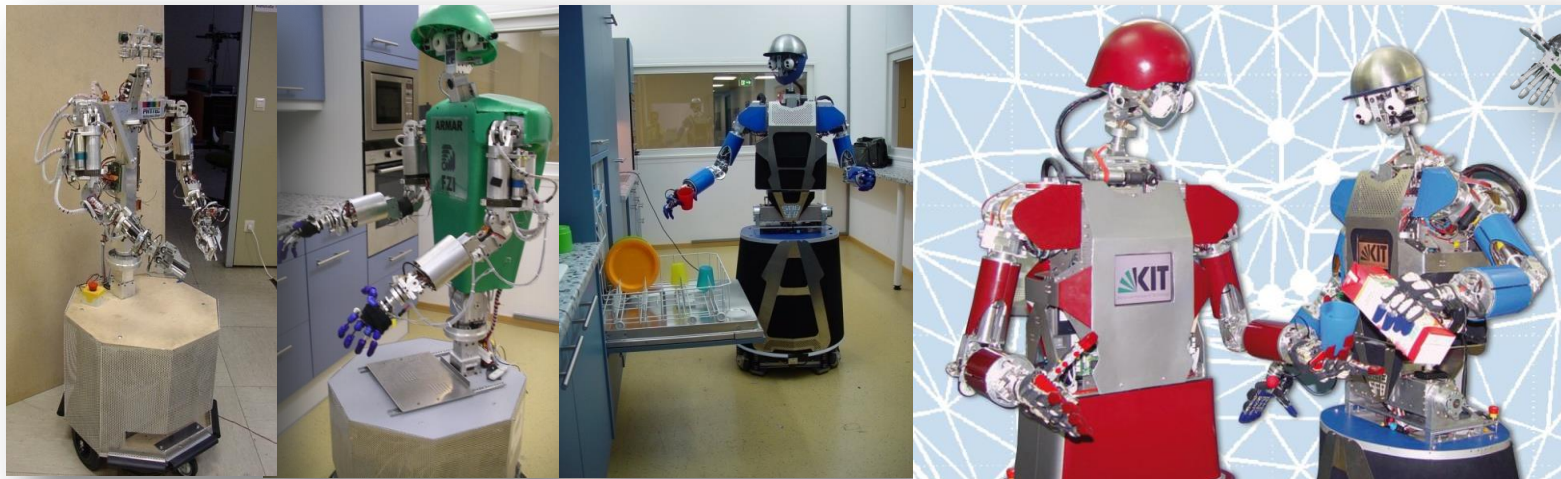


Humanoid Robotics Research @ KIT

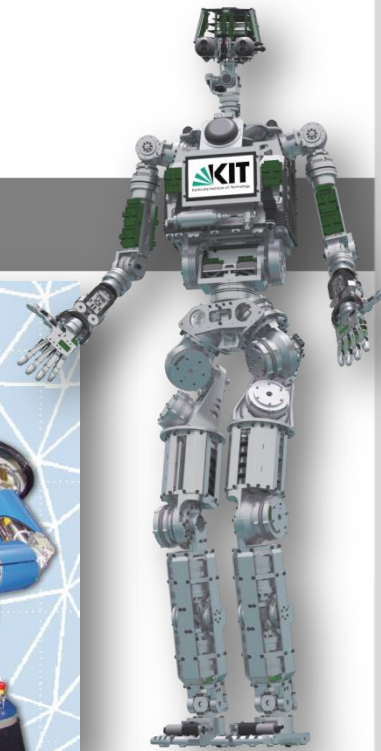
Tamim Asfour

Institute for Anthropomatics and Robotics, Humanoid Robotic Systems



<http://www.humanoid.kit.edu>

<http://h2t.anthropomatik.kit.edu>



KIT – Our Mission



Research &
Development



Higher Education



Innovation

„... to form a novel Quality of Cooperation, and to overcome the Separation between Federal Research Facilities and State Universities ...“

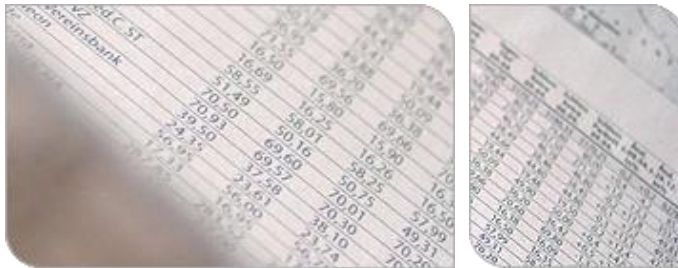
KIT in Figures

Employees
9.261

Students
23.836

364
Professors

789
Annual Budget in Million Euros



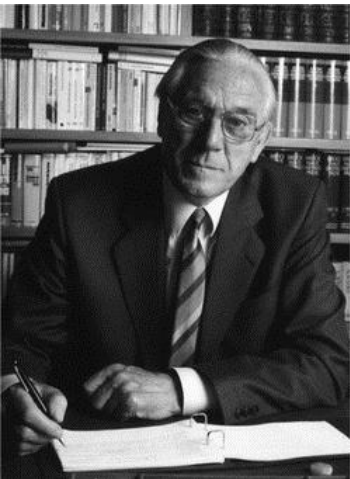
Engineering Tradition



Karl Benz: inventor of the modern automobile

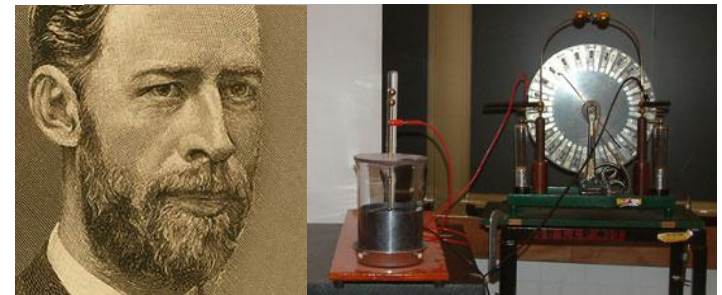


Ferdinand Braun: inventor of cathode ray tube → television



Karl Steinbuch: coined the term *Informatics* in 1957

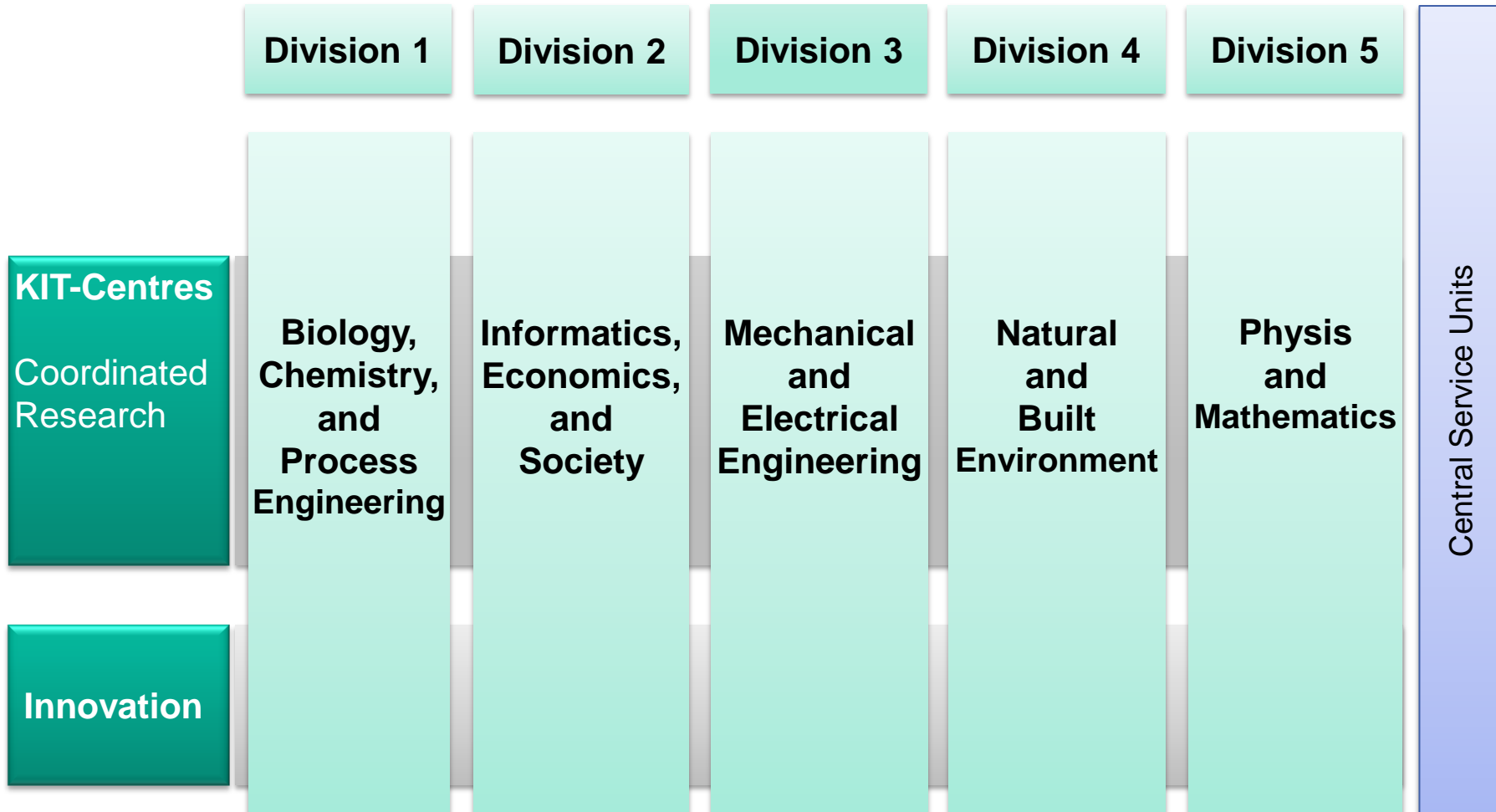
Heinrich Hertz: confirmation of electromagnetic waves



Fritz Haber: fixation of atmospheric N_2 → synthetic ammonia

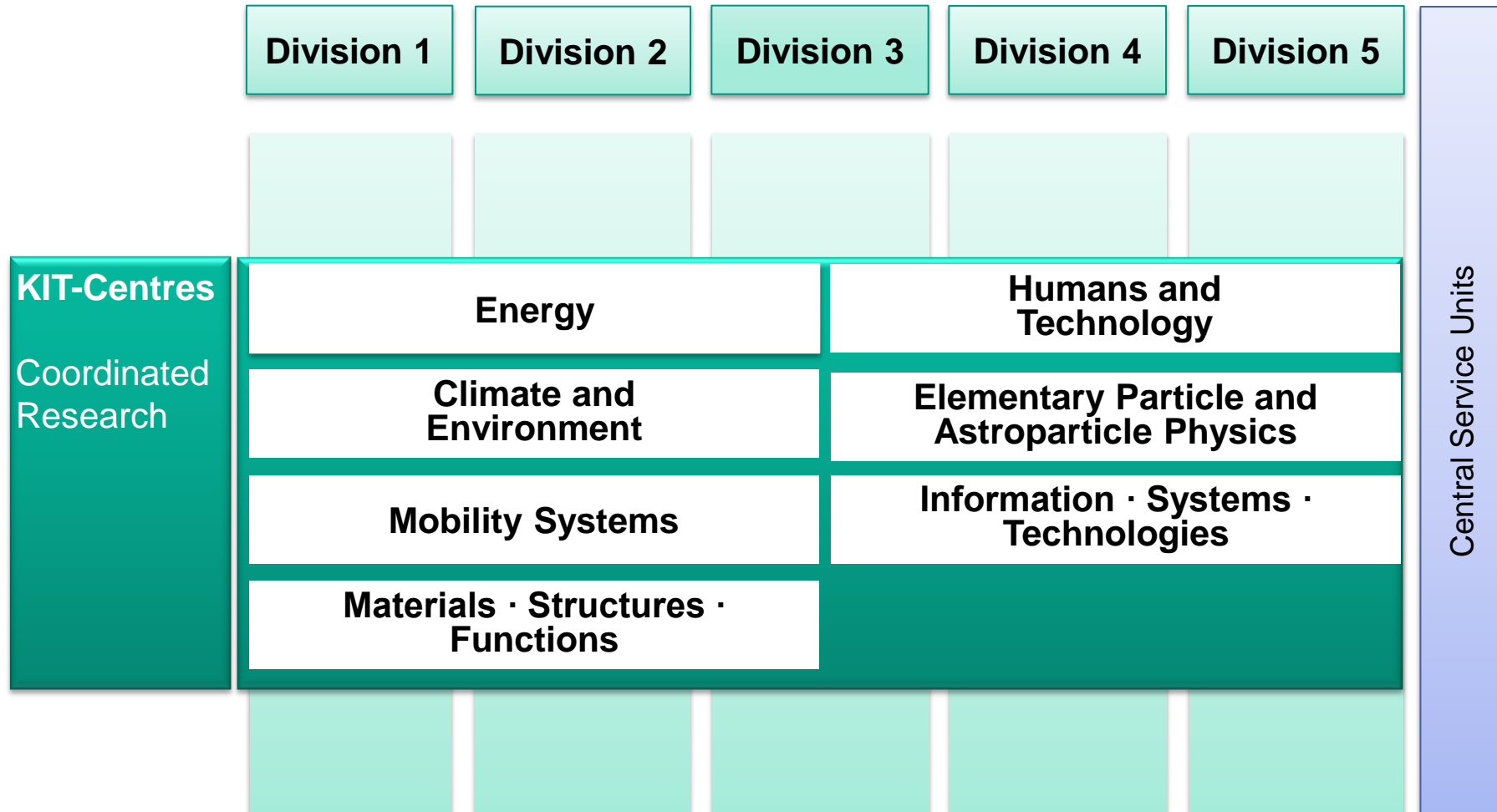
Disciplines organized in Divisions

Executive Board



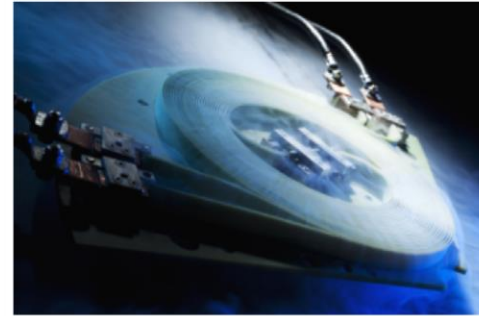
Coordinated Research in KIT-Centres

Executive Board



Energy

7 Topics
1250 Empl.



Climate and
Environment

7 Topics
500 Empl.



Mobility Systems

7 Topics
900 Empl.



Materials · Structures ·
Functions

14 Topics
1100 Empl.



Humans and
Technology

6 Topics
400 Empl.



Elementary Particle and
Astroparticle Physics

9 Topics
400 Empl.

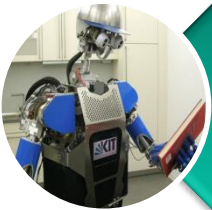


Information · Systems ·
Technologies

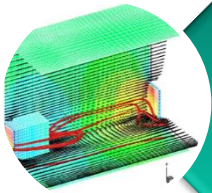
10 Topics
450 Empl.



Networked complex adaptive
Systems (COMMputation)



Robotics, Anthropomatics,
Mechano-Informatics



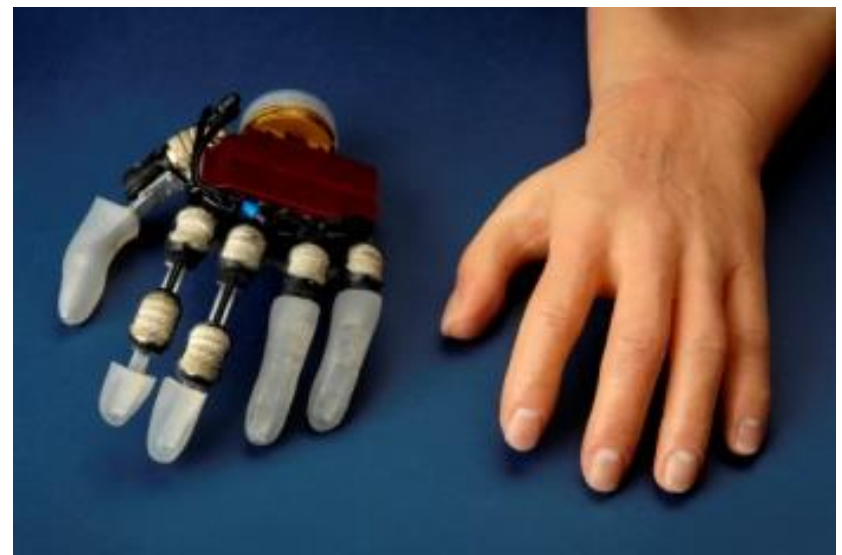
Scientific Computing and
Data-intensive Research



Dependable and Trusted
Information and Communication

Anthropomatics is ...

... the science of the symbioses between
human and machine



Institute for Anthropomatics and Robotics (IAR)

10 chairs, ca. 150 researchers

• **High Performance Humanoid Technologies**

Asfour



• **Vision and Fusion**

Beyerer



• **Humanoids and Intelligence Systems**

Dillmann



• **Intelligent-Sensor-Actuator System**

Hanebeck



• **Intelligent Industrial Robotics**

Hein



• **Cognitive Systems**

Schultz



• **Computer Vision for Human Computer Interaction**

Stiefelhagen



• **Interactive Systems**

Waibel



• **Process Control, Automation and Robotics**

Wörn



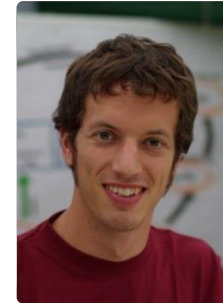
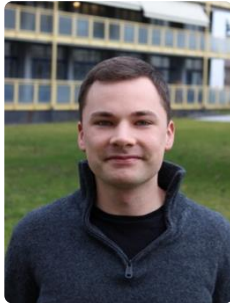
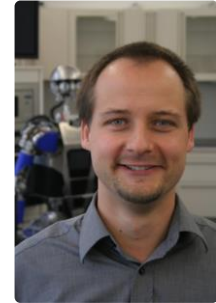
• **Applied Technical Cognitive Systems**

Zöllner

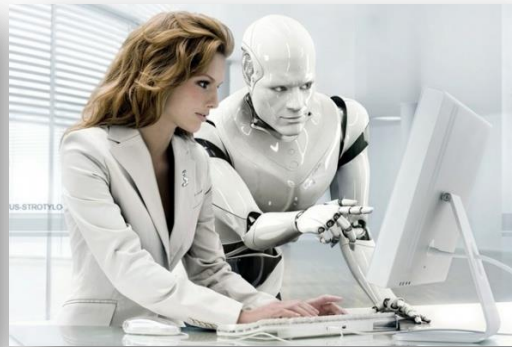


My team

Humanoids@KIT



Building Humanoids = Building Human-Centered Technologies



- Assistants and companions for people in different ages, situations, and environments to improve the quality of life
- Key technologies for future robotic systems
- Experimental platforms to study theories about humans from other disciplines

My inspiration

■ Biology

■ Science Fiction

Human performance



Roger Federer

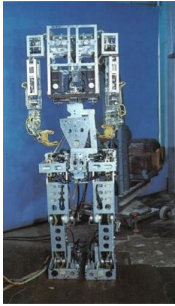


Human performance

- Johanna Quaas - oldest active Gymnast of the World!
86 years, from Halle, Germany



Humanoid robotics has made progress !



