
Online Recognition of Daily-Life Movements

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Background

Collaborative Research Center 588 Humanoid Robots - Learning and Cooperating Multimodal Robots



Background

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- Understanding
- Cooperating
- Learning



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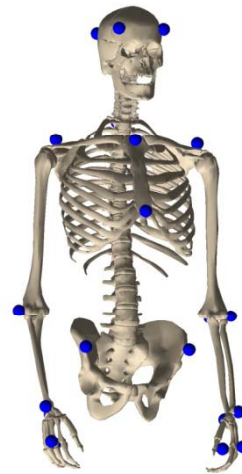
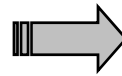
- Understanding
 - Cooperating
 - Learning
- What is the user doing?
- Needed: motion recognition system



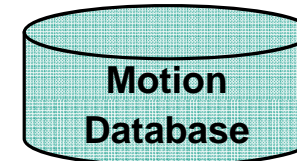
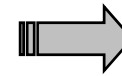
Motion Recognition System



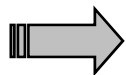
Live tracking by VICON



Mapping on human model



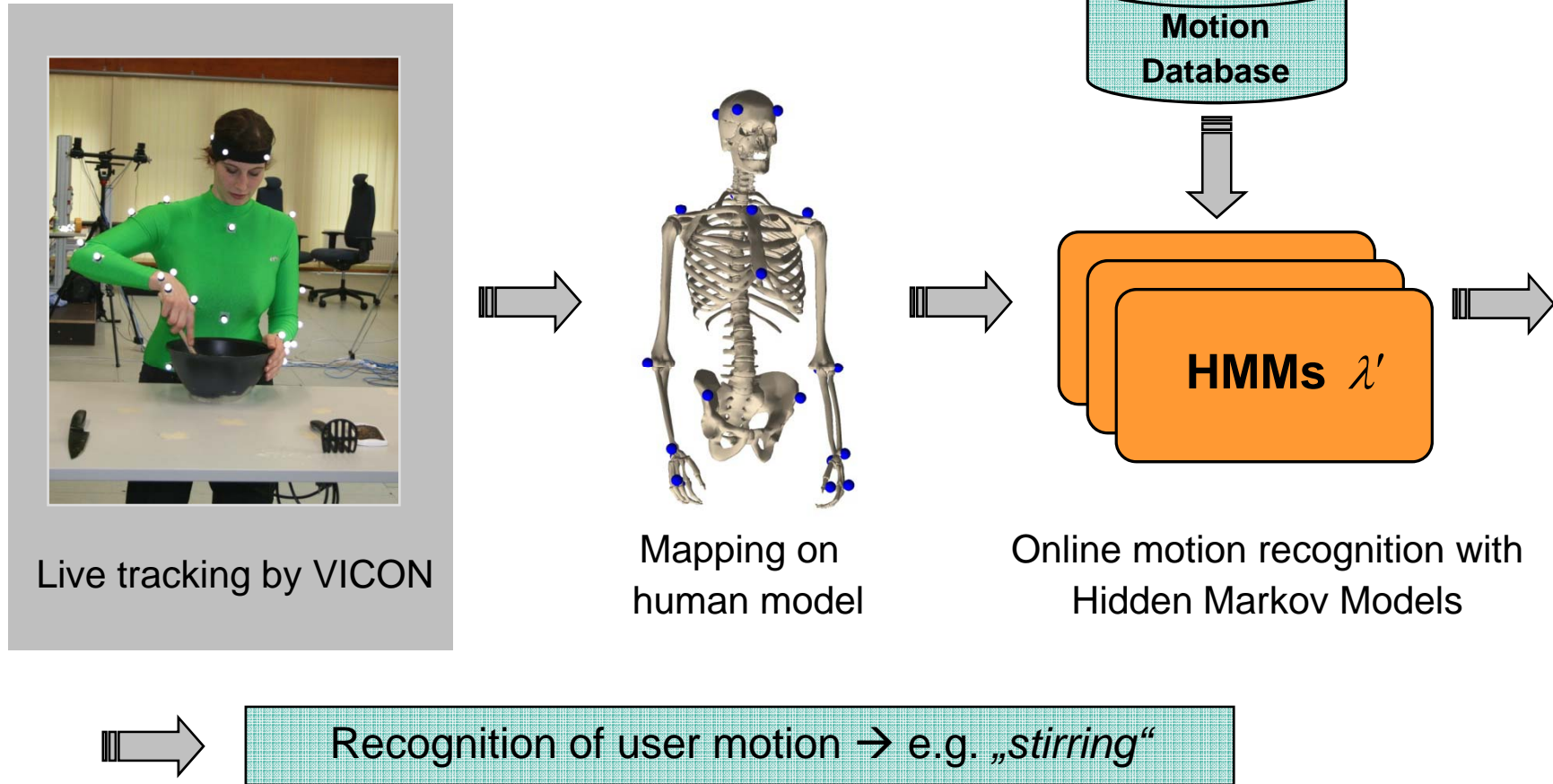
Online motion recognition with Hidden Markov Models



