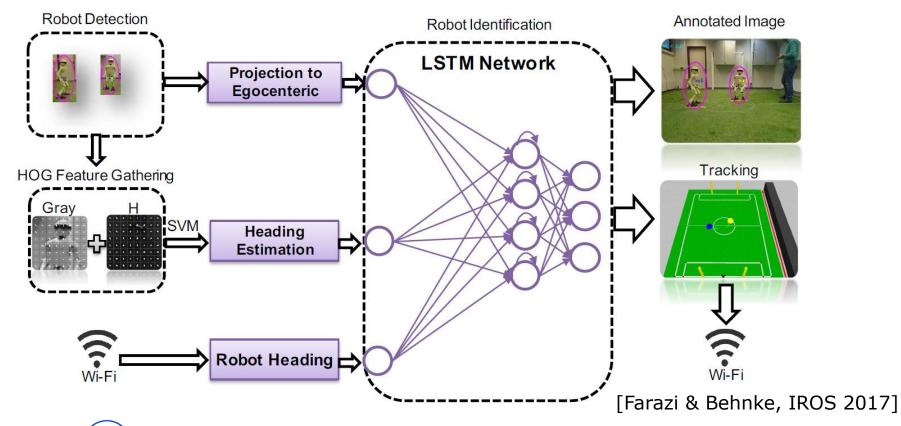
#### **Team NimbRo TeenSize 2016**





## Robot Detection, Tracking & Identification

- Based on visual detections and compass
- Learning data association





#### **Robot Detection & Pose Estimation**

Based on HoG features



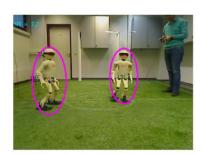




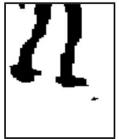




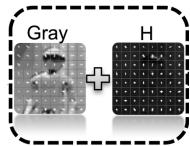
Scan line feet estimation







- Heading estimation
  - Dense HoG
  - SVM multiclass classifier



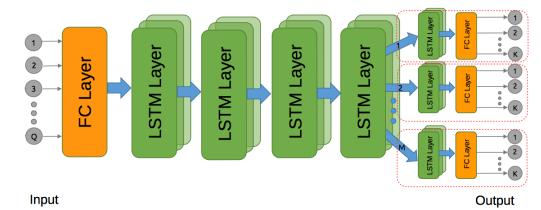


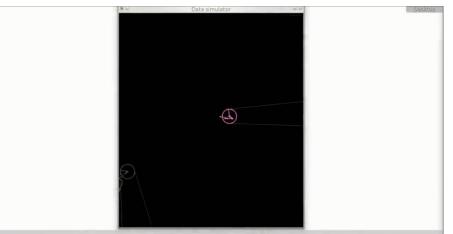
[Farazi & Behnke, IROS 2017]



#### **Learning Data Association**

- Recurrent neural network
- Training with simulated data





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Setup	M=3, N=2	M=5, N=3	M=10, N=7
Human	94.3%	86.3%	67.3%
Kalman-HA	75.6%	72.2%	53.1%
ours	96.2%	<b>87.1</b> %	66.5%

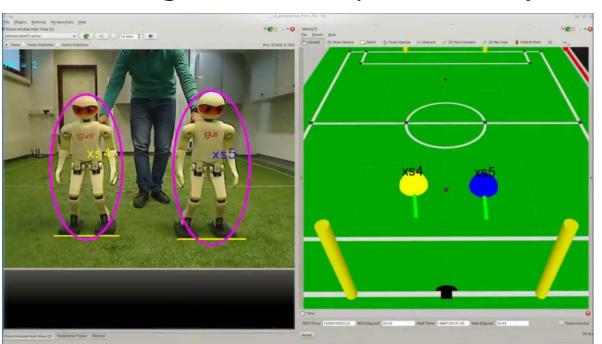
Fine-tuning on real data

[Farazi & Behnke, IROS 2017]



## **Real-Robot Experiment**

- Three Igus humanoid robots, observer in goal area
- Randomly chosen sequences, 3140 frames in total
- Partial, short term and long term occlusions, Single forward pass 4ms (≈250Hz)