WABOT-1



P2



ASIMO



DB



CB



HRP-2



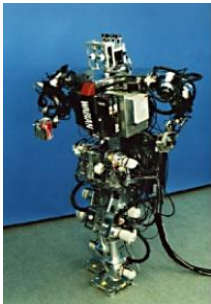
HRP-4



HRP-4C



ARMAR-IV



WABIAN



Twendy-one



ARMAR-III



iCub



kojiro



Partner Robot



HUBO



KOBIAN



Petman



Robonaut



Justin



NAO

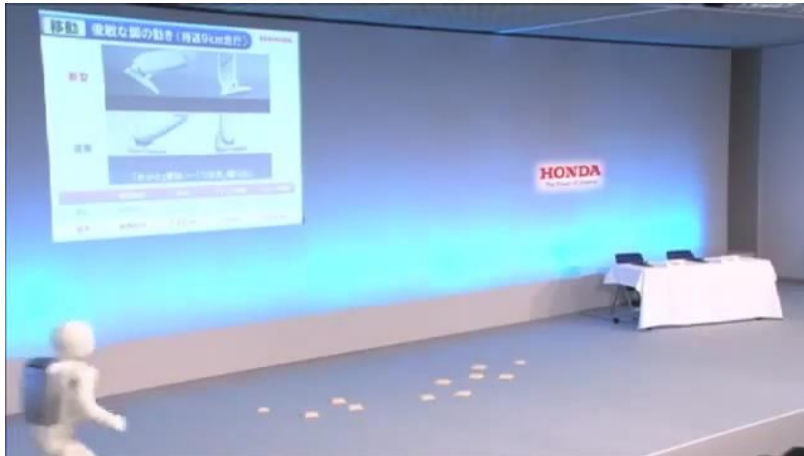


DARWIN-OP



Lola

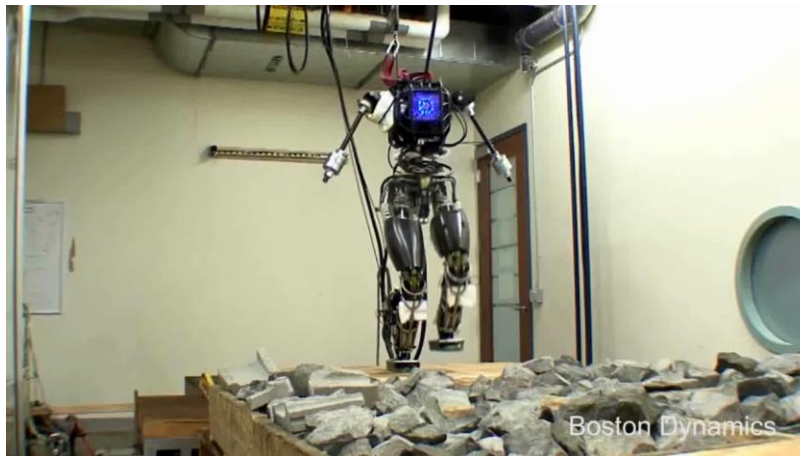
Some examples



ASIMO, Honda, Japan



HRP-4C, AIST, Tsukuba, Japan



Atlas, Boston Dynamics, USA



ARMAR, KIT, Germany

Ambitious goals have been set for humanoid robotics

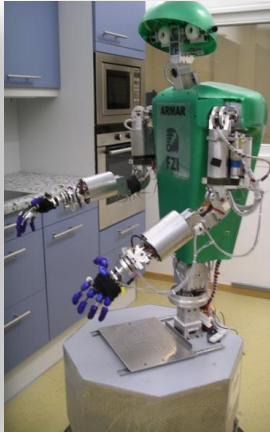
- Companions and assistants for humans in daily life
- Helpers in man-made and natural disasters
- Winners against the winner of most recent World Cup in 2050
- DARPA Robotics Challenge



Humanoid Robots @ KIT



ARMAR, 2000



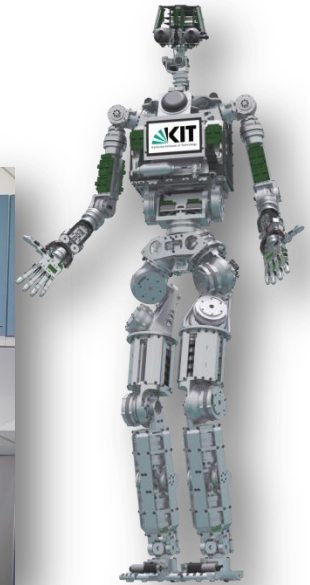
ARMAR-II, 2002



ARMAR-IIIa, 2006



ARMAR-IIIb, 2008

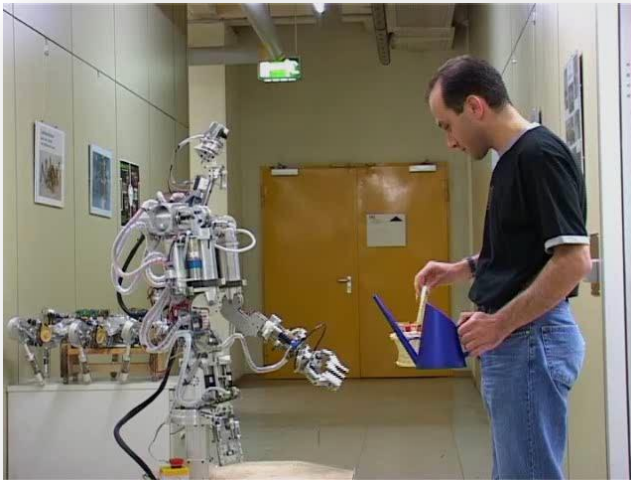


ARMAR-IV, 2011

■ Collaborative Research Center 588: Humanoid Robots - Learning and Cooperating Multimodal Robots (SFB 588)

- Funded by the German Research Foundation (DFG: Deutsche Forschungsgemeinschaft)
- 2001 – 2012
- <http://www.sfb588.uni-karlsruhe.de/>

ARMAR-I and ARMAR-II



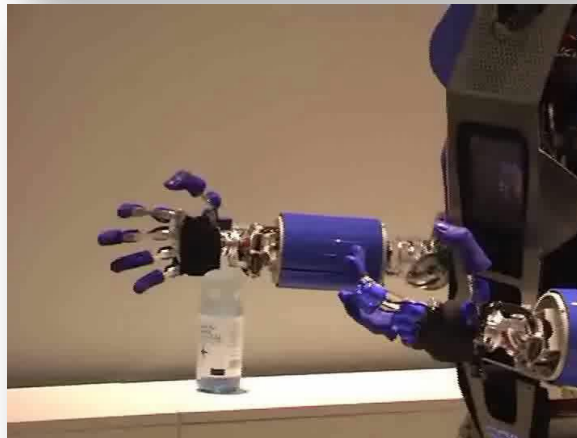
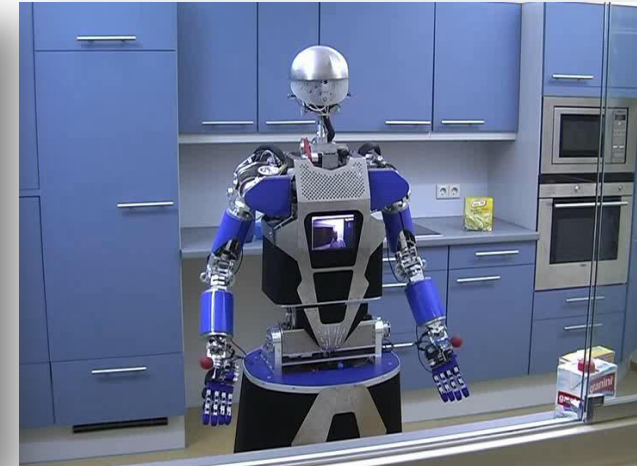
First demonstrator of the SFB 588



Demo at CEBIT 2006

ARMAR-IIIa and ARMAR-IIIb

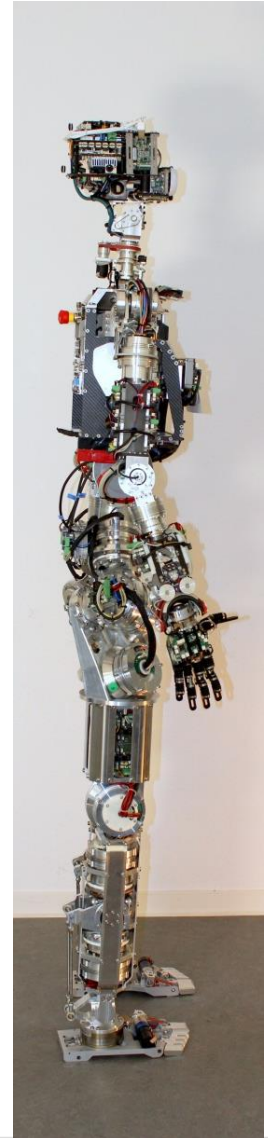
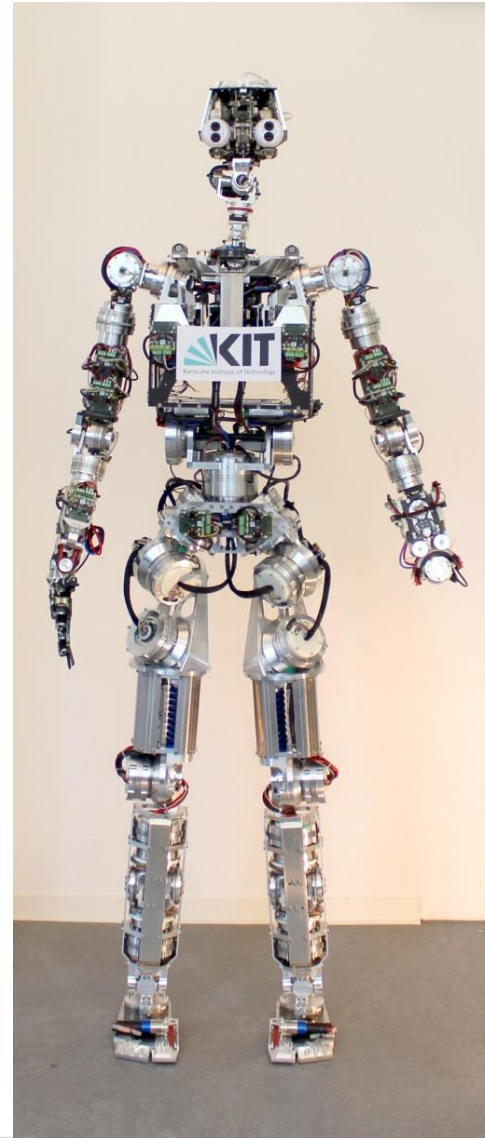
- 7 DOF head with foveated vision
 - 2 cameras in each eye
 - 6 microphones
- 7-DOF arms
 - Position, velocity and torque sensors
 - 6D FT-Sensors
 - Sensitive Skin
- 8-DOF Hands
 - Pneumatic actuators
 - Weight 250g
 - Holding force 2,5 kg
- 3 DOF torso
 - 2 Embedded PCs
 - 10 DSP/FPGA Units
- Holonomic mobile platform
 - 3 laser scanner
 - 3 Embedded PCs
 - 2 Batteries
- Weight: 150 kg



Fully integrated humanoid system

ARMAR-IV

- 63 DOF
- 170 cm
- 70 kg
- Torque-controlled!



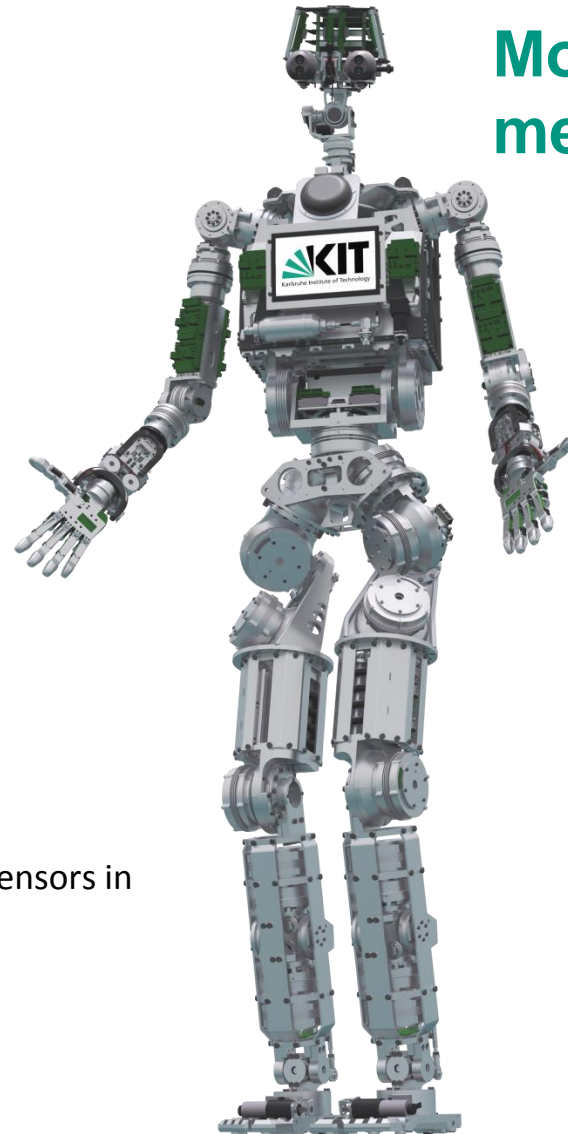
ARMAR-IV: Mechano-Informatics

- Torque controlled
- 3 on-board embedded PCs
- 76 Microcontroller
- 6 CAN Buses

- 63 DOF
 - 41 electrically-driven
 - 22 pneumatically-driven (Hand)

- 238 Sensors
 - 4 Cameras
 - 6 Microphones
 - 4 6D-force-torque sensors
 - 2 IMUs
 - 128 position (incremental and absolute), torque and temperature sensors in arm, leg and hip joints
 - 18 position (incremental and absolute) sensors in head joints
 - 14 load cells in the feet
 - 22 encoders in hand joints
 - 20 pressure sensors in hand actuators
 - ...

More than
mechatronics



ARMAR-IV

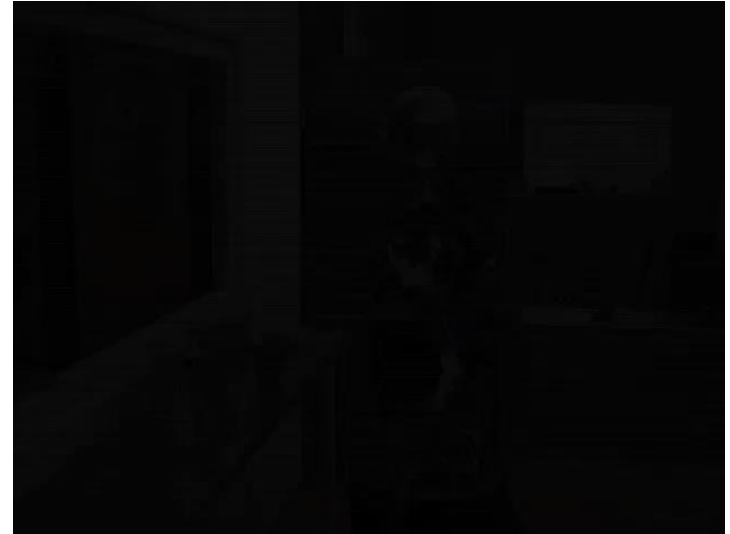
made@KIT

70 kg

170 cm

Humanoids in the real world

- Grasping and manipulation

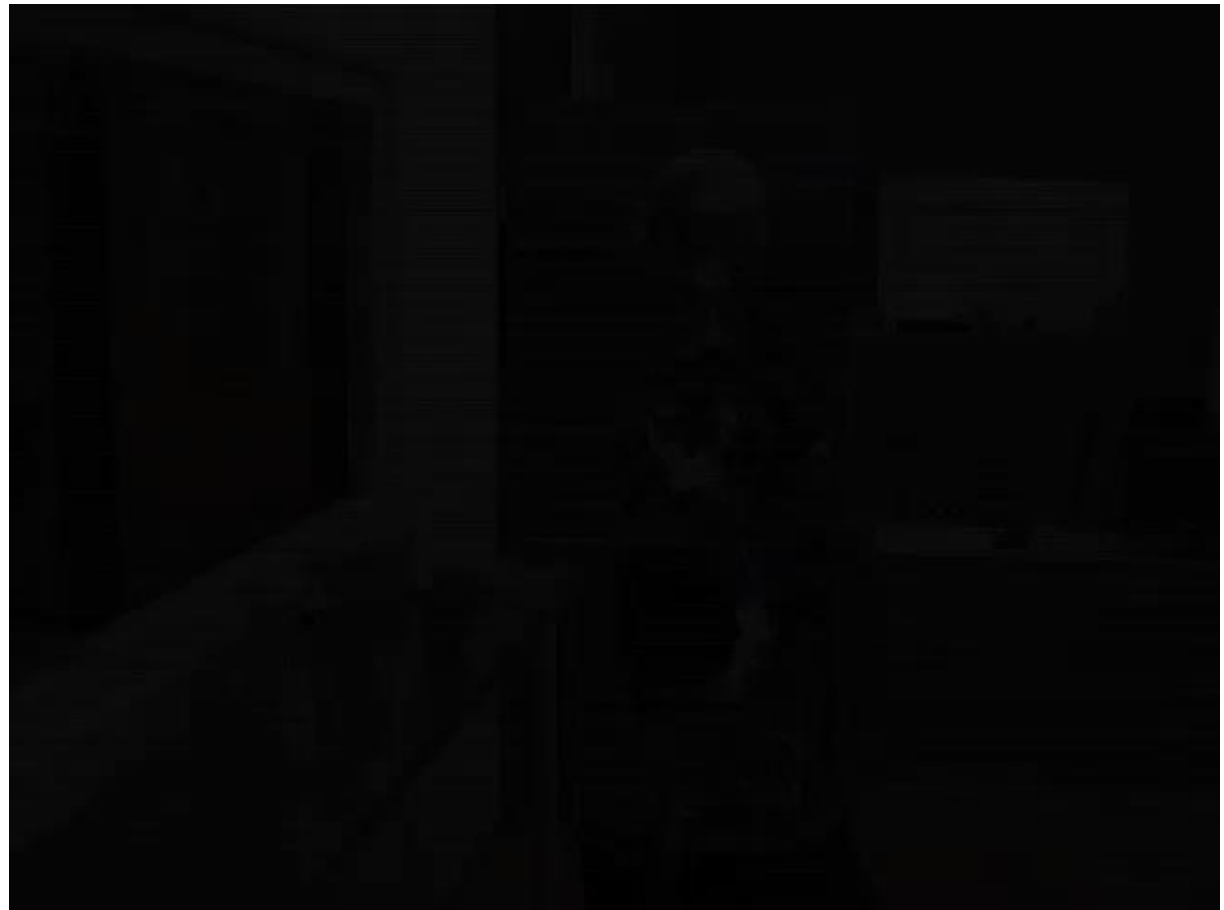


- Learning for human observation



ARMAR-III in the RoboKITchen

- Object recognition and localization
- Vision-based grasping
- Hybrid position/force control
- Combining force and vision for opening and closing door tasks
- Collision-free navigation
- Vision-based self-localisation
- Multimodal human-robot dialogs
- Continuous speech recognition
- Learning new objects, persons and words
- Audio-visual tracking and localization
- ...