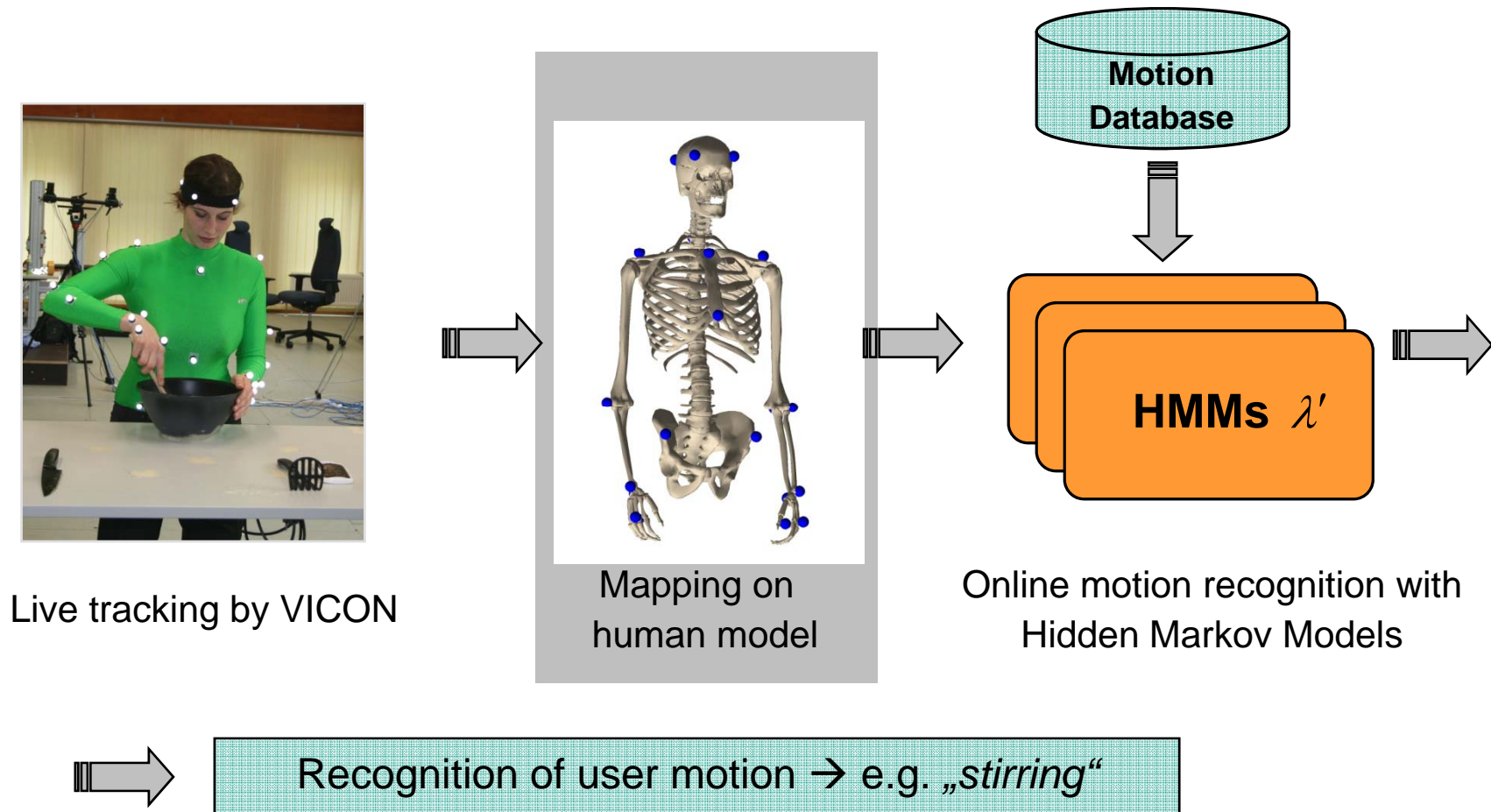
Recognition of user motion → e.g. „stirring“



Motion Recognition System



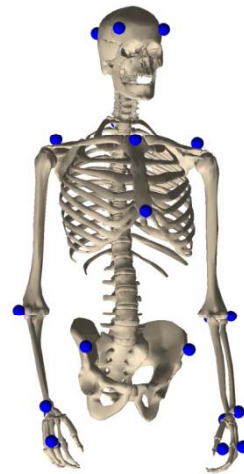
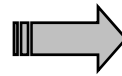
Motion Recognition System



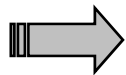
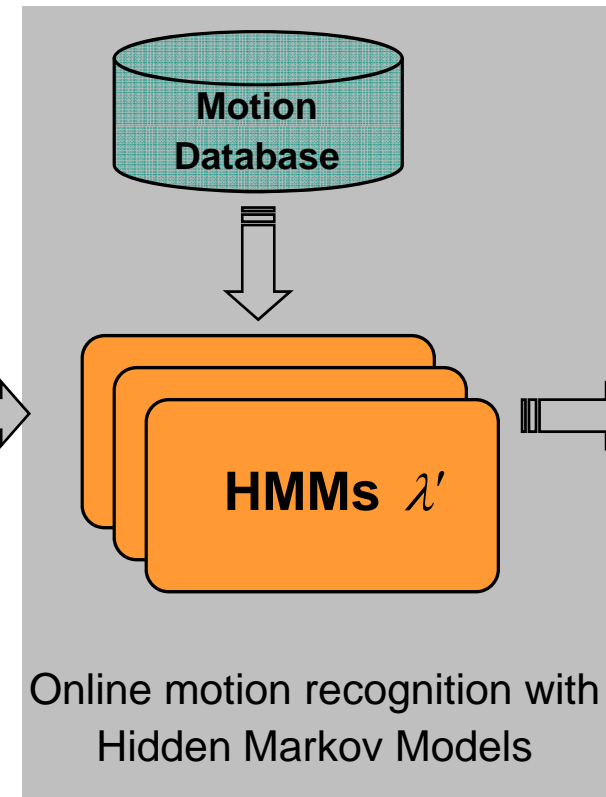
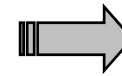
Motion Recognition System