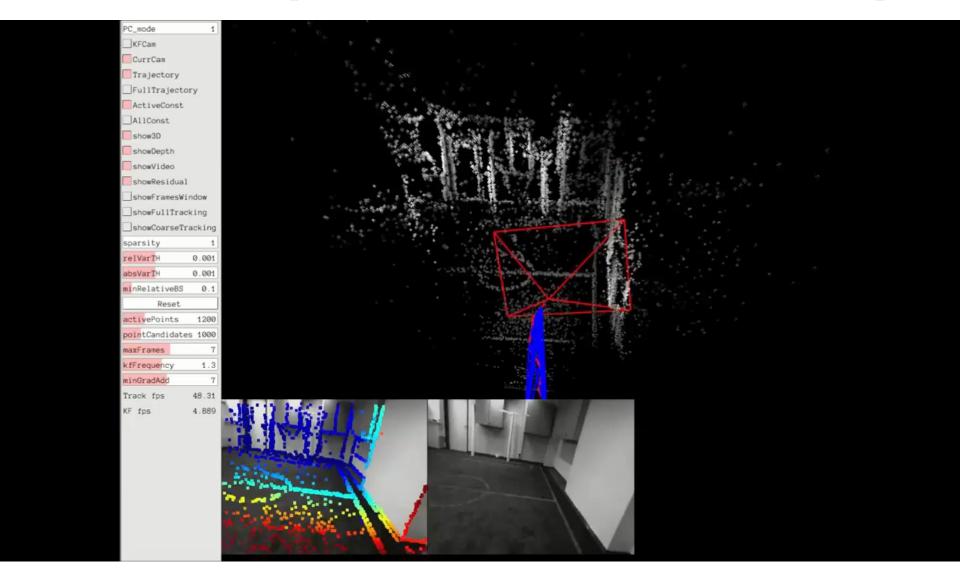
Baseline	Kalman-HA	Kalman-HA2	JPDA	Ours
Average error	0.67 m	0.30 m	$0.29\mathrm{m}$	0.22 m

Frames	200	400	800	Total
Kalman-HA	73.2%	75.5%	72.1%	73.8%
Kalman-HA2	87.2%	84.0%	86.3%	85.5%
JPDA	87.1%	84.6%	85.6%	86.3%
Deep LSTM (ours)	89.8%	90.3%	92.4%	91.1%

[Farazi & Behnke, IROS 2017]