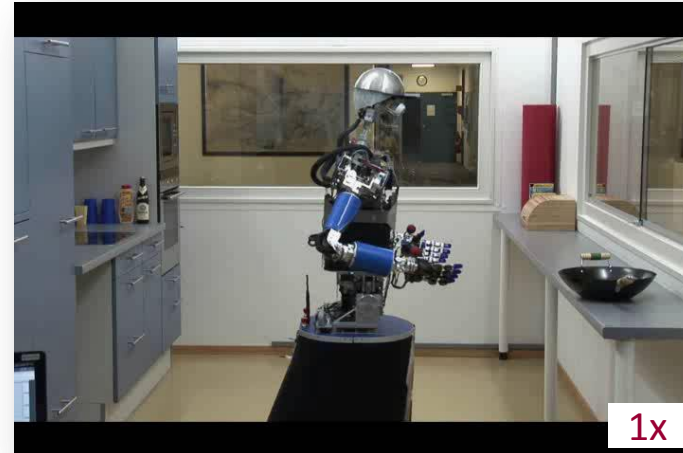
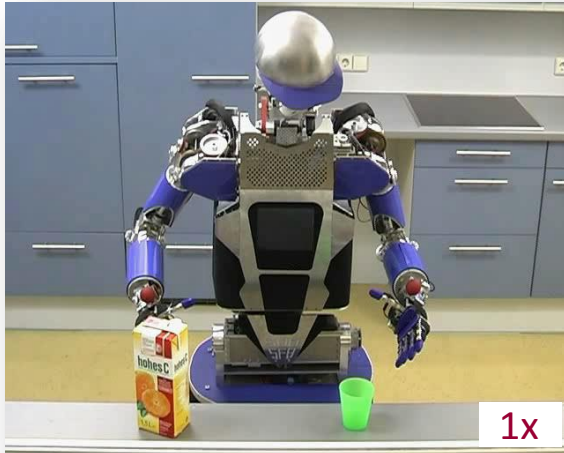


ARMAR-III in the RoboKITchen

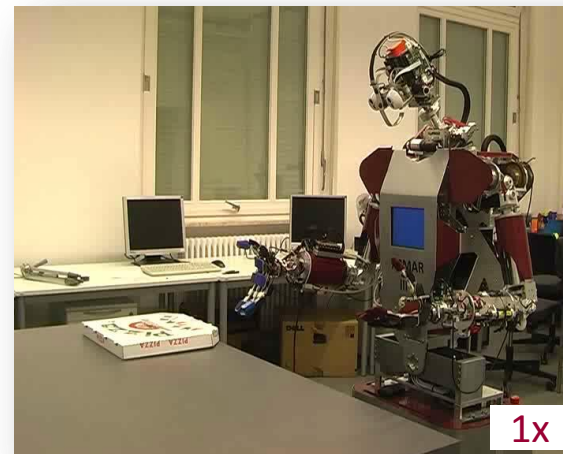
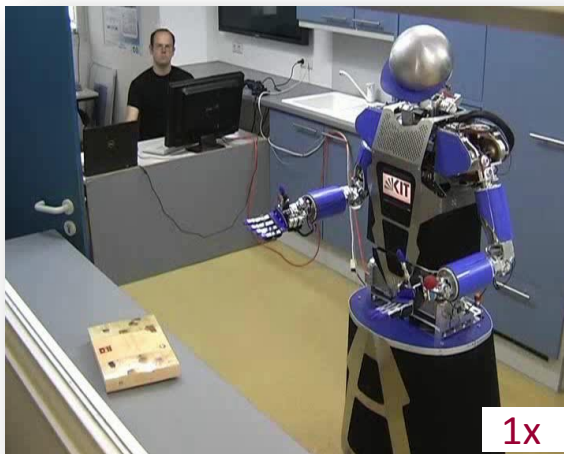
- First step towards 24/7
 - 45 minutes demonstration
 - Shown more than 950 times, since 03. February 2008, to experts and public
 - 75 times in 5 days for approx. 5000 visitors at CeBIT 2012
 - 30 times during the EFFEKTE weekend, 2013 in Karlsruhe

Advanced grasping capabilities

■ Bimanual grasping and manipulation



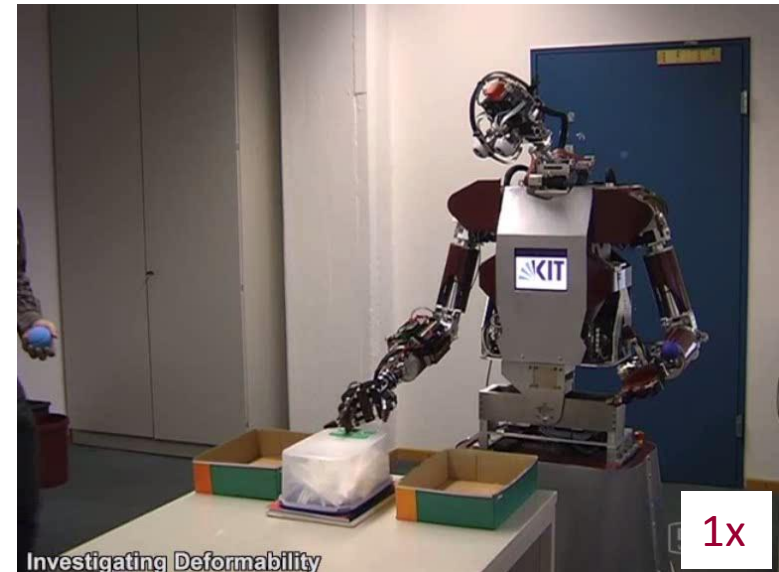
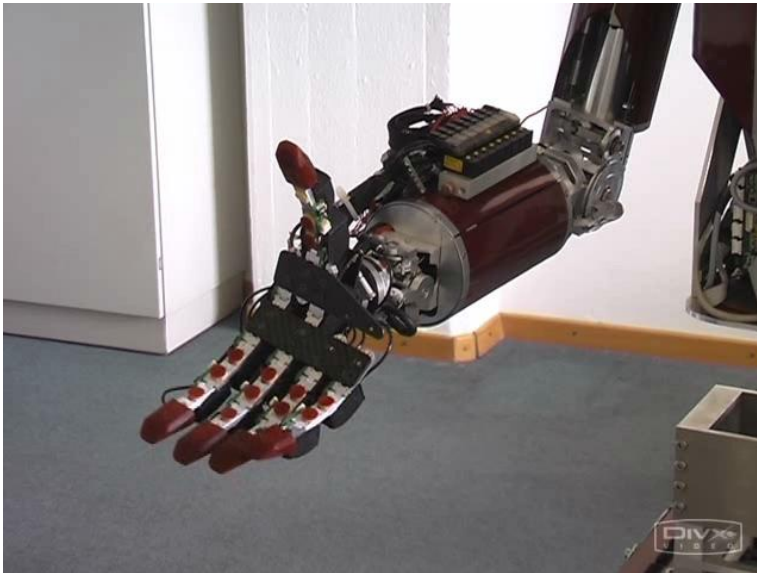
■ Pre-grasp manipulation



*Humanoids 2009
Humanoids 2010
IROS 2011
RAS 2012
RAS 2008*

Haptic exploration of unknown objects

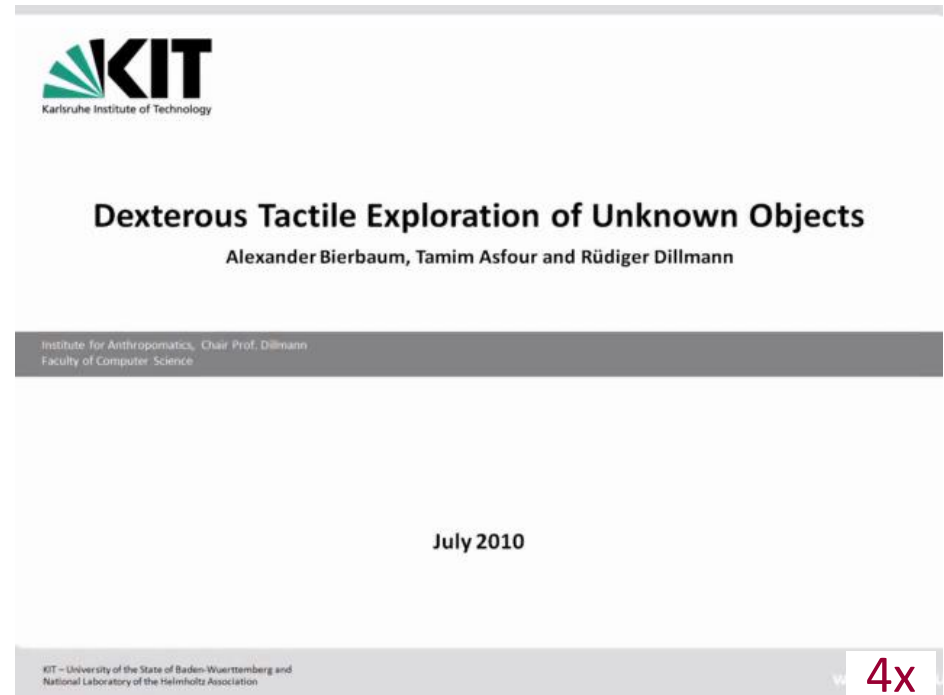
- Multisensory (contact, pressure, force, proprioception) approach for
 - Detection of contact and “objectness”
 - Assessment of object deformability
 - ...



Humanoids 2008, 2009
BioRob 2012

Haptic object exploration

- Potential field guides the robot hand along the object surface
- Oriented 3D point cloud from contact data
- Compute face pairings from 3D point
- Calculate grasping hypotheses using a geometric feature filter pipeline

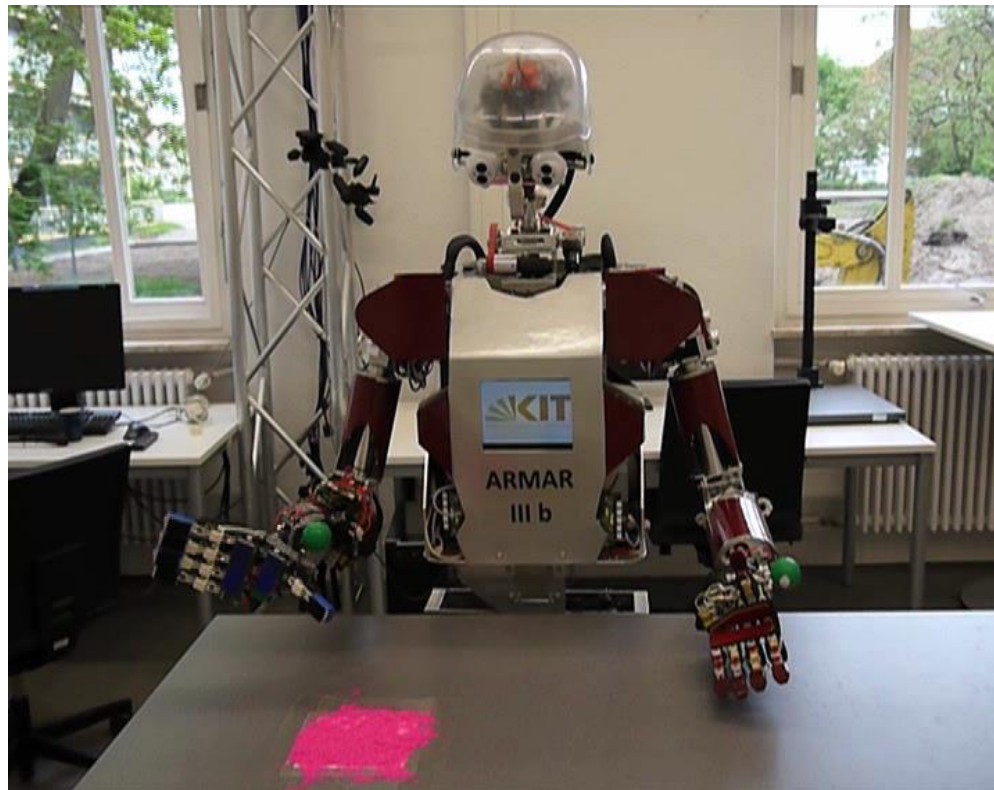


Humanoids 2008, 2009, BioRob 2012

Association between “objects” and grasping actions → “grasp affordances”

Learning object-action-effect correlations

- Objects and Actions are inseparably intertwined
→ Object-Action Complexes (OACs)



Discover, segment and grasp unknown Objects

- Physical interaction (pushing) to separate unknown object from unknown background
- Reliable, correct and complete object segmentation
- Reactive grasping based on haptic feedback:
 - No object model needed
 - No grasping planning

*ICRA 2012, Humanoids 2011, 2012
Adaptive Behavior 2013*

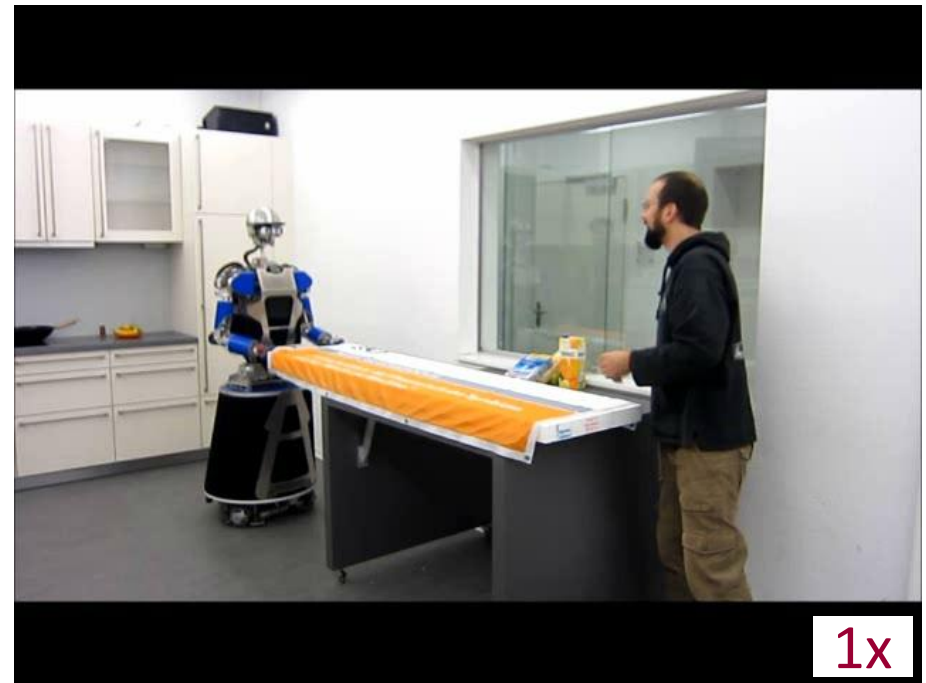
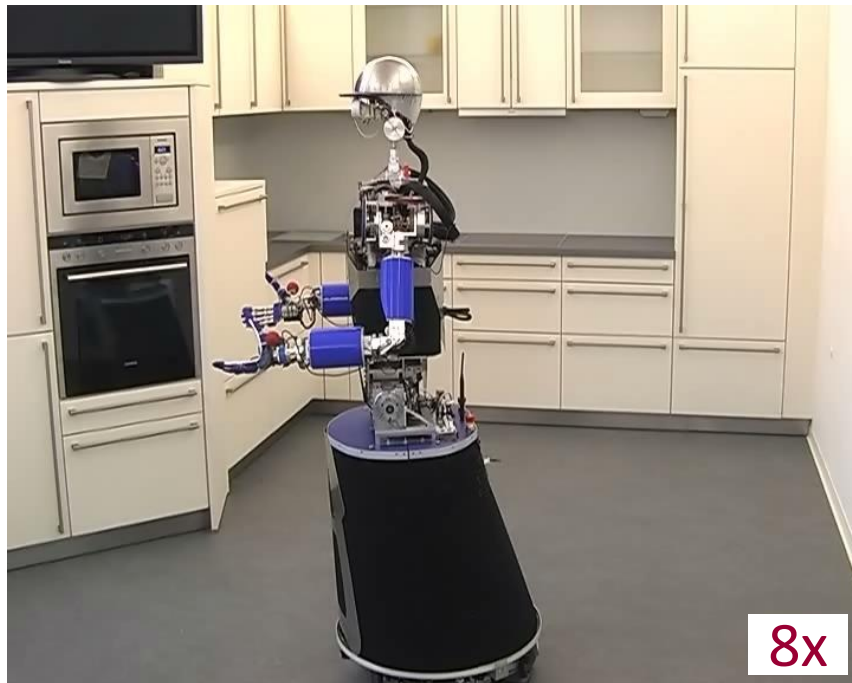


Discovery, Segmentation and Reactive Grasping of Unknown Objects

David Schiebener, Julian Schill and Tamim Asfour

Karlsruhe Institute of Technology
Institute for Anthropomatics
High-Performance Humanoid Technologies

Physical human-robot interaction



Predicting human motion “intention” based on force feedback

Learning from Observation

- Building a library of motion primitives
- Dynamic movement primitives for discrete and periodic movements



Humanoids 2006, IJHR 2008, Humanoids 2007, ICRA 2009, Humanoids 2009, TRO 2010, Humanoids 2012



On discrete and periodic motion primitives

■ Observation: **periodic motions start with a discrete part**

- Stirring – first move the hand to the vessel containing the liquid.
- Wiping – first move the hand to the surface to be wiped.
- Peeling – first bring the peeler to the potato.
- Cutting – first move the hand to the object.
- Walking – first step vs. all other steps.
- Juggling – bringing the balls into the air vs. juggling itself.

We call this non-periodic part the transient.

- ## ■ Encode the **periodic movement** and **all corresponding transients** into a **single** dynamical system (DMP)

Humanoids 2012

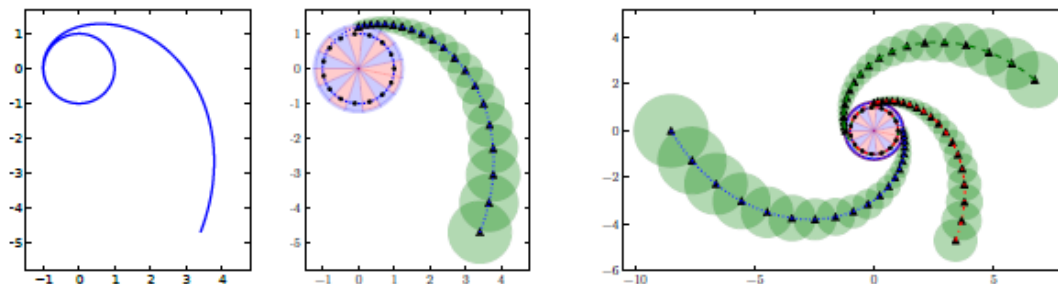
Two dimensional canonical system

■ $s(t) := (\varphi(t), r(t))$ in $\mathbf{R} \times (0, \infty)$ for φ, r solution of

$$\begin{cases} \dot{\phi} = \Omega, \\ \dot{r} = \eta(\mu^\alpha - r^\alpha)r^\beta, \\ \phi(0) = \phi_0, r(0) = r_0 \end{cases}$$

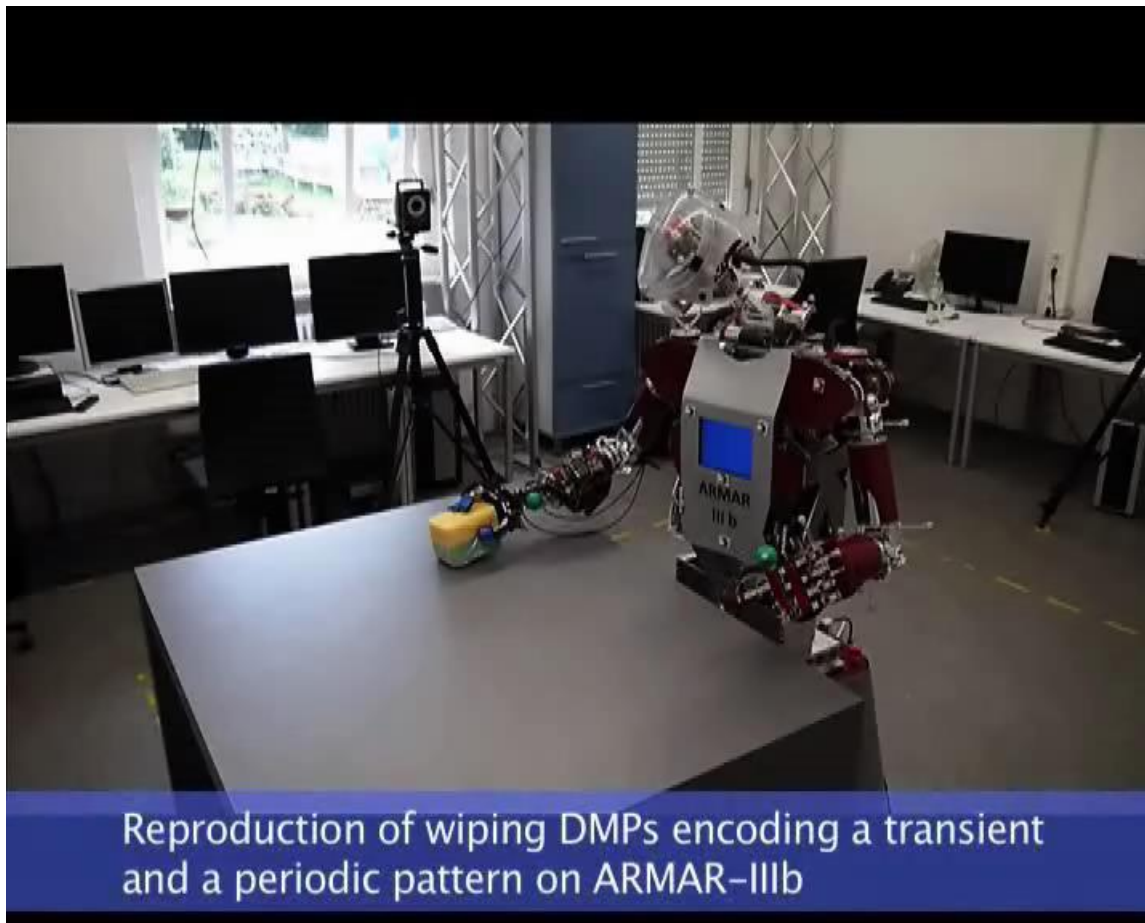
■ ψ_j “living” outside the limit cycle (transient).