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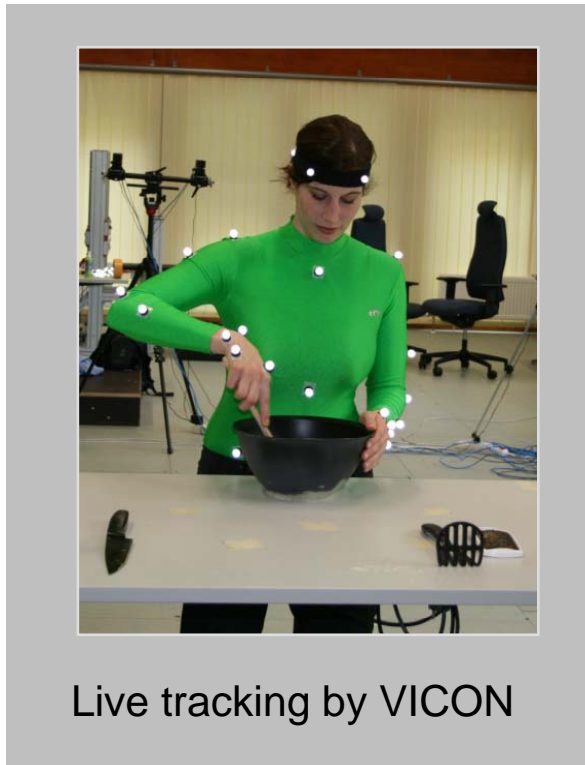
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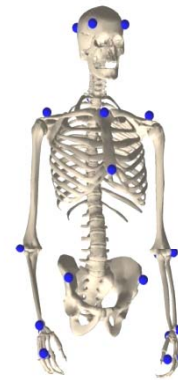
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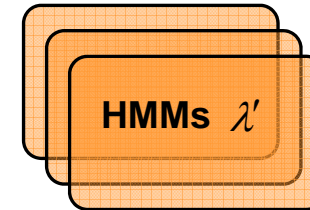
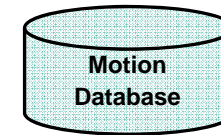
Data acquisition



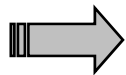
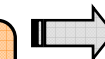
3D-Marker-
positions



Mapping on
human model



Online motion recognition with
Hidden Markov Models



Recognition of user motion → e.g. „stirring“



Motion Capture with Vicon

- BioMotion Center: Vicon system



Motion Capture with Vicon

- BioMotion Center: Vicon system
- 8 – 12 infra-red cameras
- Design of different marker sets



Motion Capture with Vicon

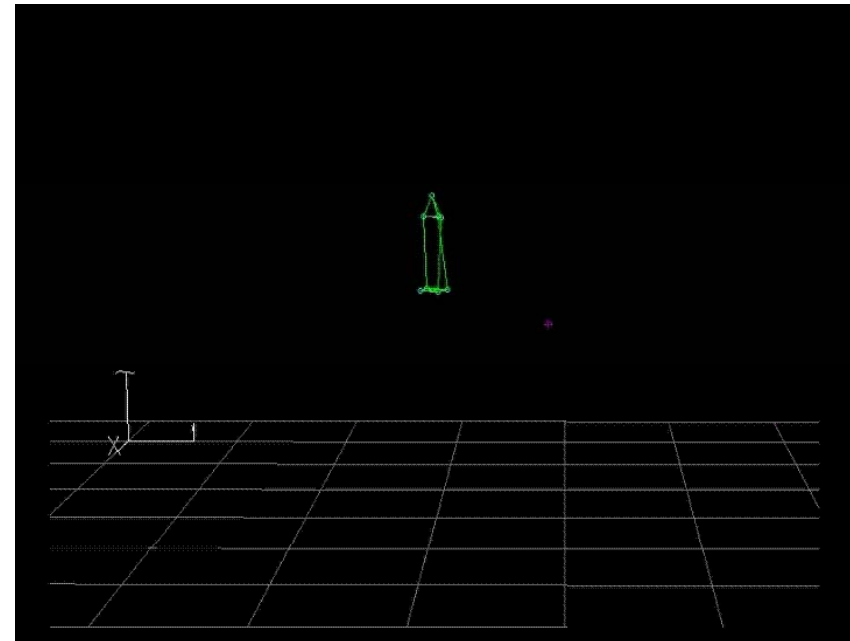
- BioMotion Center: Vicon system
- 8 – 12 infra-red cameras
- Design of different marker sets
- Marker motion Capture with Vicon
 - 2D marker positions from every camera
 - reconstruction of 3D position
 - marker labeling