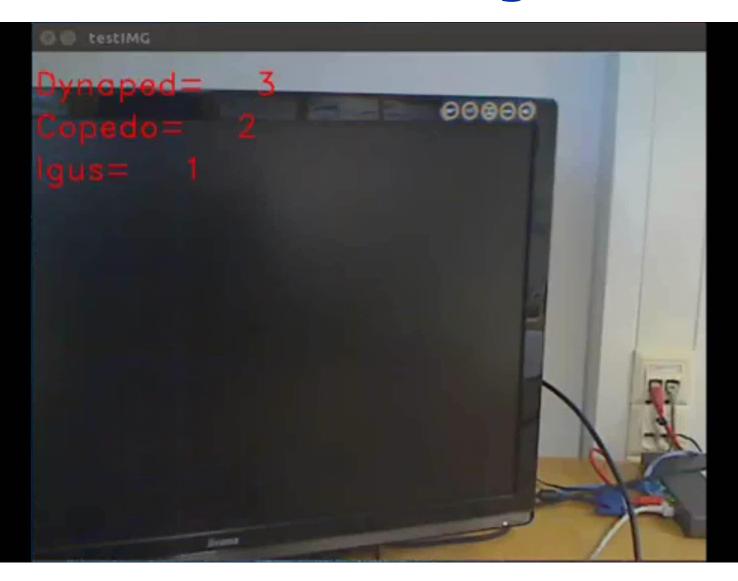


## **Direct Sparse Visual Odometry**

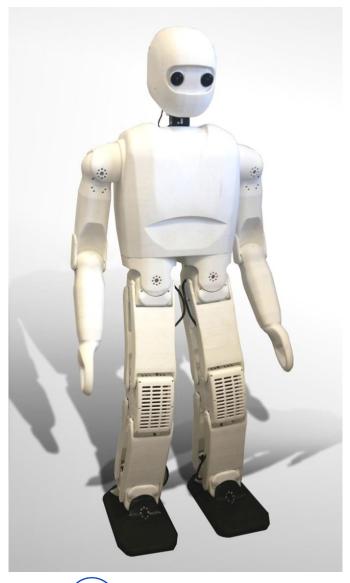




#### **AlexNet Robot Recognition**



#### NimbRo-OP2



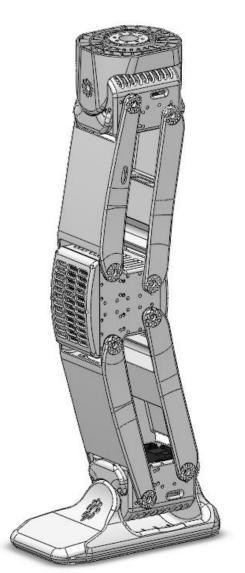
- Height: 135 cm
- Weight: 18.0 kg
- SLS printed exoskeleton
- 18 Degrees of Freedom
  - 5 DoF per leg
    - Parallel kinematics
    - 13 MX-106 actuators
    - Additional spur gears
  - 3 DoF per arm
  - 2 DoF in the neck
- SFF PC i7-7567U 3.5 GHz
- Wide-angle camera
- CM-740 with IMU
- LiPo 14.8 V, 6.6 Ah battery

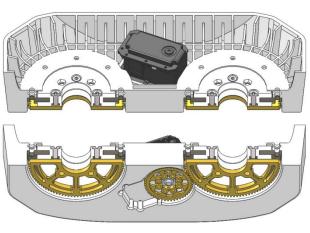
[Ficht et al. Humanoids 2017]

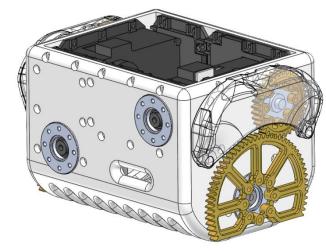


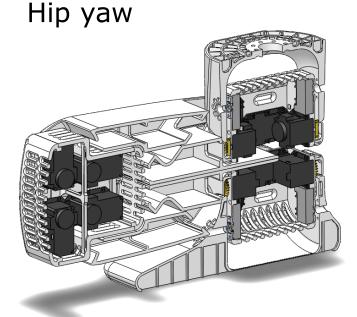
#### **Construction Details**

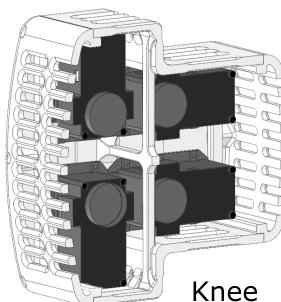
#### Hip pitch & roll













[Ficht et al. Humanoids 2017]

## RoboCup 2017 AdultSize Final



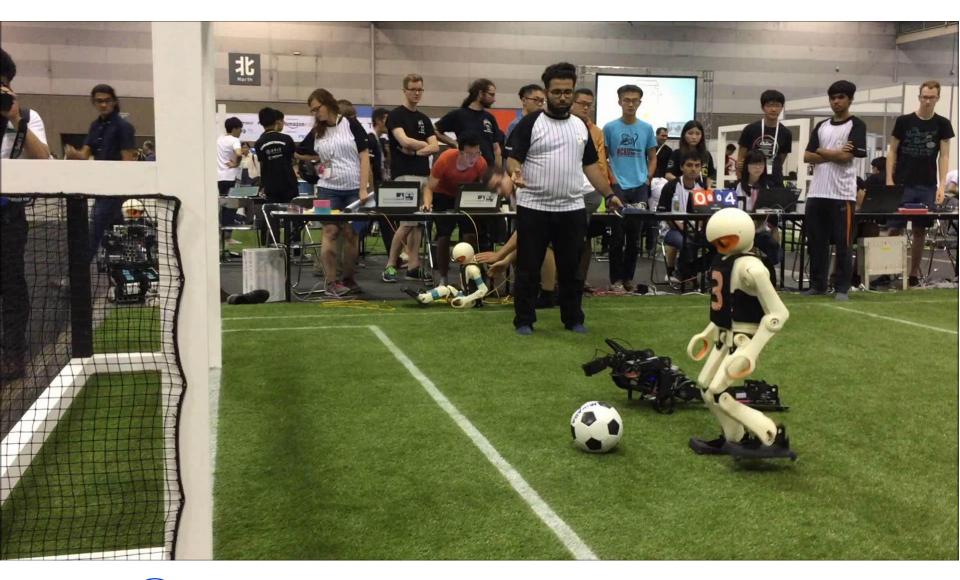


# RoboCup 2017 AdultSize Technical Challenge





## RoboCup 2017 TeenSize





## NimbRo RoboCup 2017 Trophies





#### **Conclusions**

- Capable robots for Humanoid TeenSize and AdultSize class
- Hard- and Software released
- Many challenges
  - Articulated perception
  - Dynamic full-body motions







ais.uni-bonn.de/nimbro/OP

## **Questions?**