■ φ_i “living” on the limit circle (periodic pattern).



■ Each transient has its separate set of basis functions ψ_j .

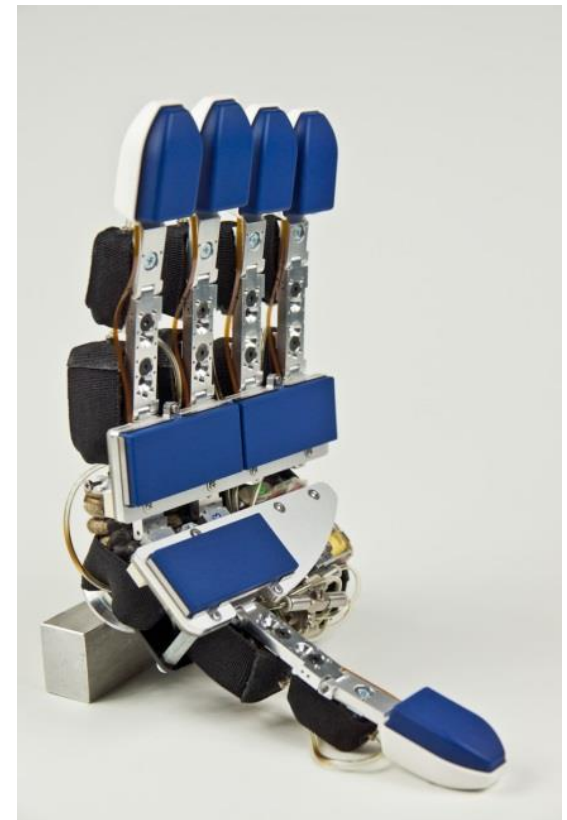
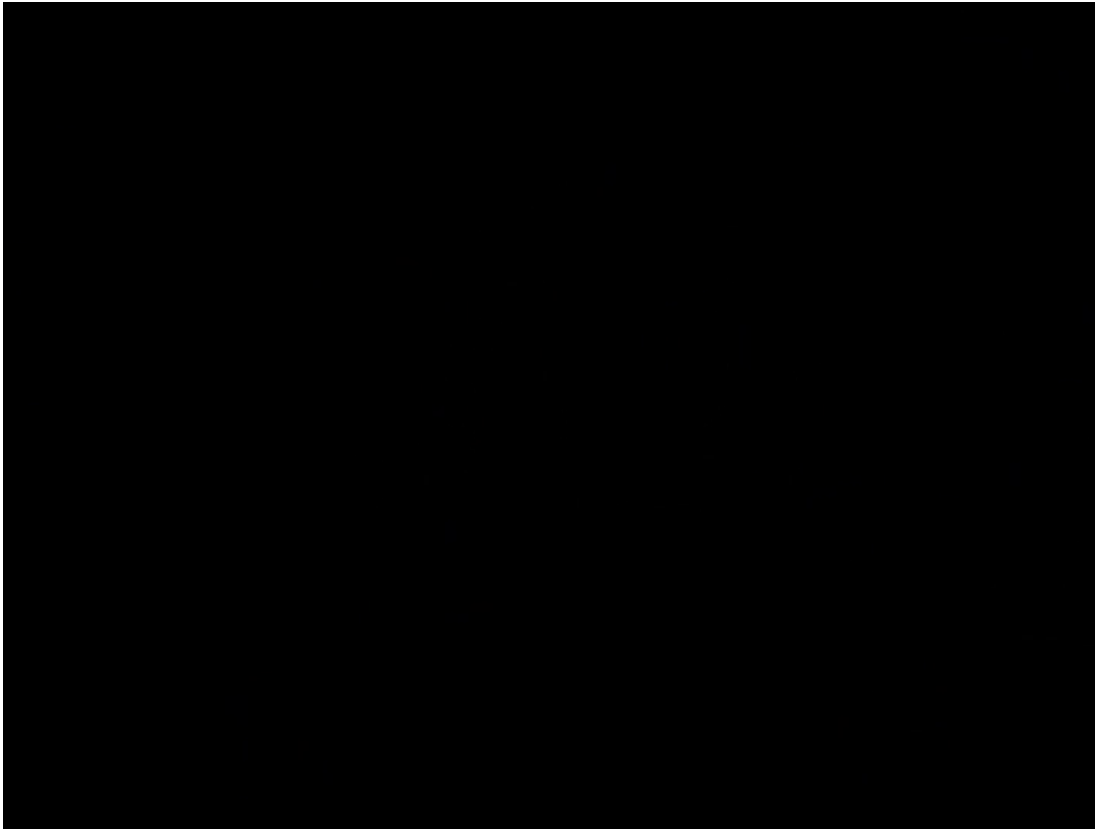
Periodic and transient motions



Humanoids 2012

Learn to wipe

- Learning associations between object properties and action parameter



Learning from Human Observation

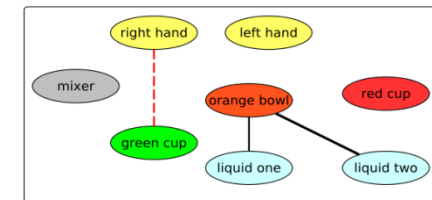


Action Sequence Reproduction

Observation of **complex task** by human demonstration



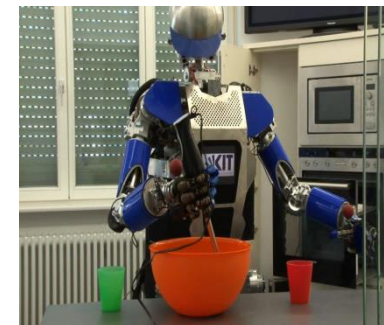
Automatic segmentation of the action sequence by detection of **object relation changes**



Association of action segments with known Object-Action Complexes (OACs)

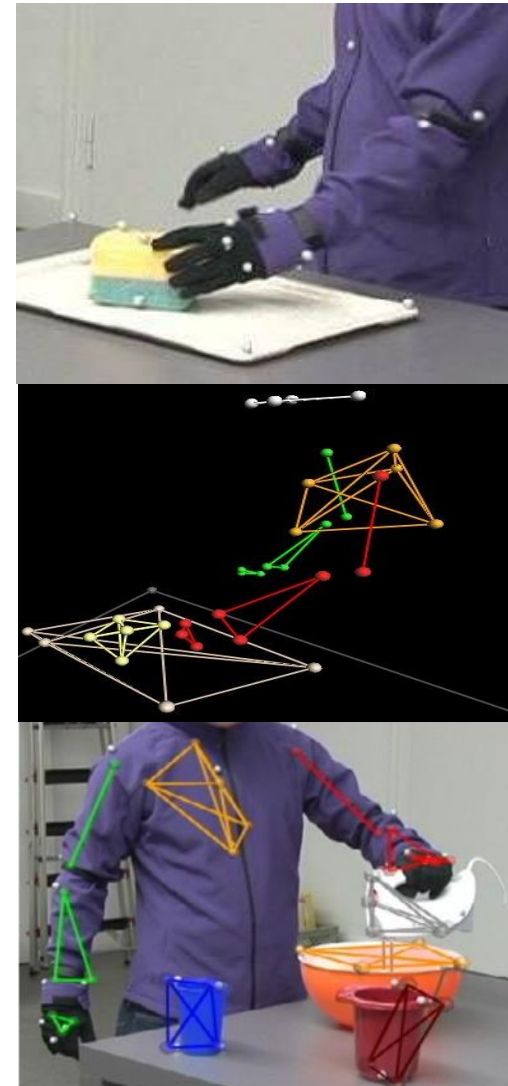


Reproduction of task with **sequence** of OACs



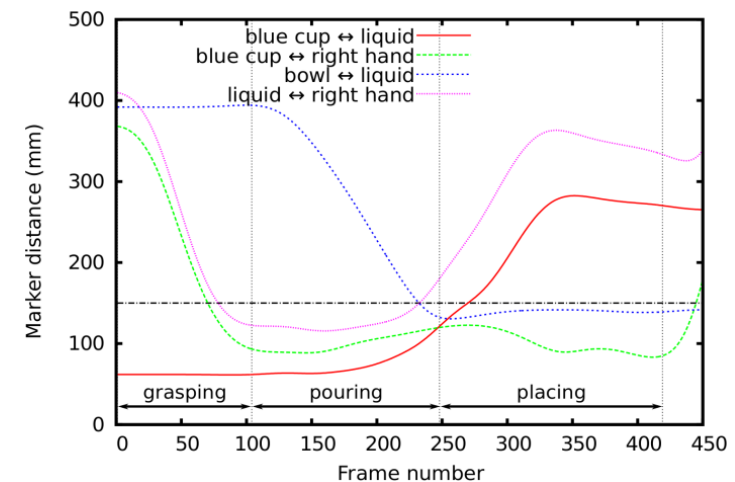
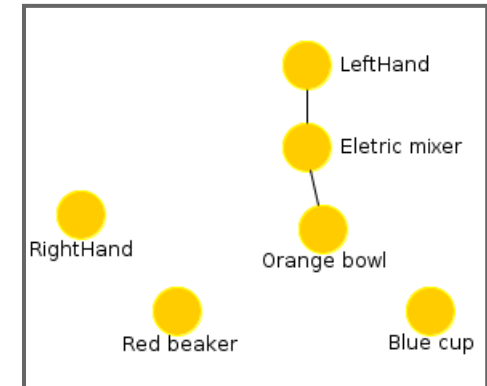
Observation of task demonstration

- **Vicon with 10 cameras**
 - Precision ($\sim 1\text{mm}$)
 - Frame rate (200Hz)
- **Markers** attached to human body and all objects
- Markers **grouped by** human body and all objects
 - **Relative position** to other markers of a group trained and used for marker identification
- Extraction of **3D trajectories** of all markers



Automatic Action Segmentation

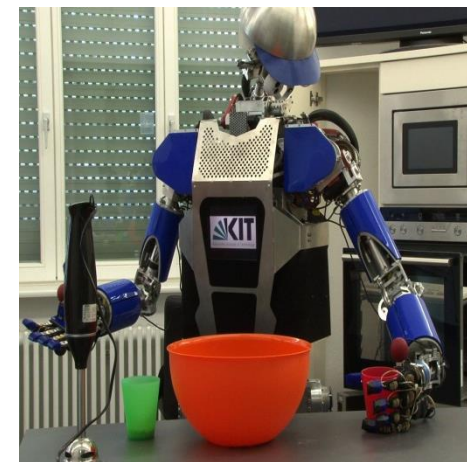
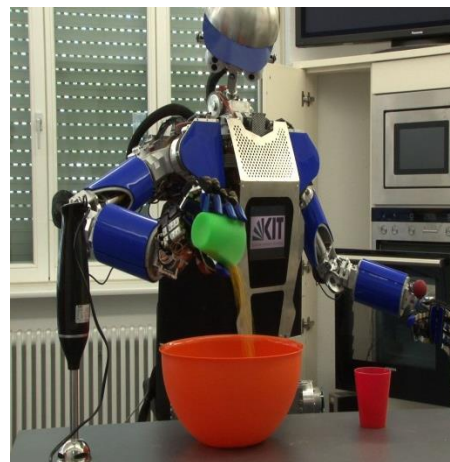
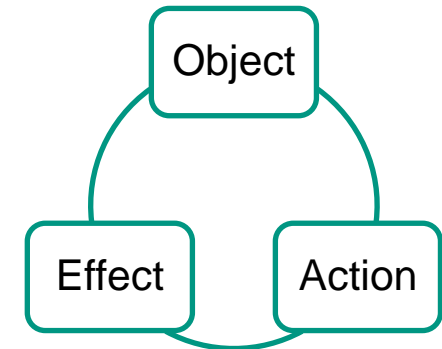
- Segmentation based on detection of **key frames**
- Based on the **Semantic Event Chains** concept (Wörgötter et al.)
- Detection of **key frames** based on **change of marker distance**
- Key frames are determined by **object relation changes**
 - Touching/Non-Touching relations between objects
- **World states** stored for every key frame
 - World state is represented as **object relations**



Object-Action Complex library

- Object-Action Complex (simplified)
 - **Motor actions** with object/agent information
 - Object pos.
 - Force/torque information
 - ...
 - Stored with **preconditions** and **effect** on the world

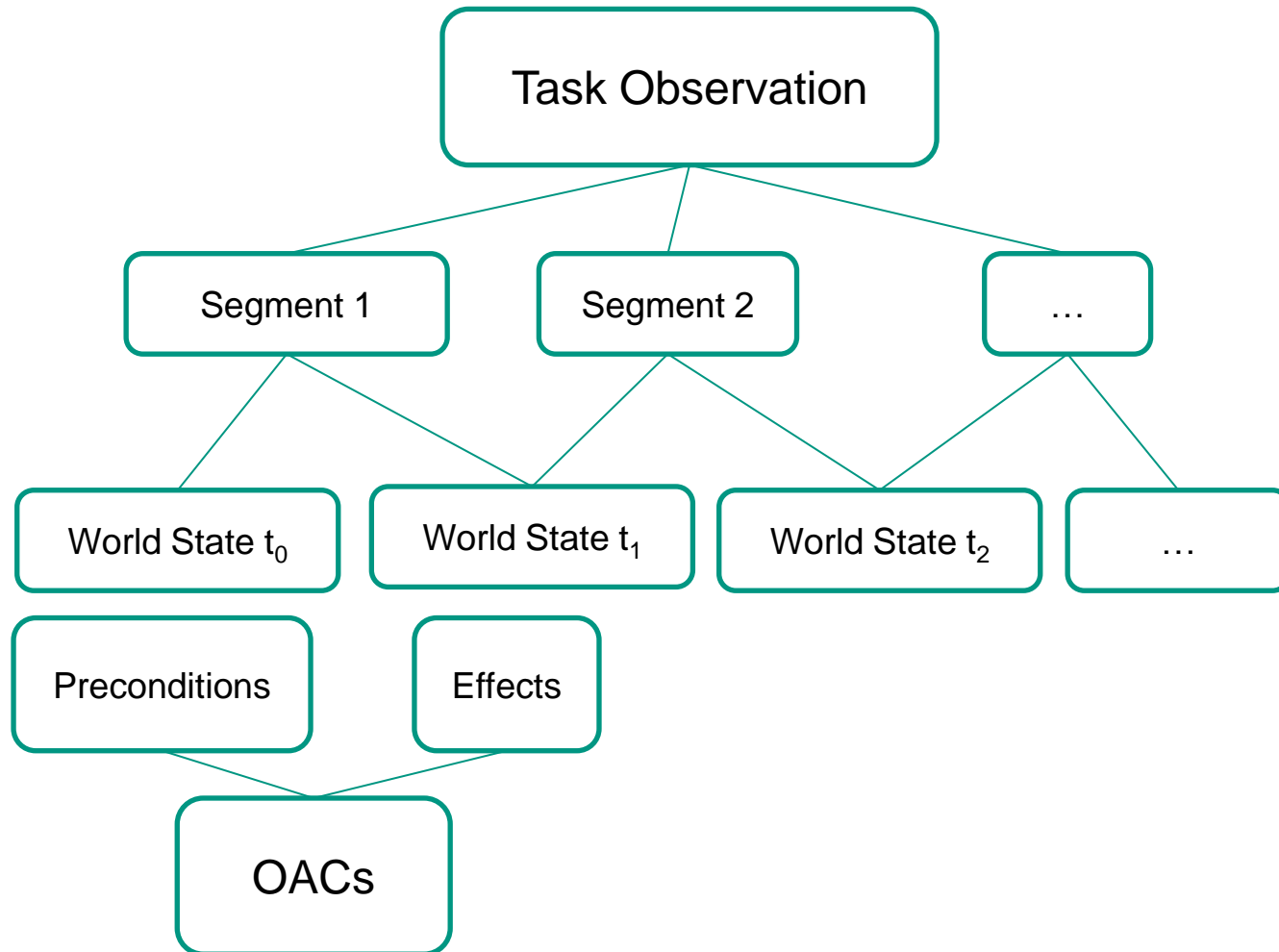
- Basic skills predefined in library:
 - grasping
 - placing
 - pouring
 - ...



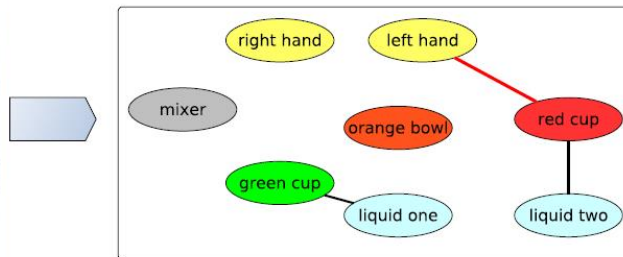
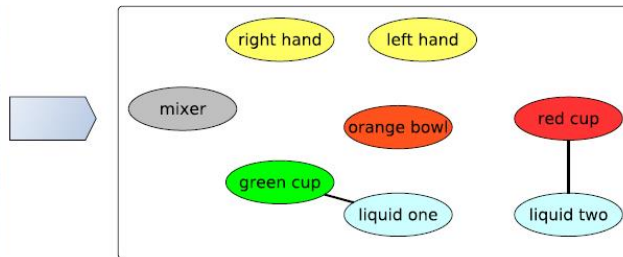
Associating OACs with observed action segments

- OACs have **preconditions** and **effects**
- **Observed segments** are between 2 key frames
 - **Key frames** have the current **world state** attached
- Preconditions and effects of OACs are **subsets** of a world state
- **Matching** preconditions and effects of OACs with observed world states
 - Find an OAC for all observed segments

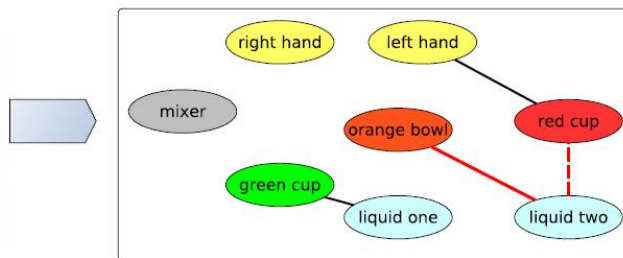
Associating OACs with observed action segments



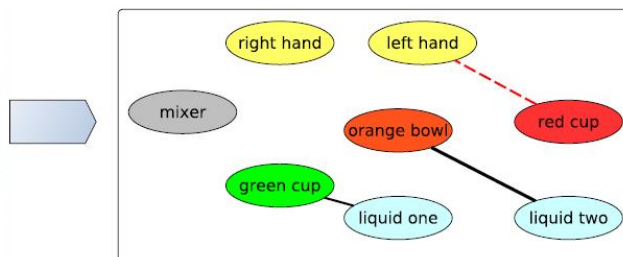
Example sequence of actions



Pre: Left hand <-> Nothing
Grasp
Effect: Left hand <-> Red cup



Pre: Left hand <-> Red cup
Liquid two <-> Red cup
Pour
Effect: Liquid two <-> Orange bowl



Pre: Left hand <-> Red cup
Place
Effect: Left hand <-> Nothing



Learning from observation: OAC replacement

Inference of OACs for Structural Bootstrapping based on Prior Knowledge and Human Demonstration

M. Wächter, M. Do, D. Schiebener and T. Asfour

Master Motor Map (MMM)

Master Motor Map (MMM) – Motivation

- Design of humanoid robots

→ models of body parts are needed

- Various human motion capture systems action recognition systems, imitation systems, visualization modules, and robot systems for reproduction

→ Unified representation is needed!