Motion Capture with Vicon



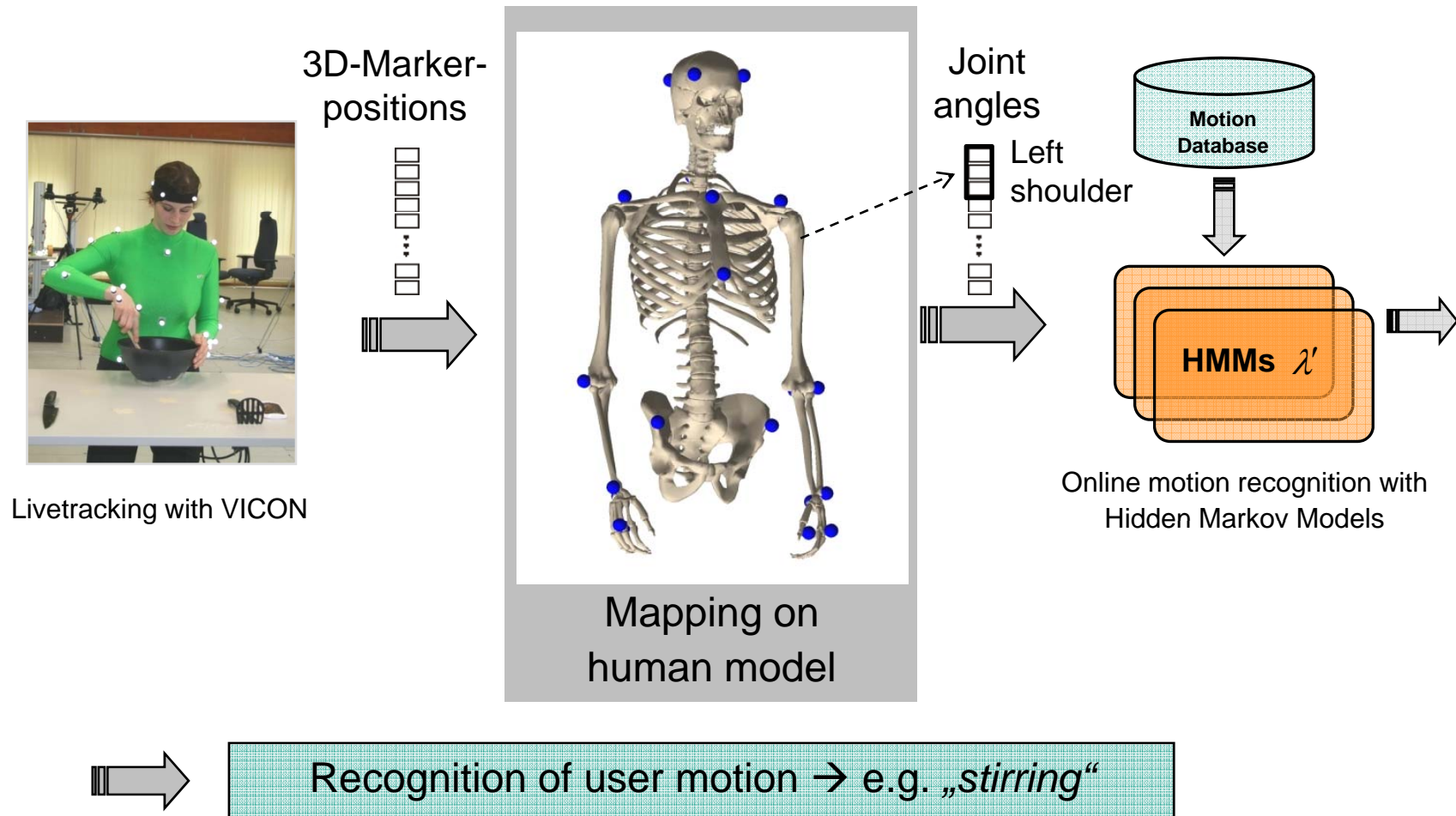
Example: Motion capture setting



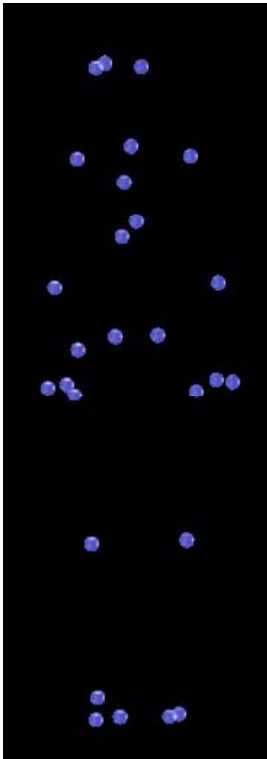
Result: Vicon record



Joint angle generation



Estimation of joint angles



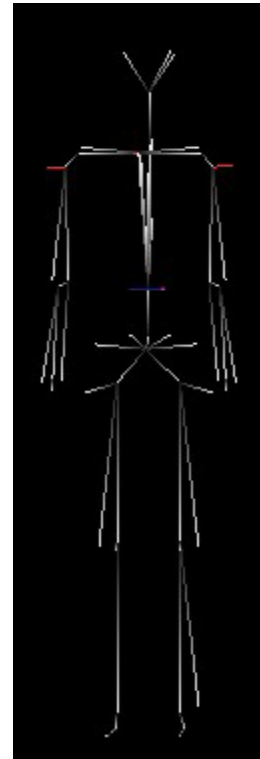
Input:
3D marker positions



Estimation of joint angles



+

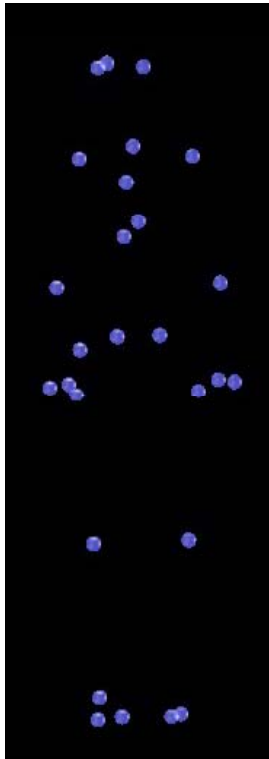


Input:
3D marker positions

Human model with
markers and joints

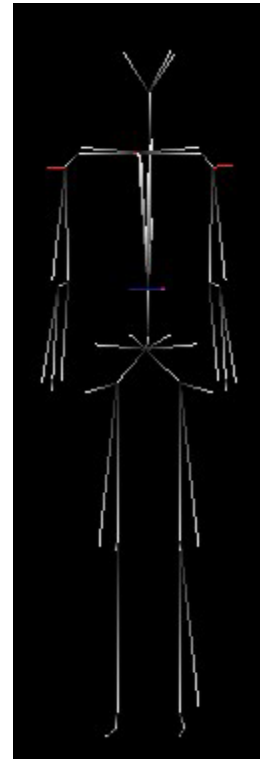


Estimation of joint angles



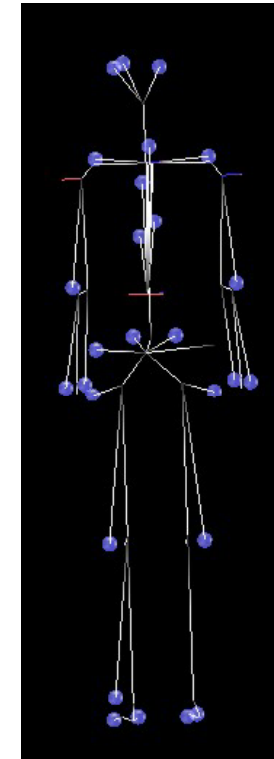
Input:
3D marker positions

+



Human model with
markers and joints

=

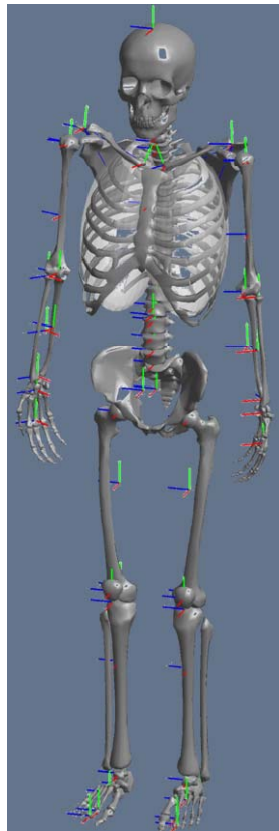


Pose
reconstruction

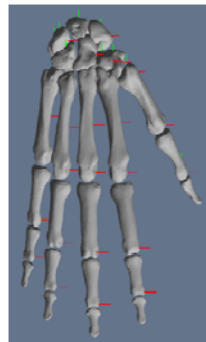


CRC 588 Human Model

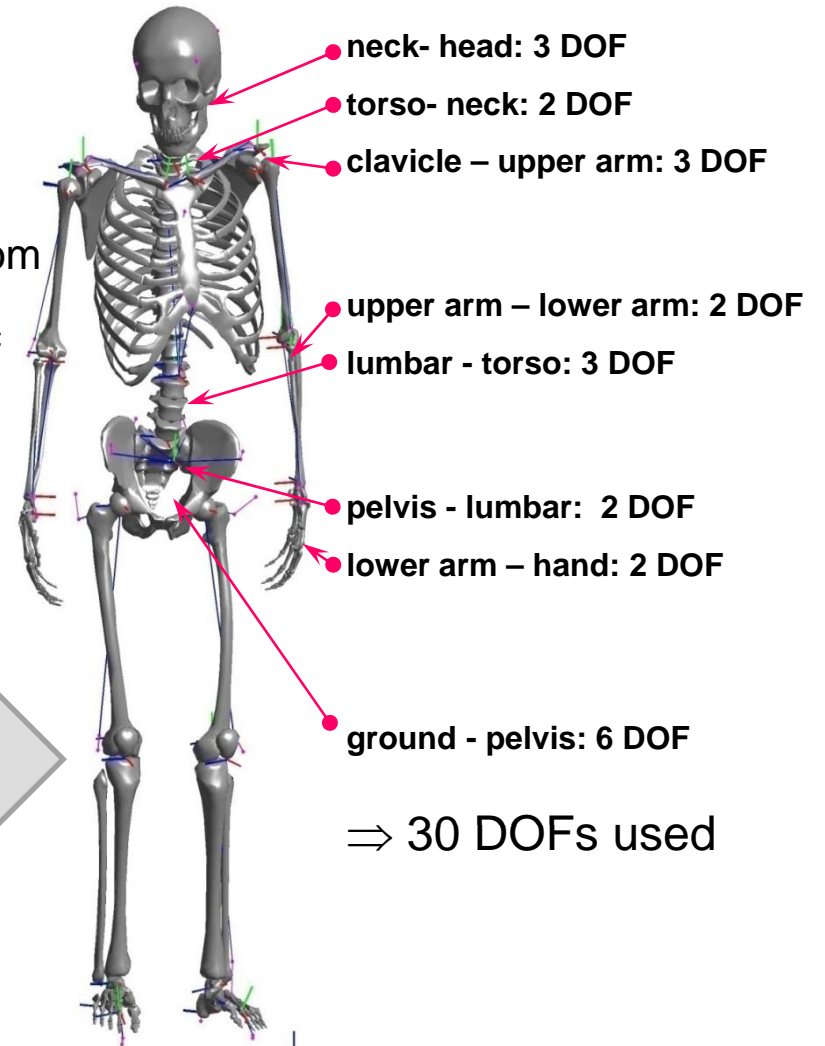
Kinematic human model



- full definition of the human body:
→ maximum of 108 degrees of freedom
- following real human joint kinematic
- basis for reduced models, marker sets and joint angle reconstruction