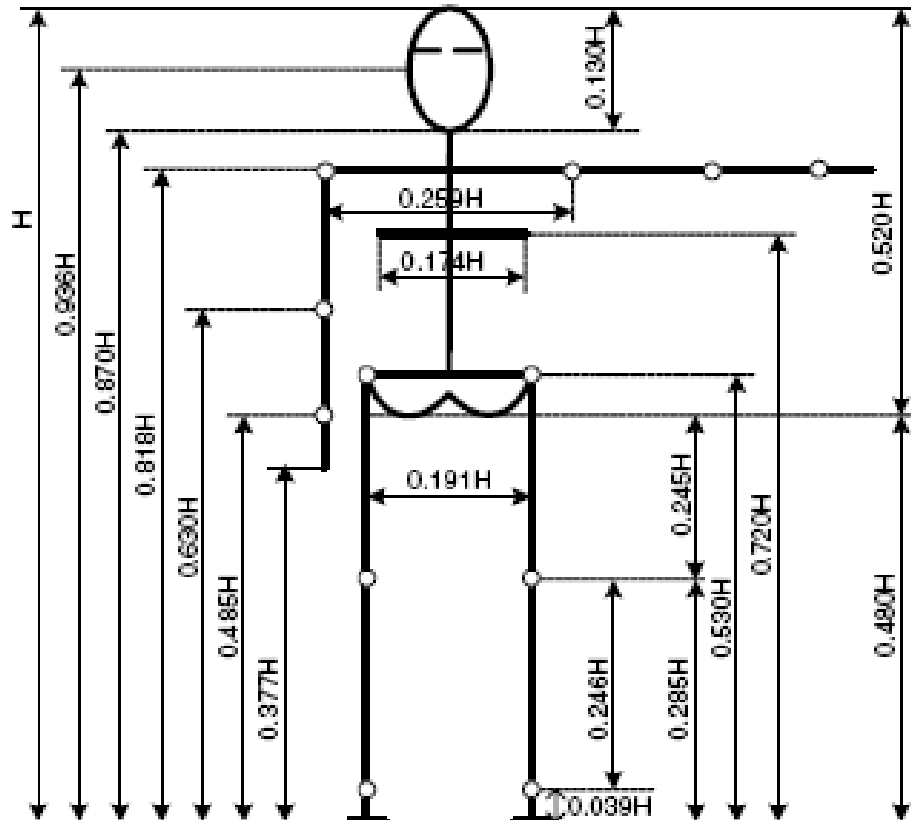
Master Motor Map (MMM)

- Reference model of the human body
 - For humanoid robot design
 - Imitation of human actions
 - Action recognition
 - Visualisation of human movements
- Interfaces and data structures for the transfer of motor knowledge between different embodiments

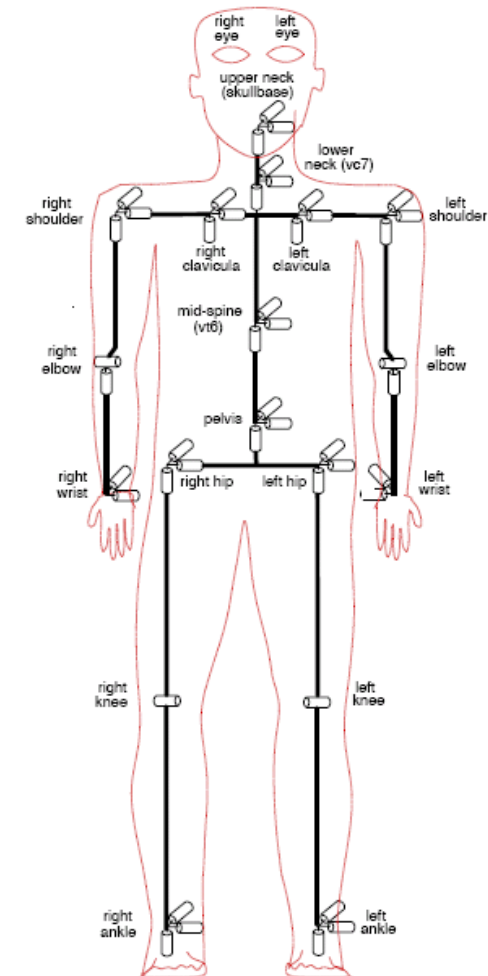
*Pedram Azad, Tamim Asfour and Ruediger Dillmann. **Toward an Unified Representation for Imitation of Human Motion on Humanoids.** IEEE International Conference on Robotics and Automation, 2007*

*S. Gärtner, M. Do, C. Simonidis, T. Asfour, W. Seemann and R. Dillmann, **Generation of Human-like Motion for Humanoid Robots Based on Marker-based Motion Capture Data,** 41th International Symposium on Robotics (ISR), pp. 1 - 8, 2010*

Human body model



D.A. Winter, Biomechanics and Motor Control of Human Movement, John Wiley & Sons Inc. 1990

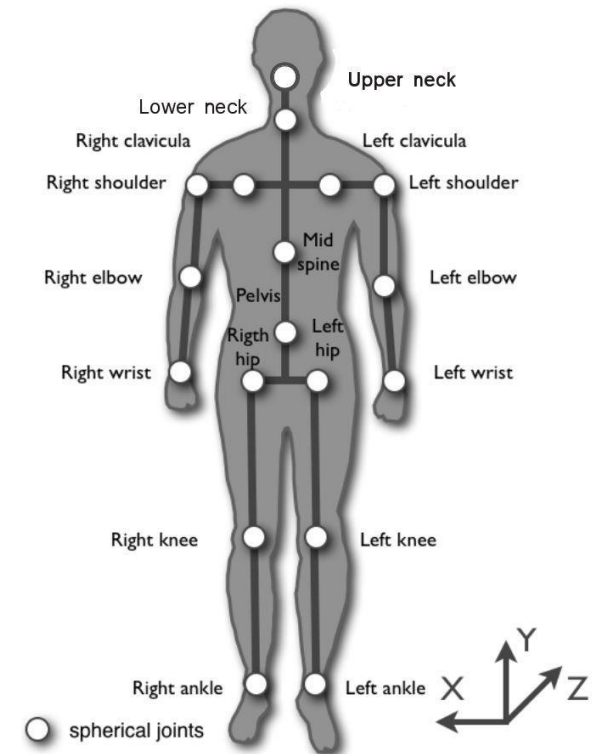


P. Azad, T. Asfour, and R. Dillmann, "Toward an Unified Representation for Imitation of Human Motion on Humanoids," in IEEE International Conference on Robotics and Automation, Rome, Italy, April 2007.

Master Motor Map (MMM)

■ Reference model of the human body

- **Kinematic** model: joints and segment lengths
- **Dynamic** model: segment mass, center of mass and moments of inertia
- **Statistic/anthropomorphic** model: Segment properties (e.g. length, mass etc) defined as a function (regression) of global parameters (e.g. body height, weight)
- **96 DoFs** (44 for the hands)



(ICRA 2007, ISR 2010, Humanoids 2012)

MMM Specification

- Types of Model:
 - Kinematic Model
 - Arrangement of segments and joints
 - Statistic/ Anthropomorphic Model
 - Body Segment Properties (Winter 2005)
 - Dynamic Model
 - Kinematic model enriched with pre-defined segment properties e.g. mass, length etc.

- Each MMM model is defined as XML Schema

- MMM Software <https://i61mmm.ira.uka.de/>

Statistic/ Anthropomorphic Model

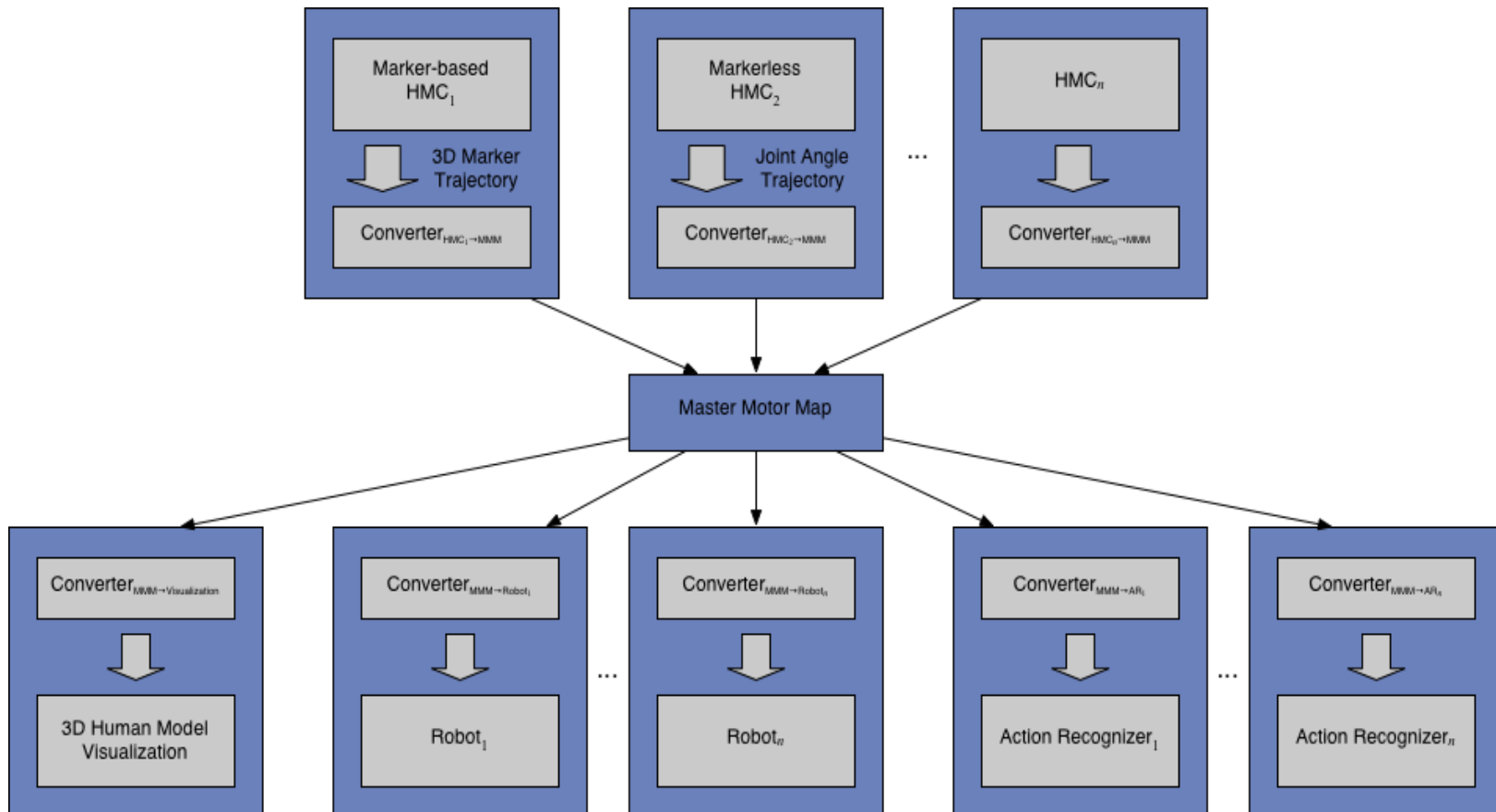
- Body segment properties (e.g. length, mass etc) are defined as a function (regression) of certain global parameters (e.g. body height, weight etc.)
- Models have been discovered and verified by various researchers (see for example de Leva 1996, Winter 2005, Pronost et al., 2006)

D.A. Winter, Biomechanics and Motor Control of Human Movement, John Wiley & Sons Inc. 1990

P. de Leva, "Adjustments to zatsiorsky-seluyanov's segment inertia parameters," J. of Biomechanics, vol. 29, no. 9, pp. 1223 – 1230, 1996.

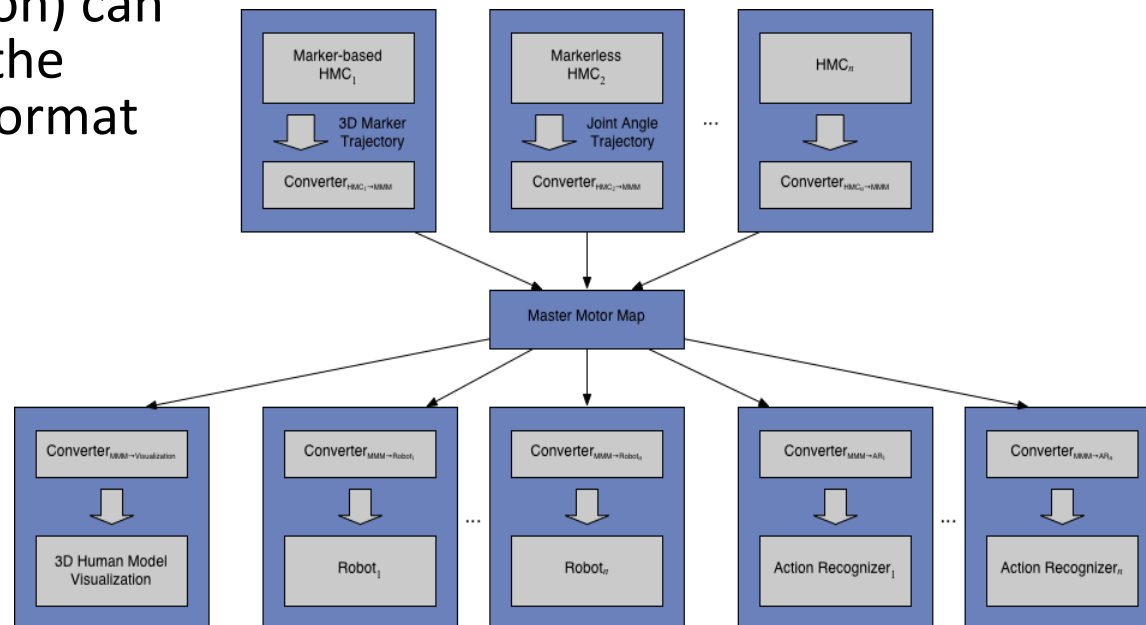
Master Motor Map (MMM)

- Interfaces and data structures for the transfer of motor knowledge between different embodiments



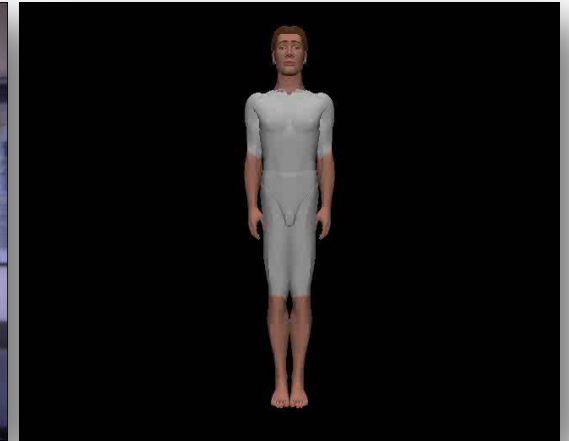
Master Motor Map (MMM)

- Replacement of any module (perception, recognition, visualization, reproduction) can be guaranteed by using the MMM as the exchange format
- All perceptive module convert their output to the MMM format
- All recognition and reproduction modules convert the MMM format to their specific internal representation

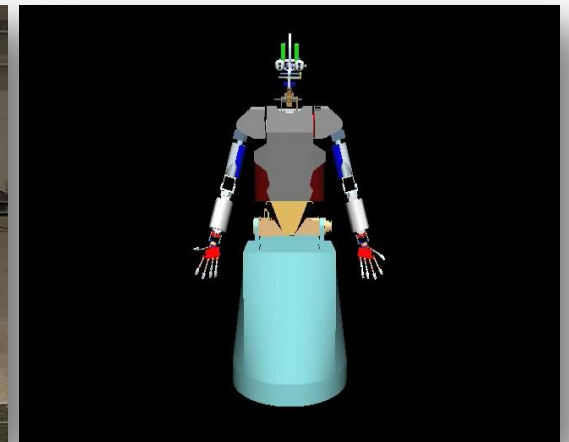
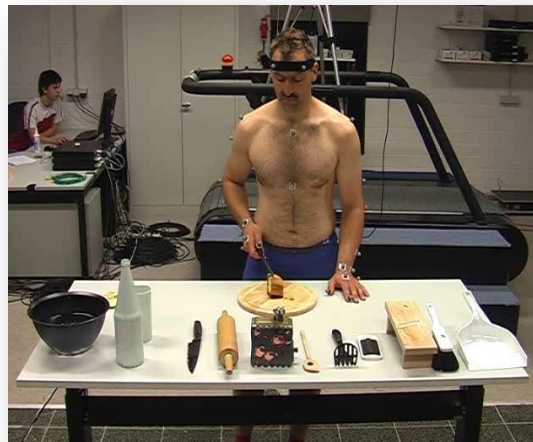


Motion reproduction using MMM

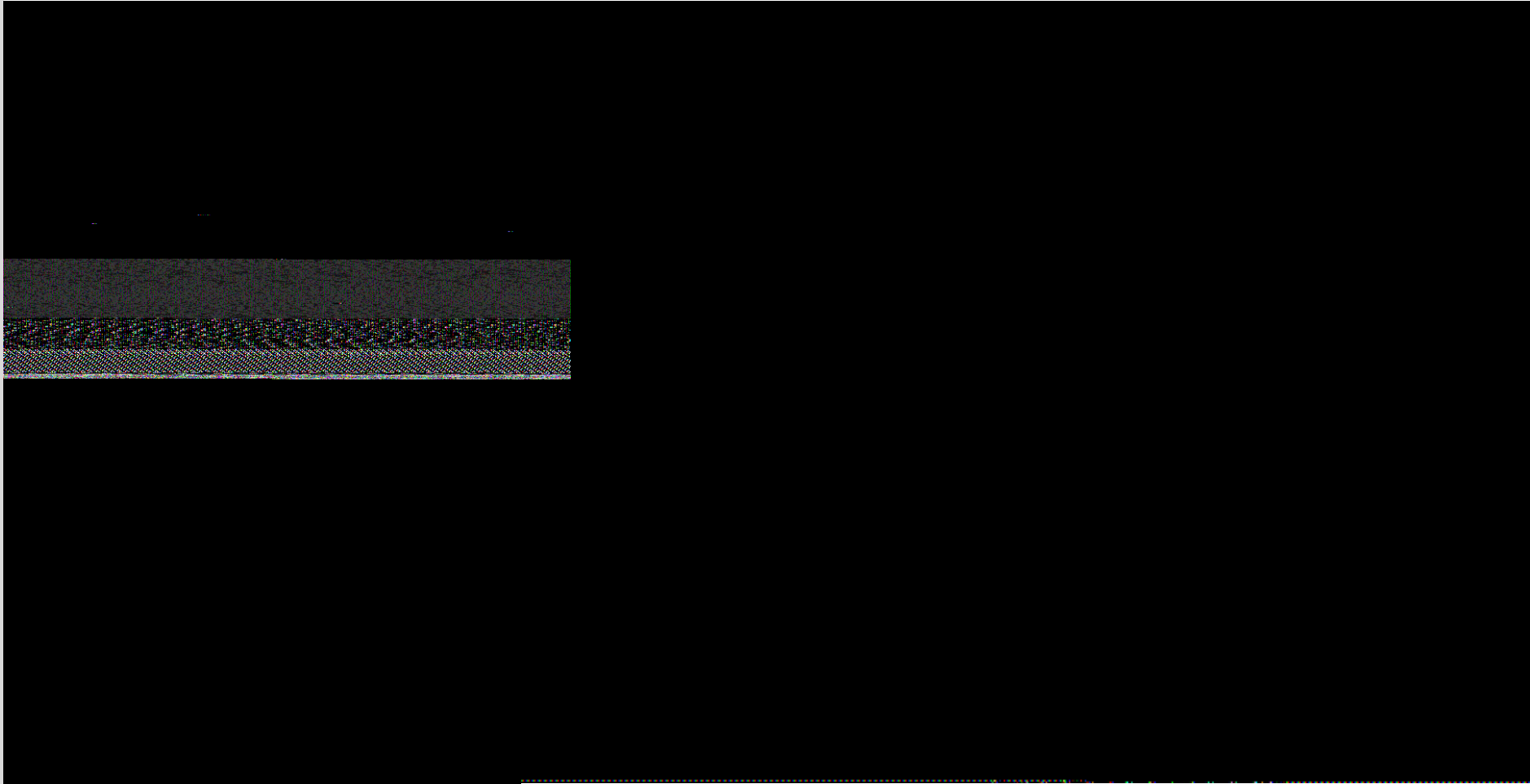
- Data from stereo-based markerless human motion capture system



- Data from VICON system (SFB 588)

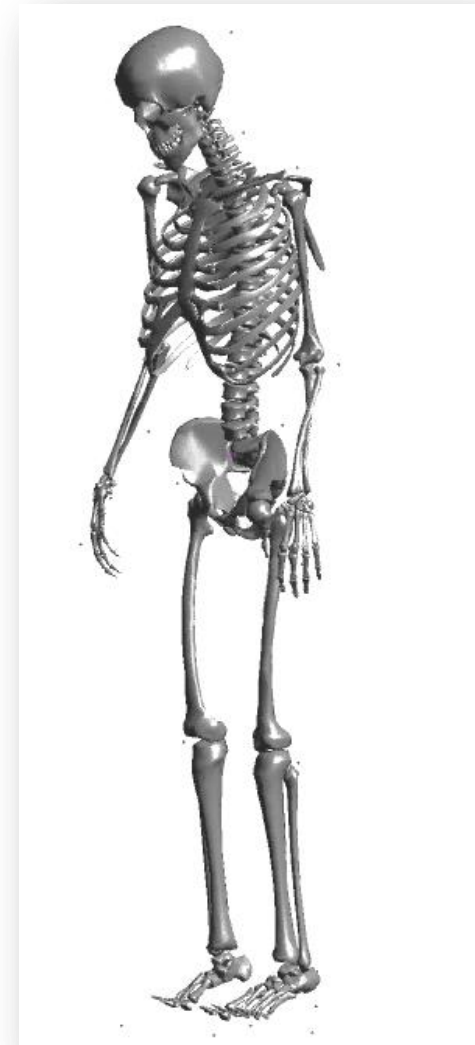
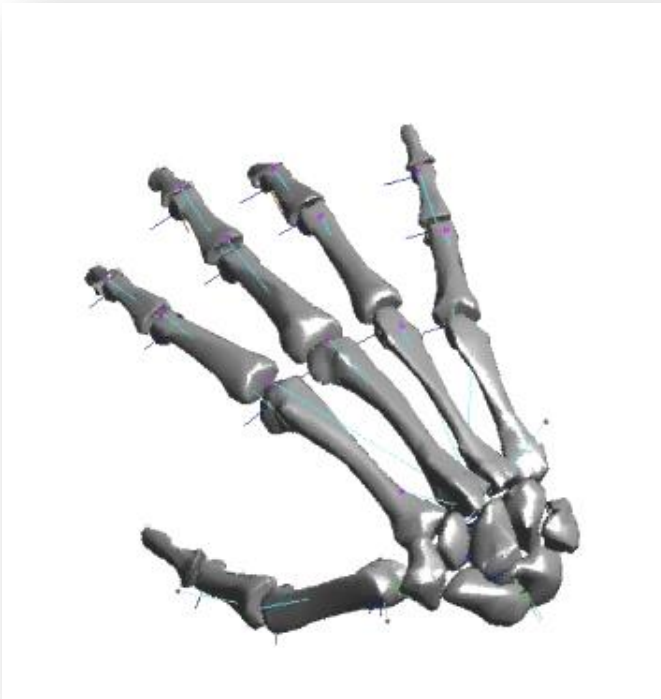


Motion Reproduction using MMM

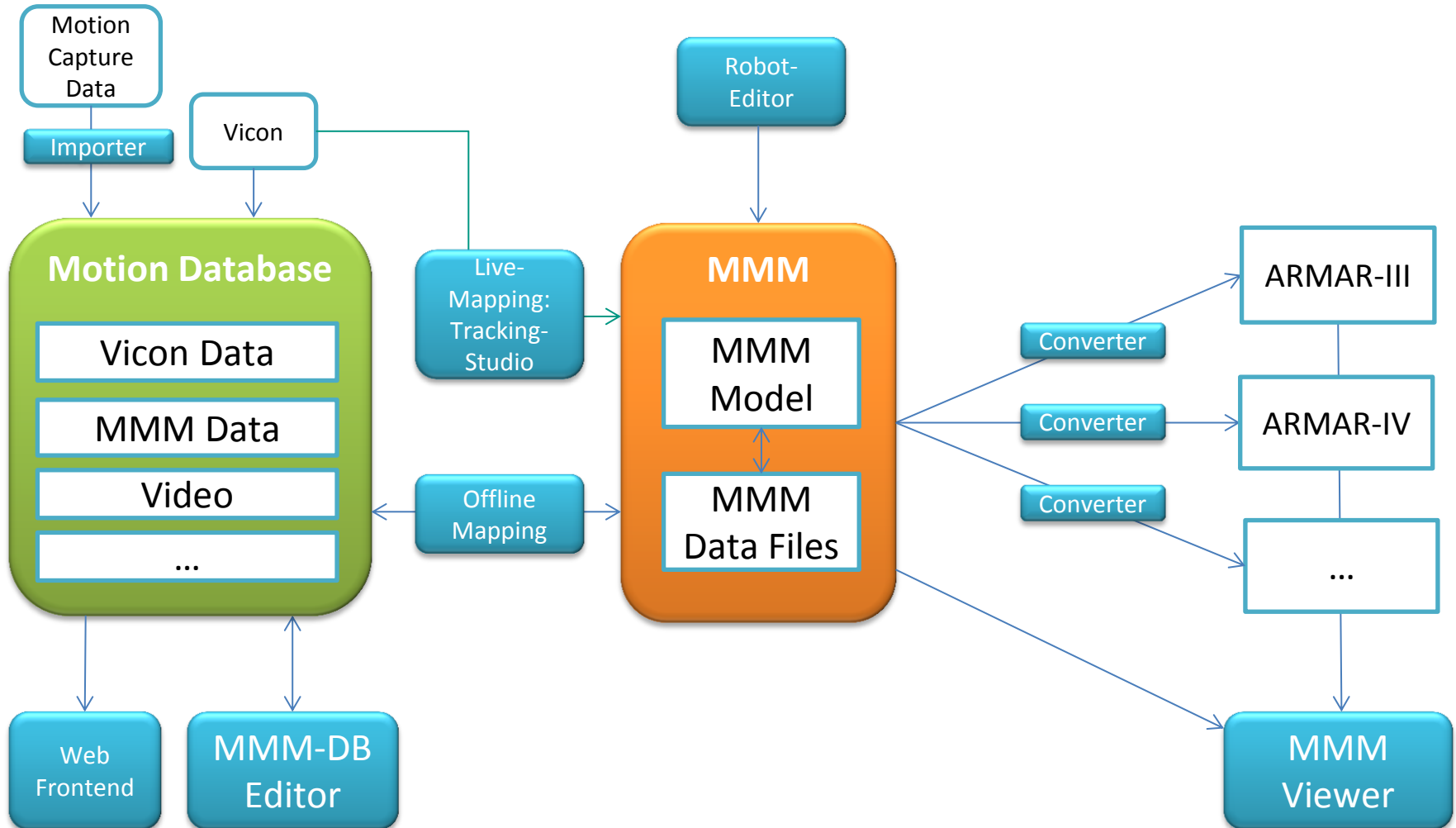


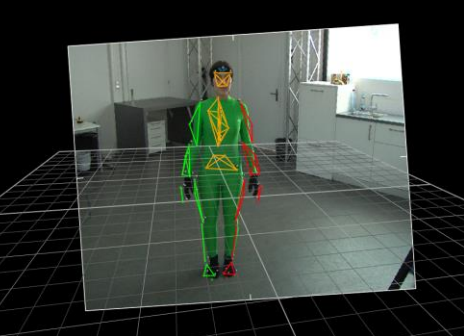
SFB 588

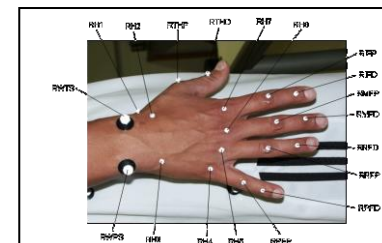
Motion Reproduction using MMM



Master Motor Map - Framework

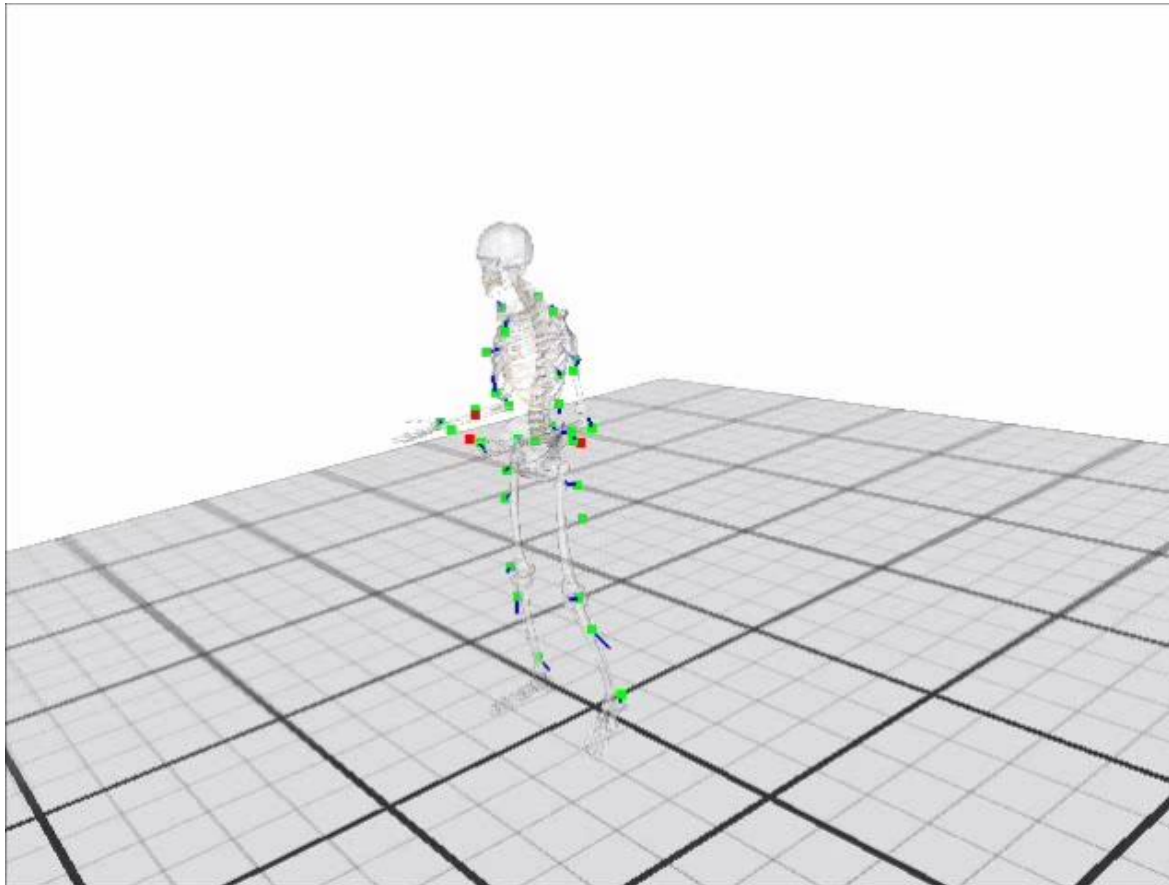


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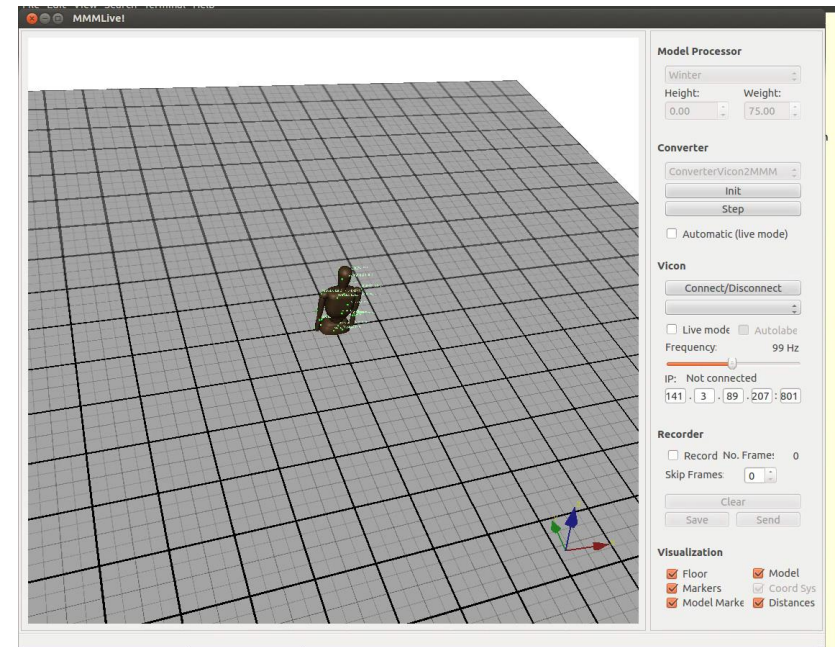
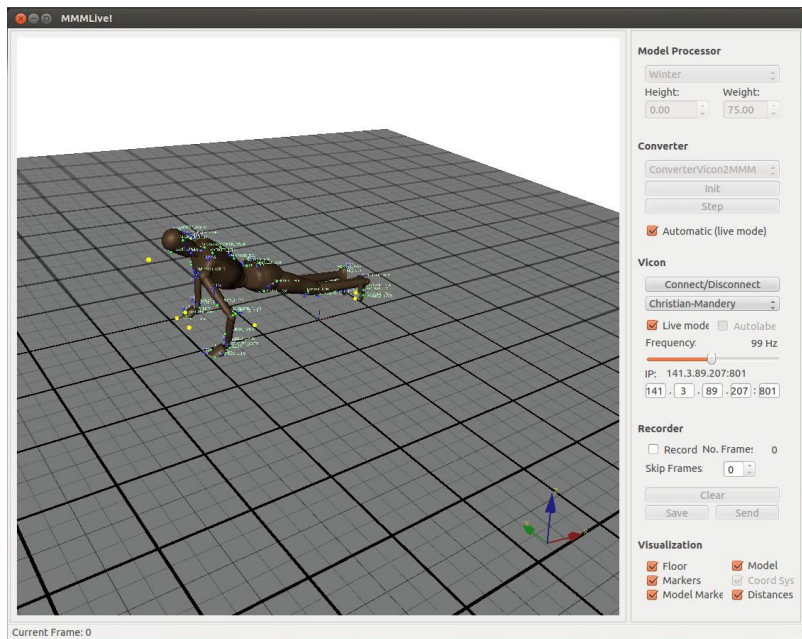


Large scale motion database & MMM

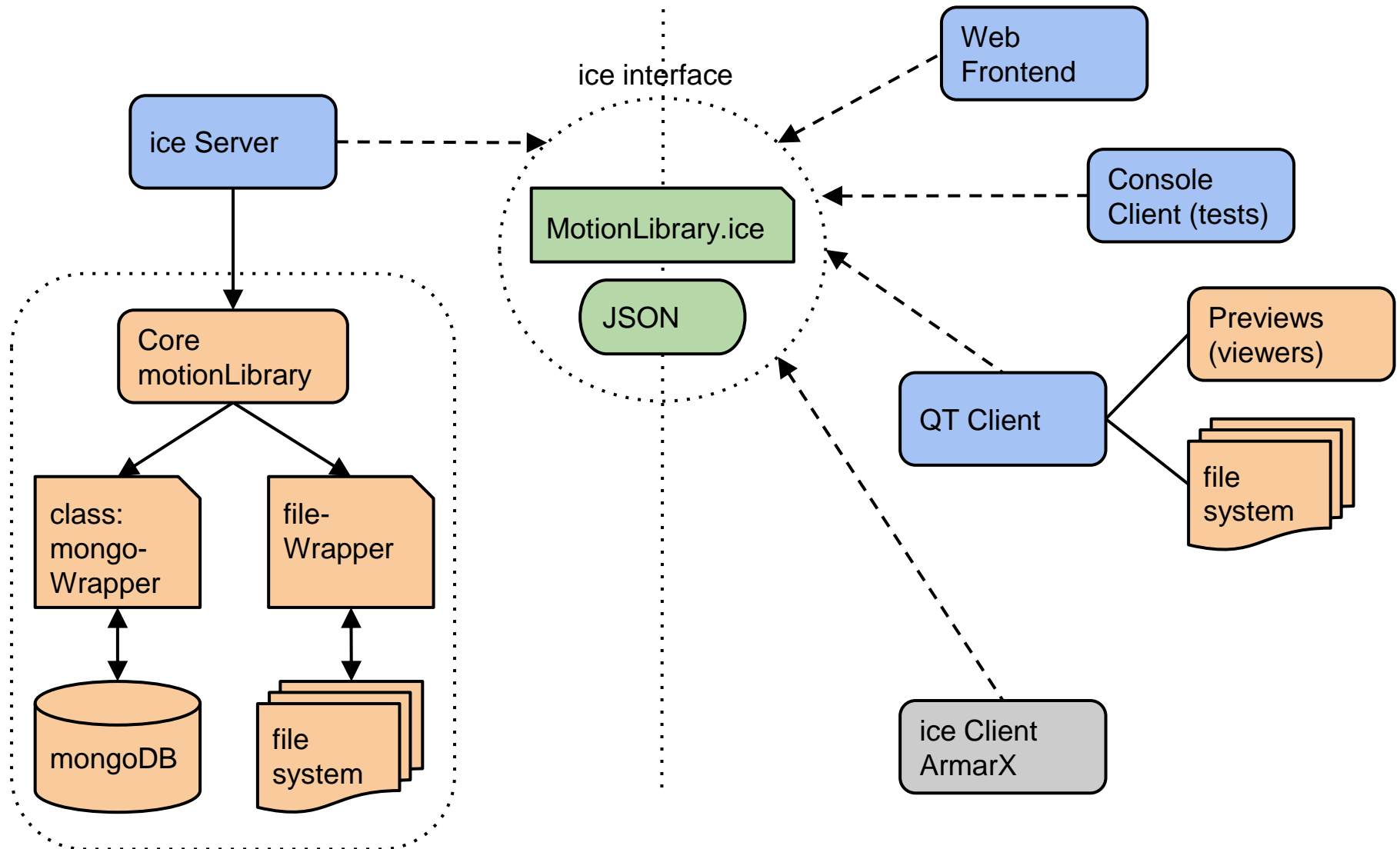
- Tracking Studio: Live mapping from human to MMM



Live Motion Capture and Mapping



MMM Tools



MMM Library & Tools

■ MMM Core

- C++ Library
- I/O, XML, Raw Marker Data, Tools, Conversions
- No dependencies (just Boost)

■ Mapping / Converter

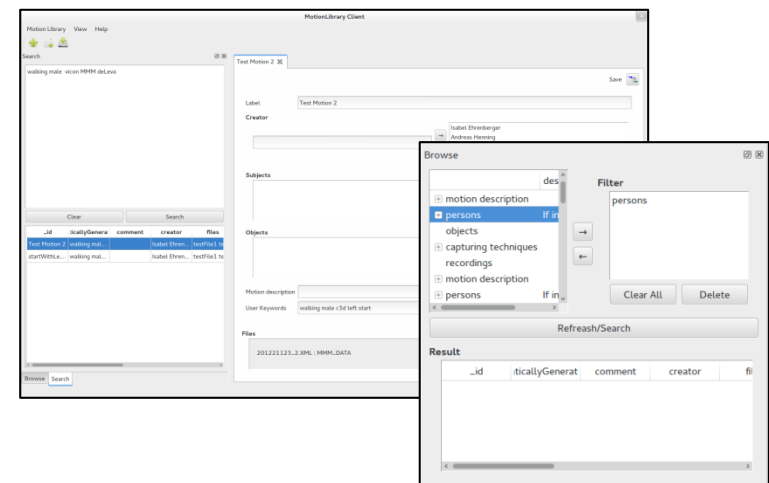
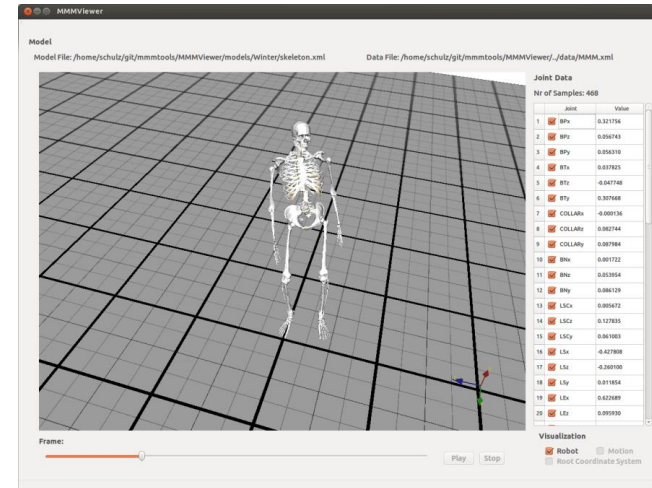
- Vicon -> MMM model
- MMM -> Robots (ARMAR III, ARMAR IV)
- MMM -> Other robots (iCub, COMAN, HRP, ...)