Example with reduced degrees of freedom (DOFs)



Estimation of joint angles

Main Idea

- Optimize the joint angles of the model so, that the distance of the actual marker set and the result of the forward kinematics of the reconstructed pose becomes minimal:

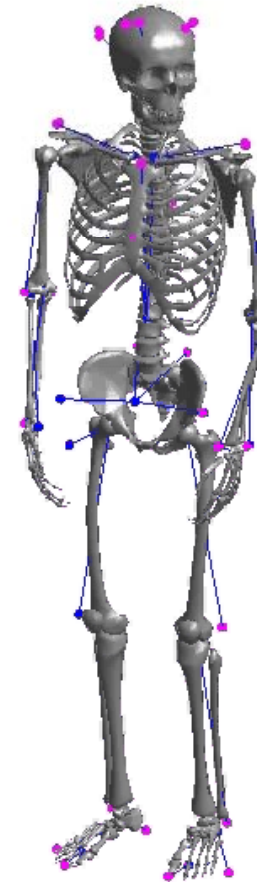
$$d(x, x_0) = \sqrt{(x_a - x_{0a})^2 + (x_b - x_{0b})^2 + (x_c - x_{0c})^2}$$

$$\min_{x \in X} \sum (d(x, x_0) \times \text{weight}(x_0))^2$$

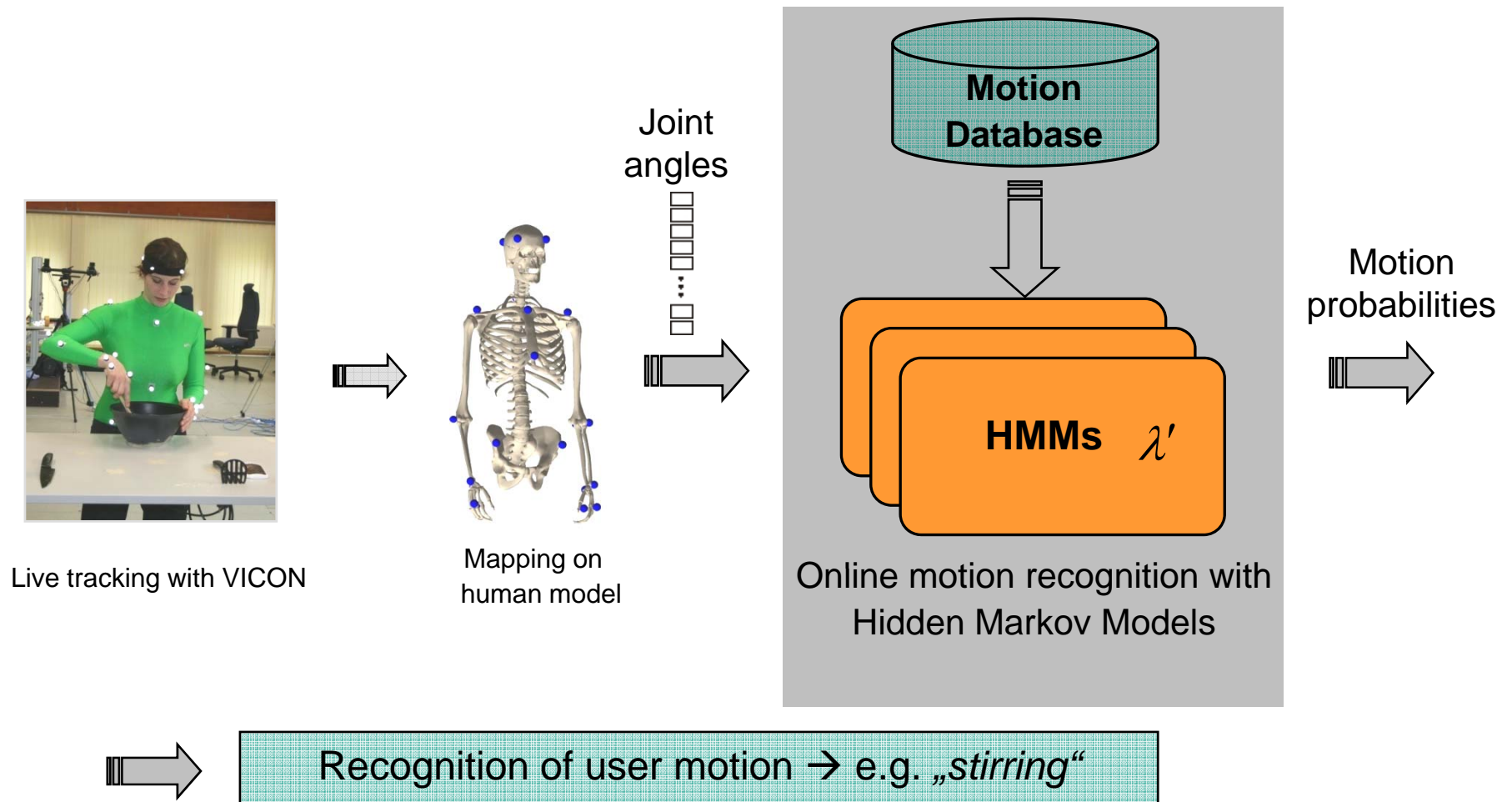
X : the set of the result of the forward kinematics