■ MMM Viewer

- 3D Model Viewer
- MMM / Marker Data
- Robots, Motions, Contacts, ...

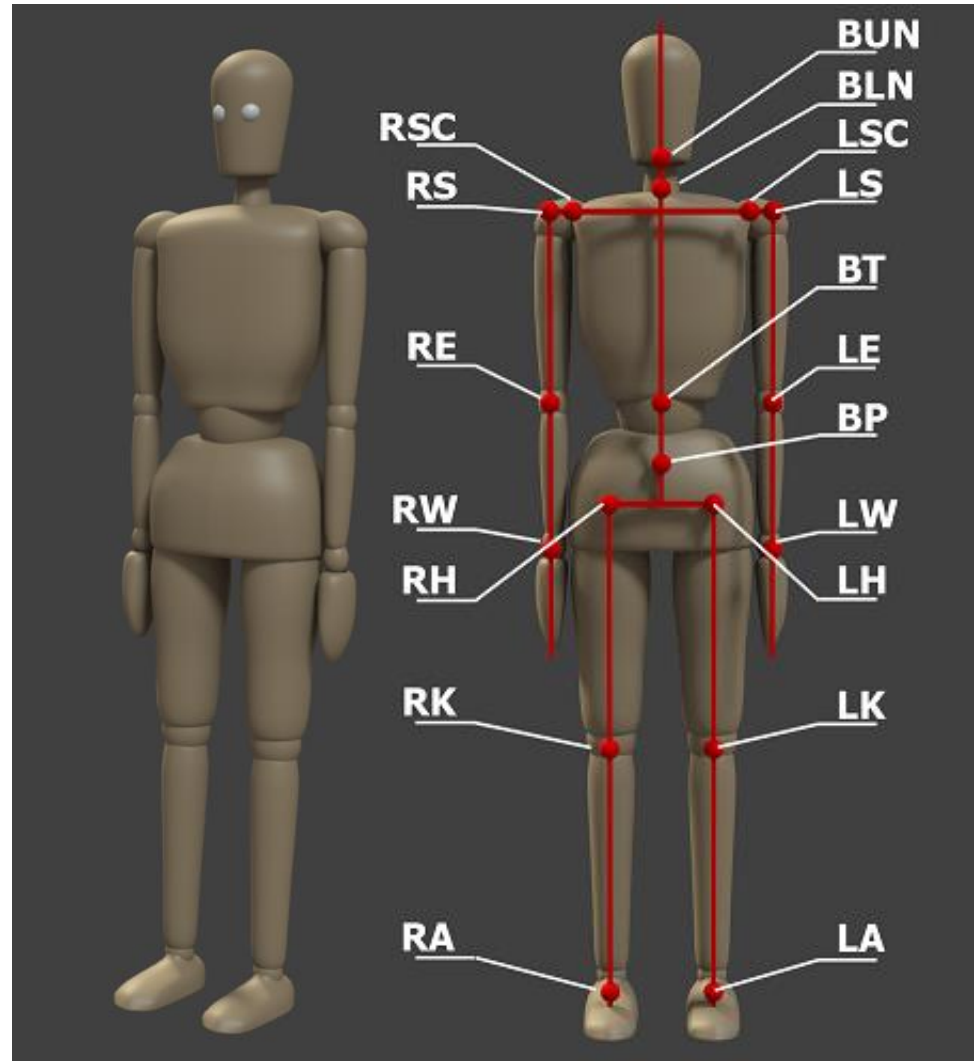
■ MMM Database

- Server
- Client/Applications, Search, Web Frontend



MMM new visualization

- Modelled in Blender
- 50 DoF are included (out of 96)
- Scalable
 - Every segment is in relation to the total height of the model
- Includes dynamical properties for every segment
 - Weight
 - Center of Mass (CoM)
 - Inertia-tensor
 - Scales with total weight



References

■ Our previous work

- S. Lengagne, O. Terlemez, S. Laturnus, T. Asfour and R. Dillmann, **Retrieving Contact Points Without Environment Knowledge**, IEEE-RAS International conference on Humanoid robots, pp. 841 - 846, 2012
- S. Gärtner, M. Do, C. Simonidis, T. Asfour, W. Seemann and R. Dillmann, **Generation of Human-like Motion for Humanoid Robots Based on Marker-based Motion Capture Data**, 41th International Symposium on Robotics (ISR), pp. 1 - 8, 2010
- Pedram Azad, Tamim Asfour and Ruediger Dillmann. **Toward an Unified Representation for Imitation of Human Motion on Humanoids**. IEEE International Conference on Robotics and Automation, 2007

■ Others

- David A. Winter. **Biomechanics and Motor Control of Human Movement**. John Wiley & Sons, Inc. 2005
- P. de Leva, **Adjustments to Zatsiorsky-Seluyanov's Segment Inertia Parameters**, J. of Biomechanics, vol. 29, no. 9, pp. 1223 – 1230, 1996.
- Nicolas Pronost, Georges Dumont. **Validating re-targeted and interpolated locomotions by dynamics-based analysis**. Proceedings of the 4th international conference on Computer graphics and interactive techniques in Australasia and Southeast Asia. 2006
- Michael Gleicher. **Retargetting Motion to New Characters**. SIGGRAPH 2008

Execution on Armar III



What's next

ARMAR-V

- First step towards humanoid robots with **multiple functions** and for **dual use**

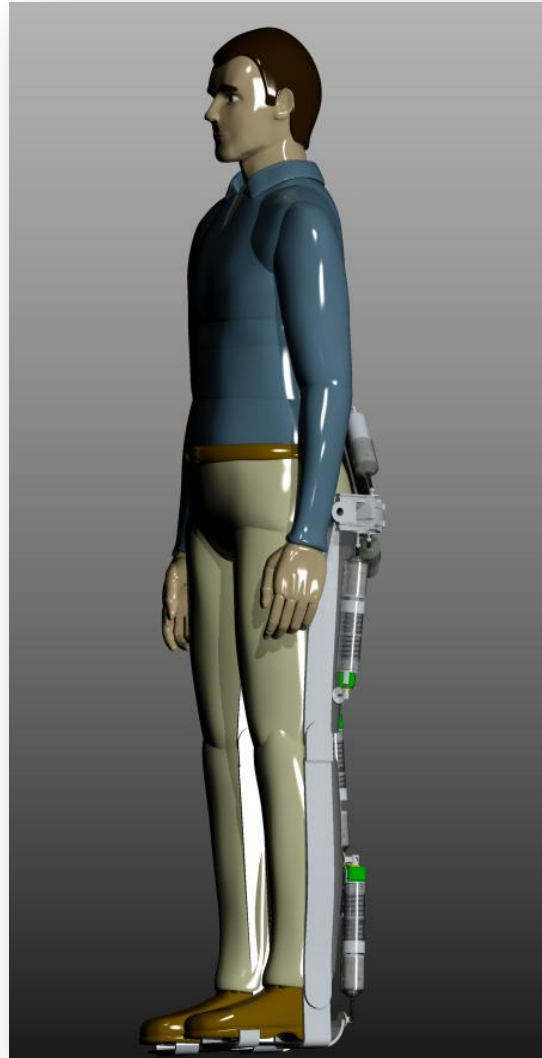
Helper, Assistant
and Companion



Wearable Humanoid
„Body suit“

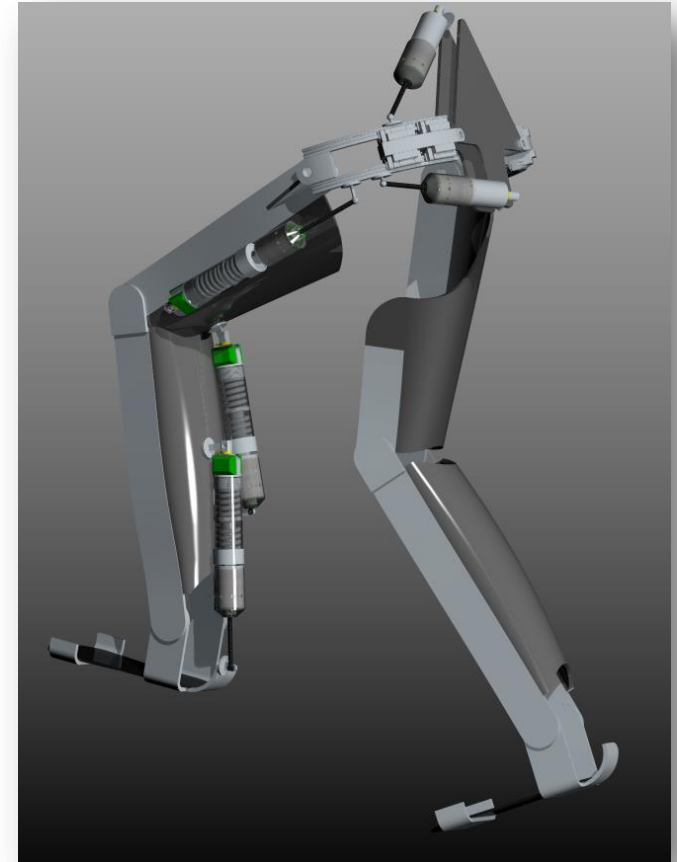
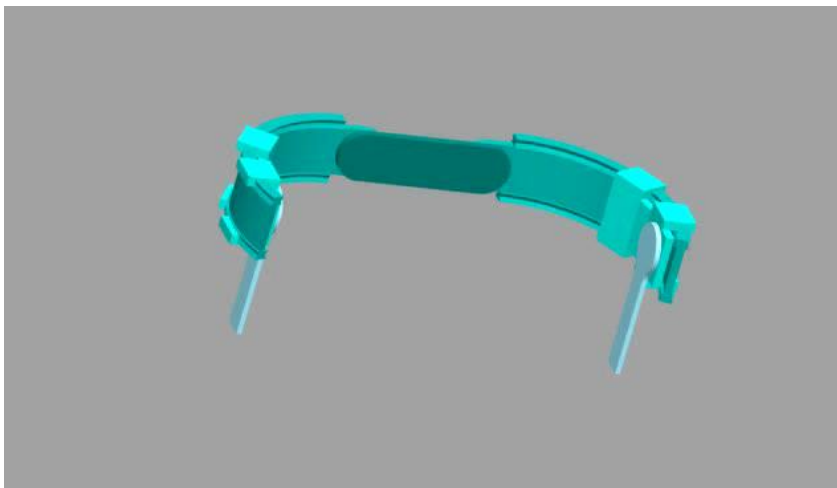


ARMAR-V: Legs

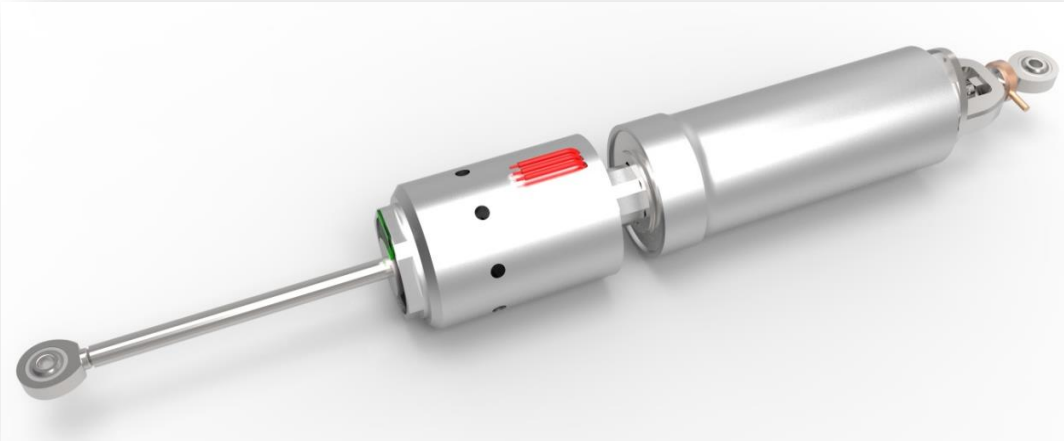


ARMAR-V Legs

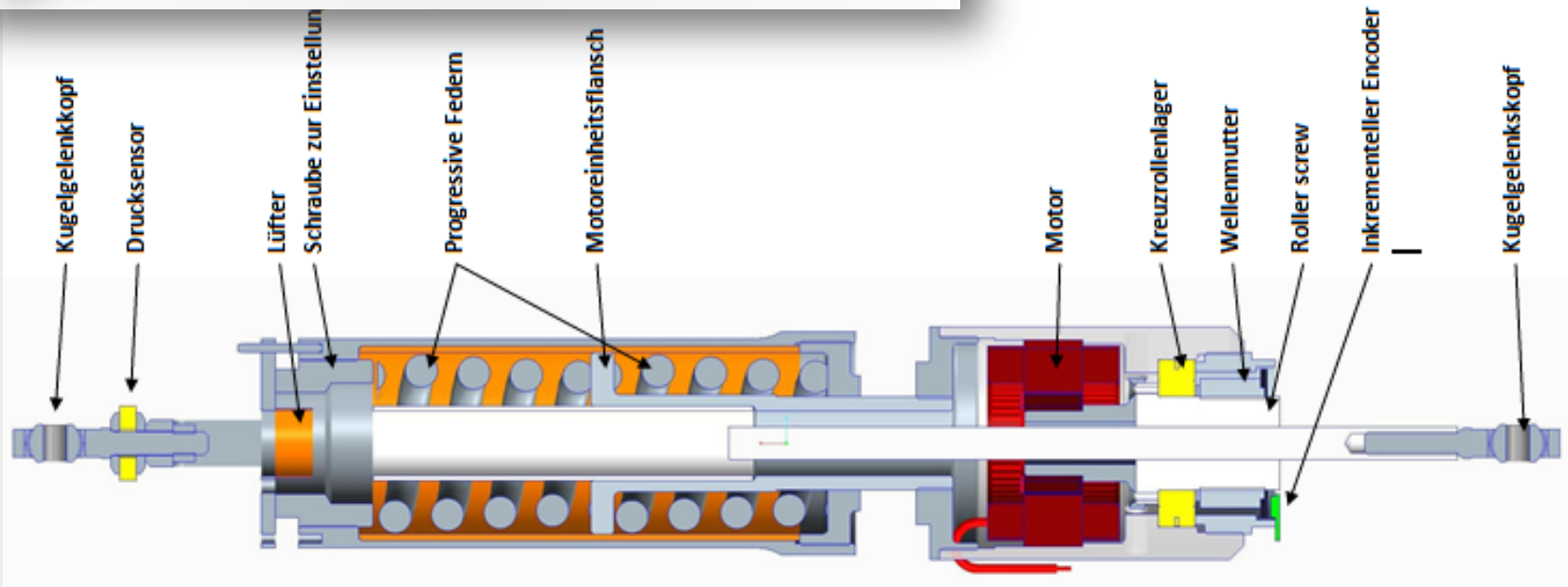
- 5 actuated DOFs in total in each leg
 - 3 DOFs in the hip
 - 1 DOF in the knee
 - 1 DOF in the ankle
- Serial elastic actuation in 3 pitch – DOFs
 - Adjustment of elasticity
- Joint peak torques ~ 120 Nm



ARMAR-V Legs – New Linear Elastic Actuators

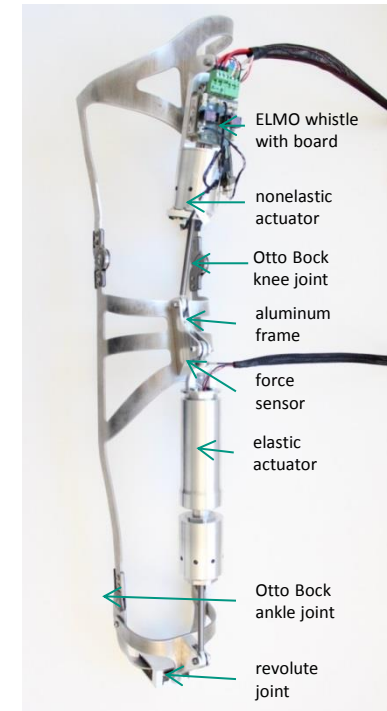
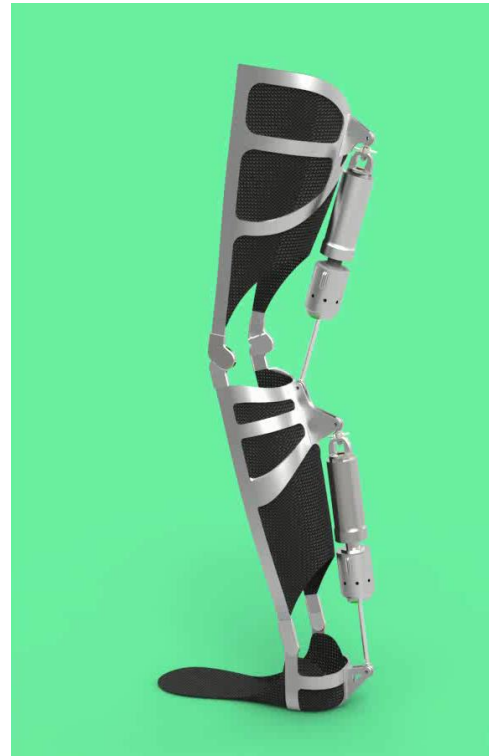


- Maximal axial force 2900 N
- Force at nominal motor torque 930 N
- Speed 300 mm/sec

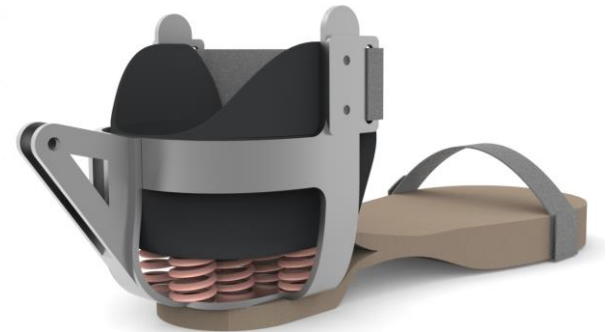
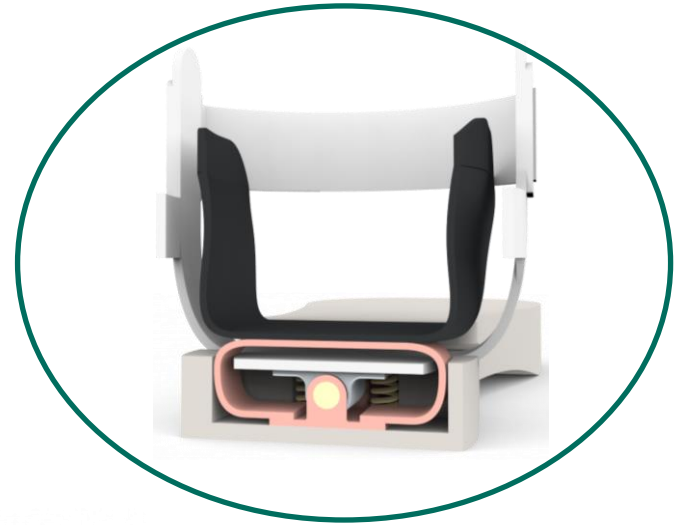
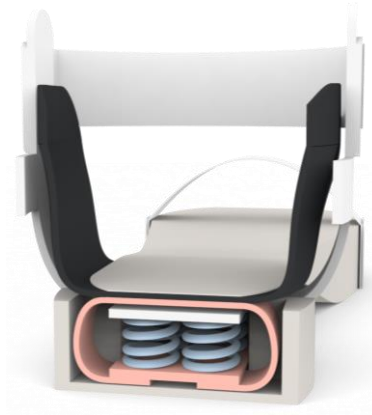


First version – with two elastic actuators

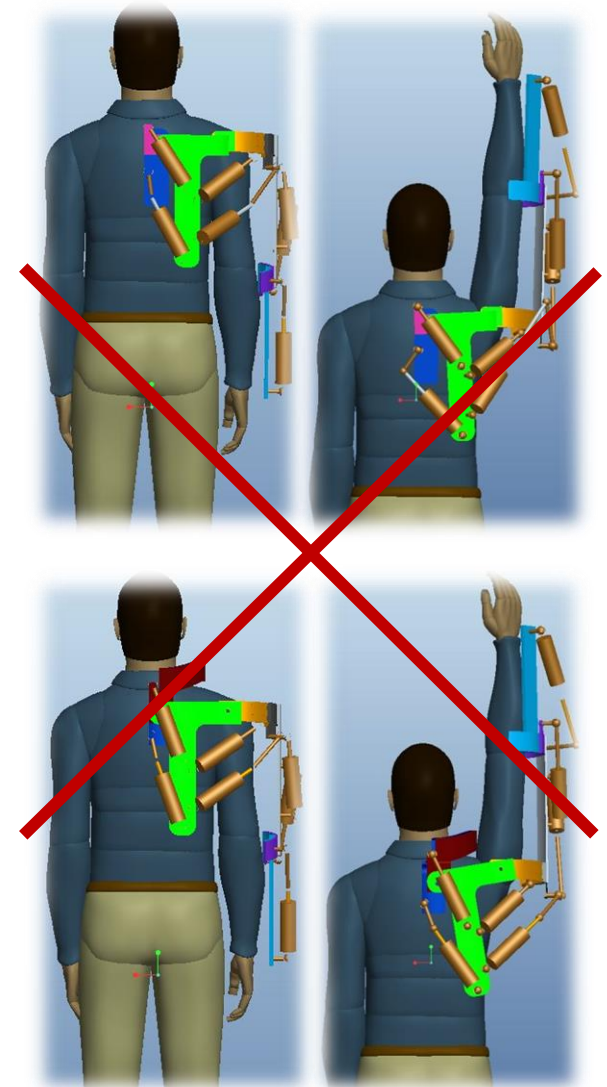
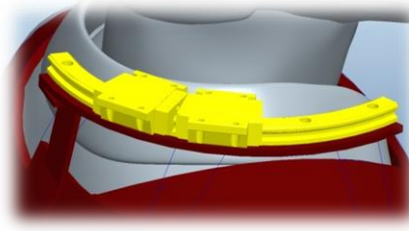
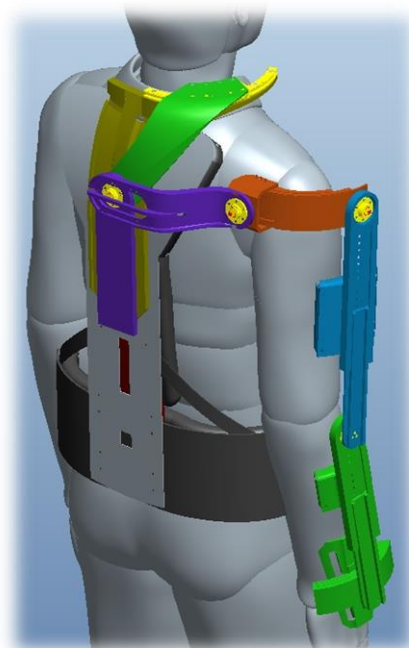
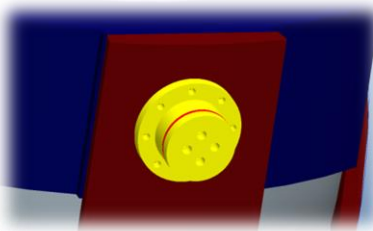
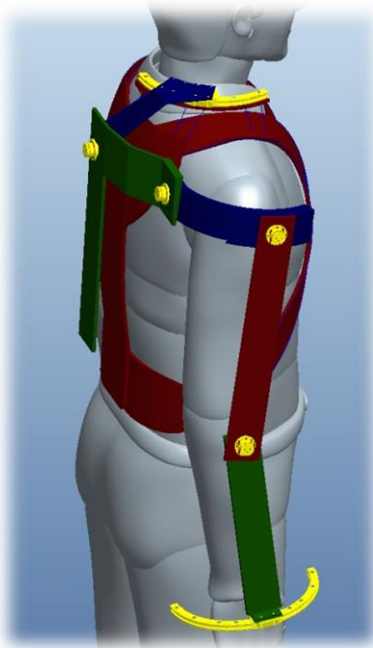
- Progressive springs integrated in the actuators for energy storage and reuse
- Serial-elastic type with 2 progressive springs
- Manually adjustable stiffness by increasing preload on springs



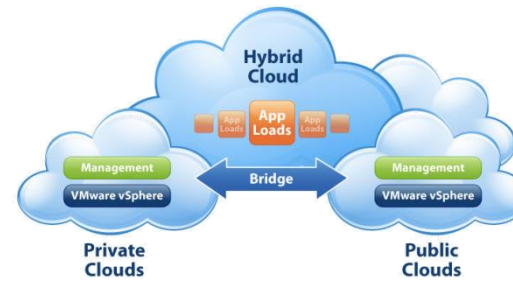
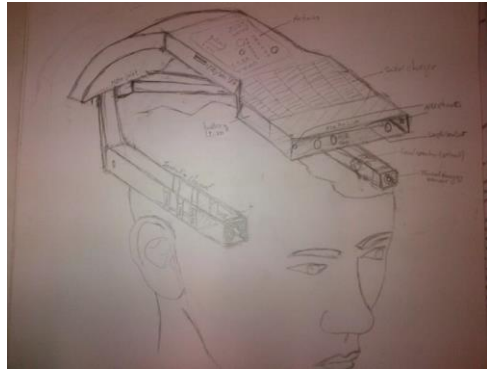
Foot and ankle joint – concepts



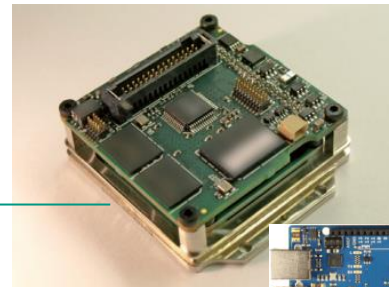
ARMAR-V: Upper body



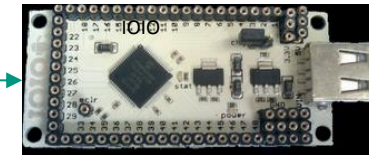
ARMAR-V: Head/ smart Helmet



E-Nose



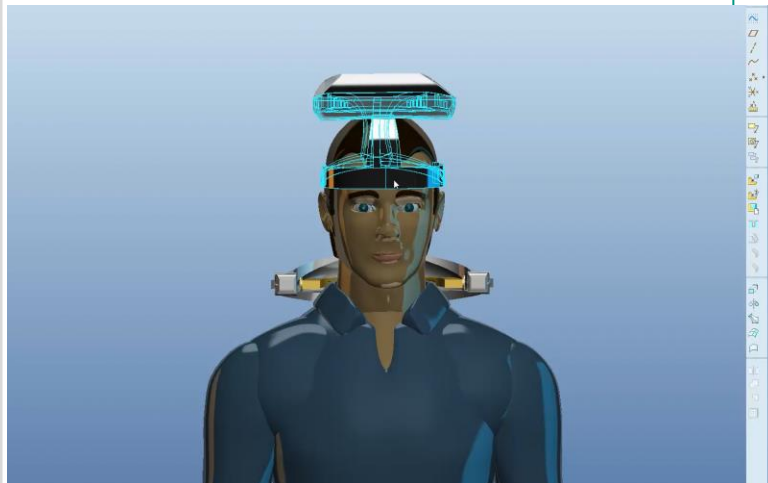
Vibration
motors



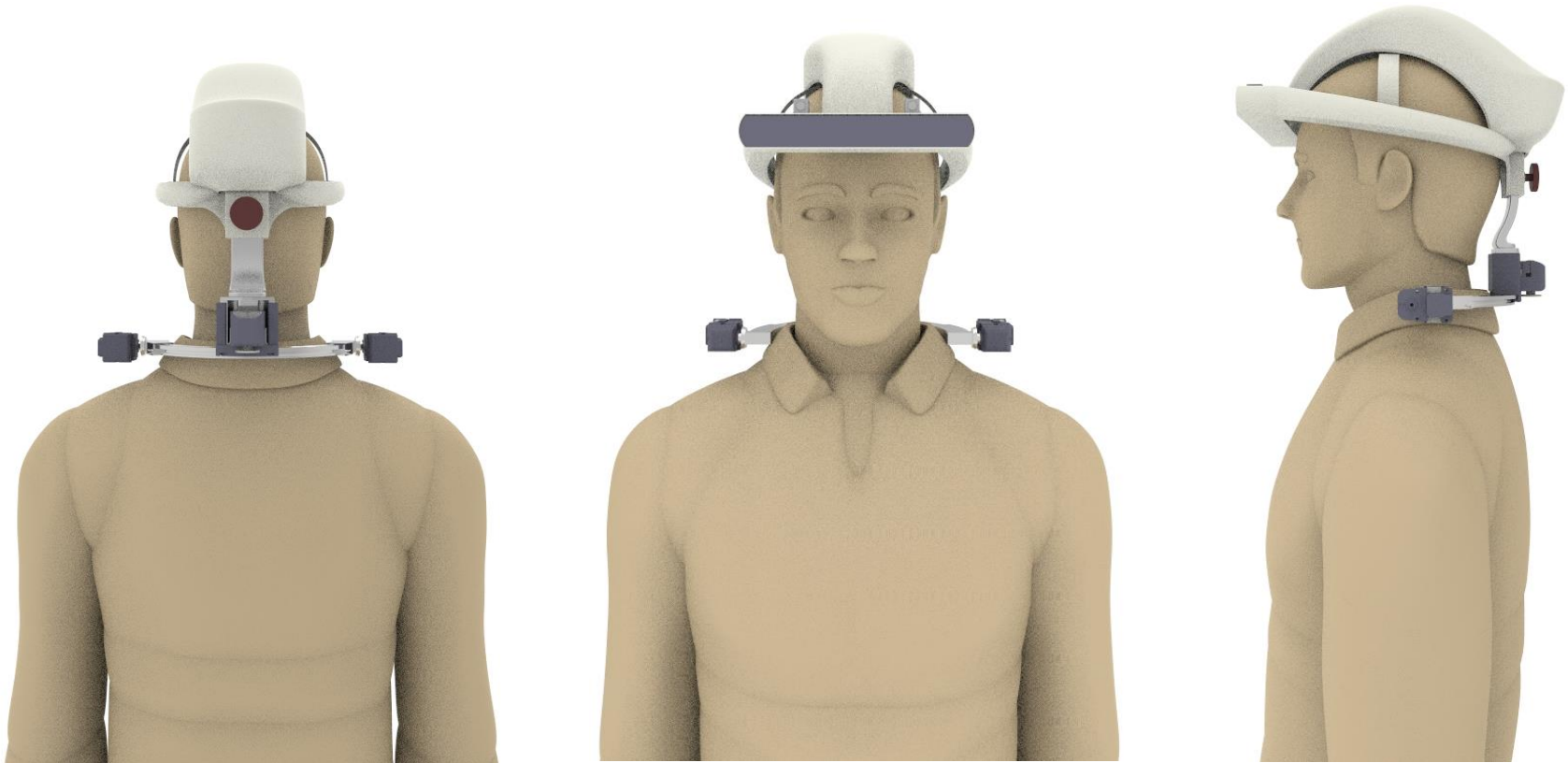
Solar charger



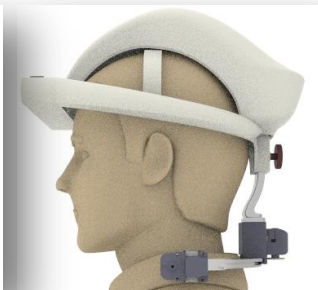
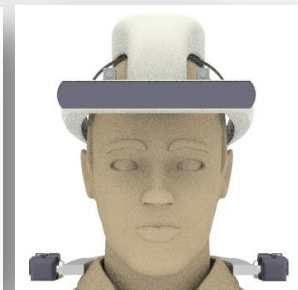
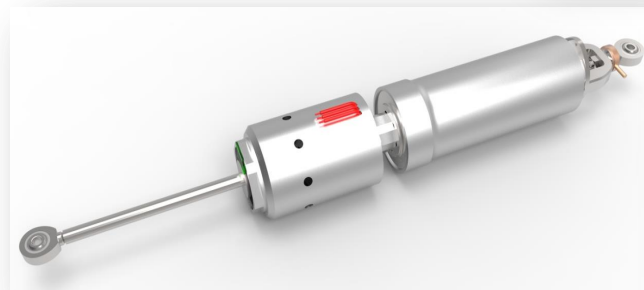
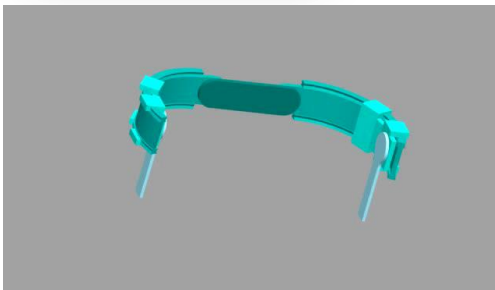
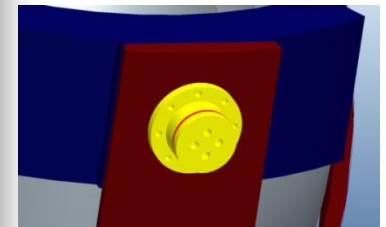
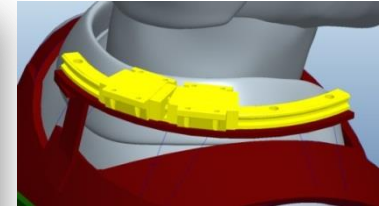
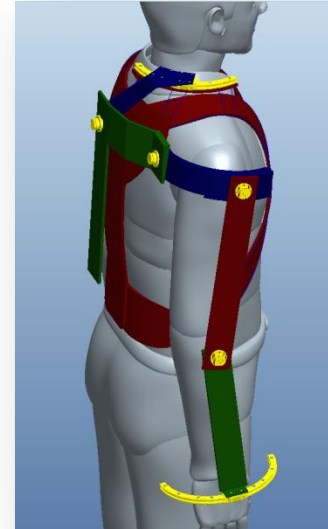
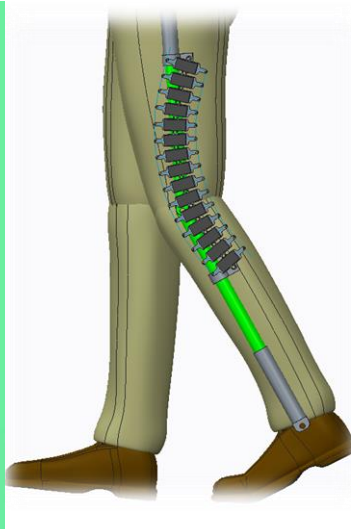
- Vision
- Audio
- Chemical
- Ultra low power thermal cameras
- Solar charger
- ...



ARMAR-V: Head/ smart Helmet



ARMAR-V: Wearable Humanoid

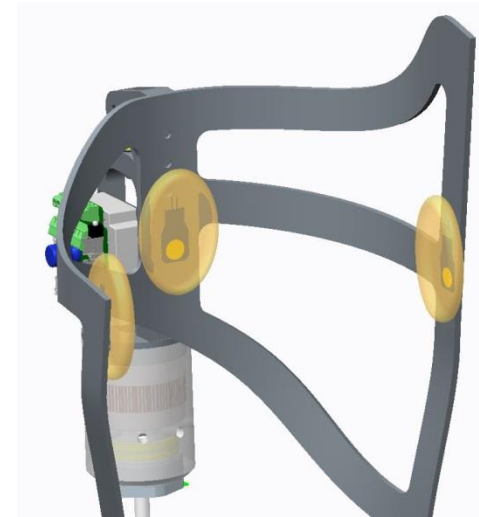


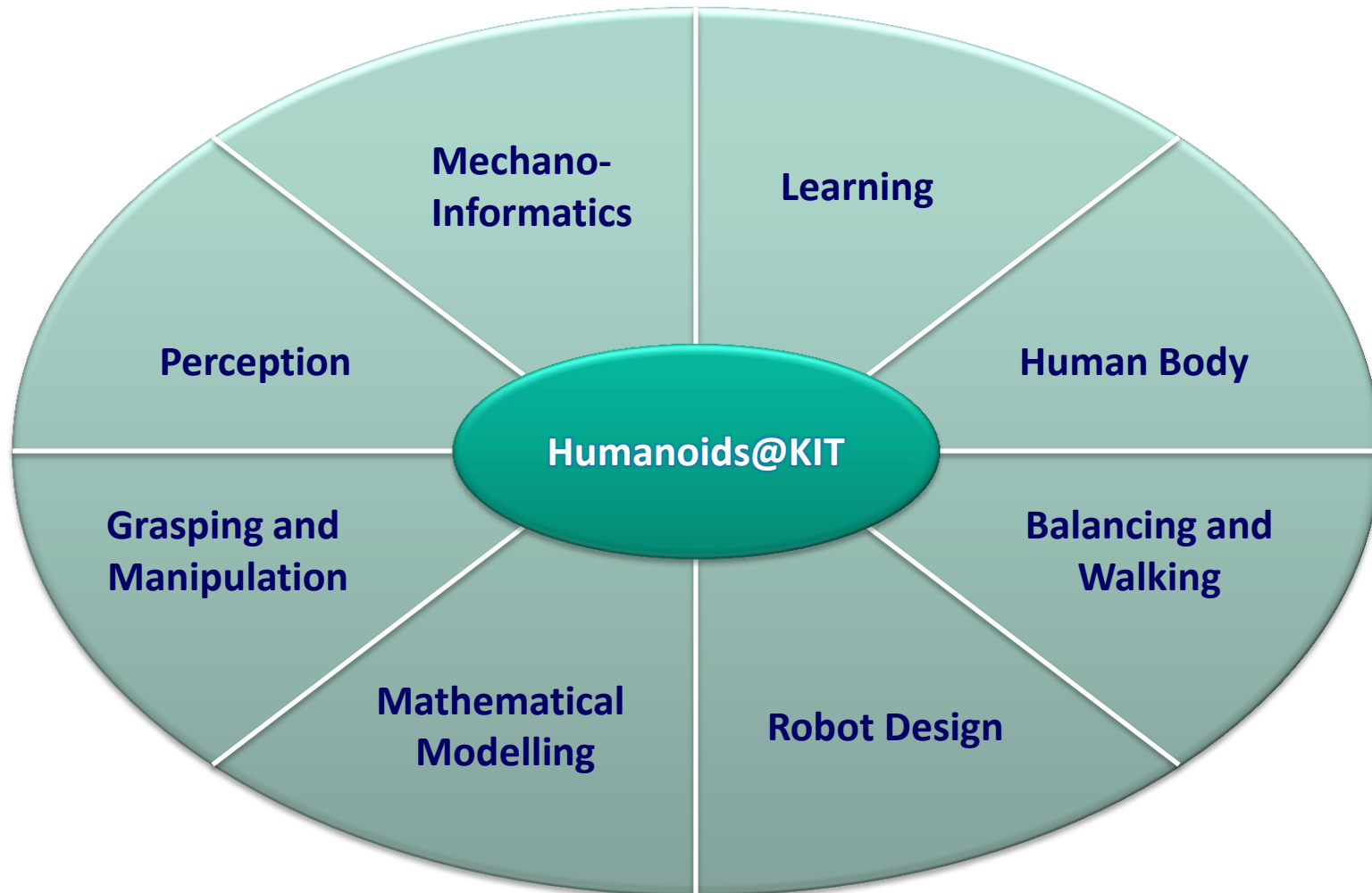
- Maximal „axial“ force 2900 N
- Force at nominal motor torque 930 N
- Speed 300 mm/sec

ARMAR V: Interface to the human body

■ Force sensor suit

- Non-invasive, EMG-free Interface to the human body
- Learn interaction force pattern between human and suit and use them for prediction
- EMG unreliable
- EMG can only be used to train a classifier as well as to study correlations between EMG pattern and force pattern “**feel the muscle activation**”





Thanks to ...

■ German Research Foundation (DFG)

- SPP 1527 autonomous-learning.org (2010 -)
- SFB/TR 89 www.invasic.de (2009 -)
- SFB 588 www.sfb588.uni-karlsruhe.de (2001 - 2012)



■ European Commission

- Xperience www.xperience.org (2012-2015)
- Walk-Man www.walk-man.eu (2013-2017)
- Koroibot www.koroibot.eu (2013-2016)
- PACO-PLUS www.paco-plus.org (2006-2011)
- GRASP www.grasp-project.eu (2008-2012)



■ Karlsruhe Institute of Technology (KIT)

- Professorship "Humanoid Robotic Systems"
- Heidelberg-Karlsruhe Research Partnership (HEiKA)



Thanks for your attention