$\text{weight}(x_0)$: the weight of the markers

→ Reconstructed body pose / motion



Motion recognition



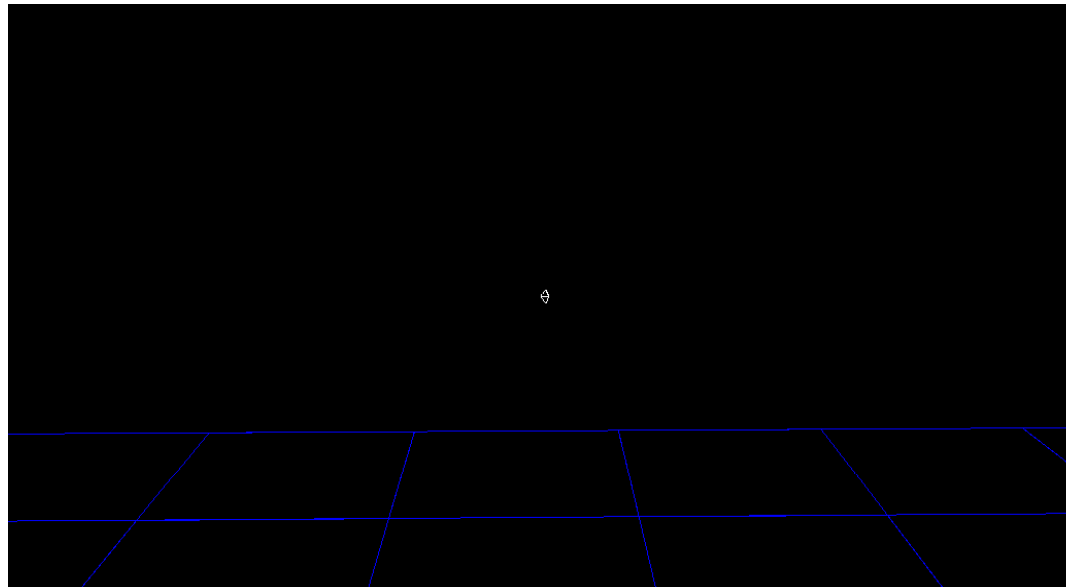
Problem: human motion

- Every motion sequence is unique
 - Individual characteristics
 - Anatomie, mental condition, concentration...
 - Environmental conditions
 - Utensils (bags, etc...), interaction with other subjects ...



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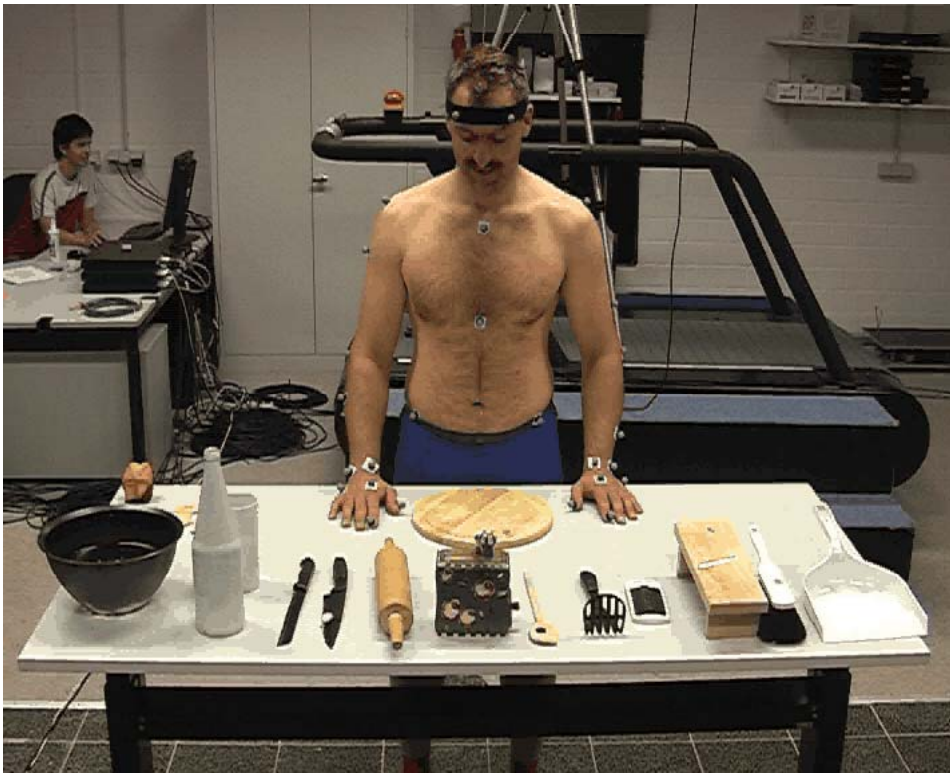
For stochastic modelling → Lots of training data needed



Solution - Motion segmentation

- Solution

- Segmentation of motion sequences into smaller motion units
- Modelling of motion units (motion phases)
- Recognition of motion sequences by concatenation of motion phases

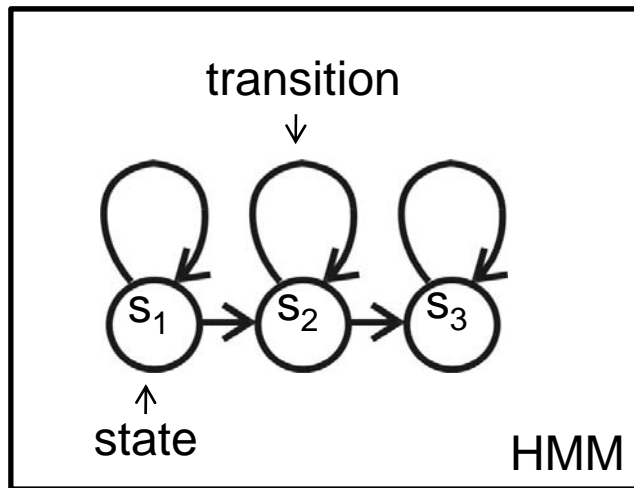


- idle_position
- picking_glass
- idle_position
- picking_bottle
- idle_position_bottle
- pouring
- putting_away_bottle
- idle_position
- putting_away_glass
- idle_position



Example: HMM for picking_bottle

- HMM for motion phase „picking a bottle“



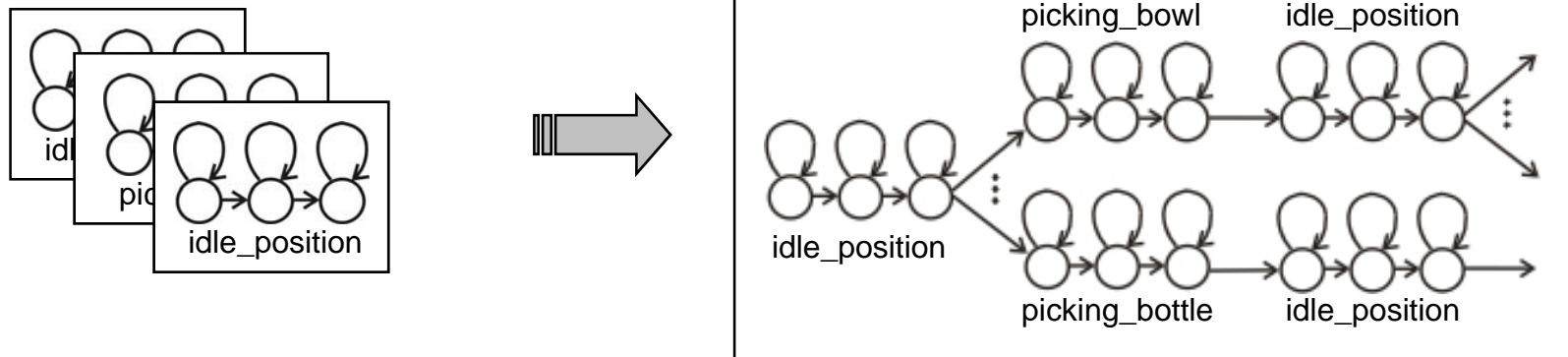
states

- s_1 : move hand towards the bottle
- s_2 : grasp the bottle
- s_3 : take bottle to target position

- Linear 3-state model
- Every state models on phase of the picking process

Recognition of continuous motion

- Context-free grammar

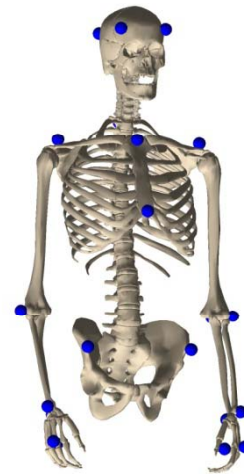
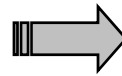


- Context-free grammar
 - Generated by experts (knowledge)
 - Recognition of limited set of sequences

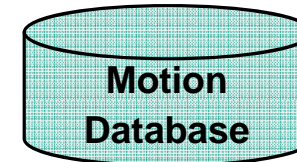
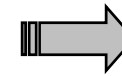
Motion Recognition System



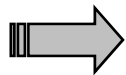
Live tracking by VICON



Mapping on human model



Online motion recognition with Hidden Markov Models

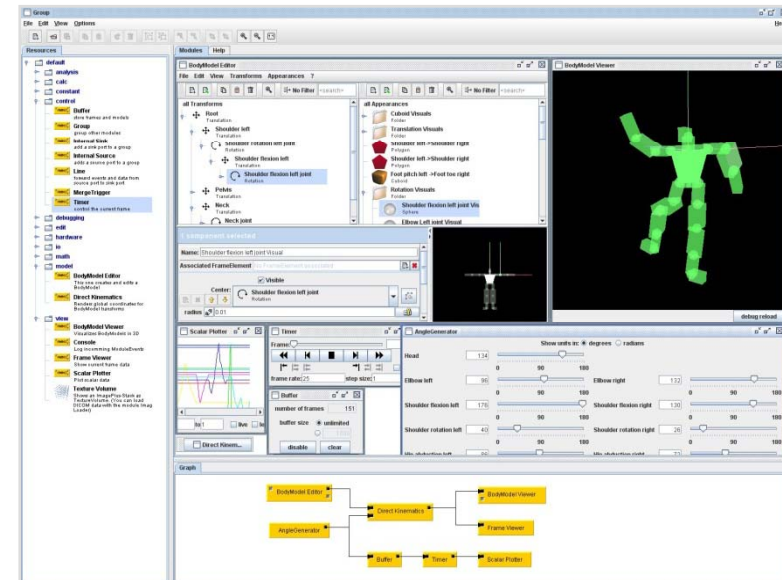


Recognition of user motion \rightarrow e.g. „stirring“



Online Motion Recognition

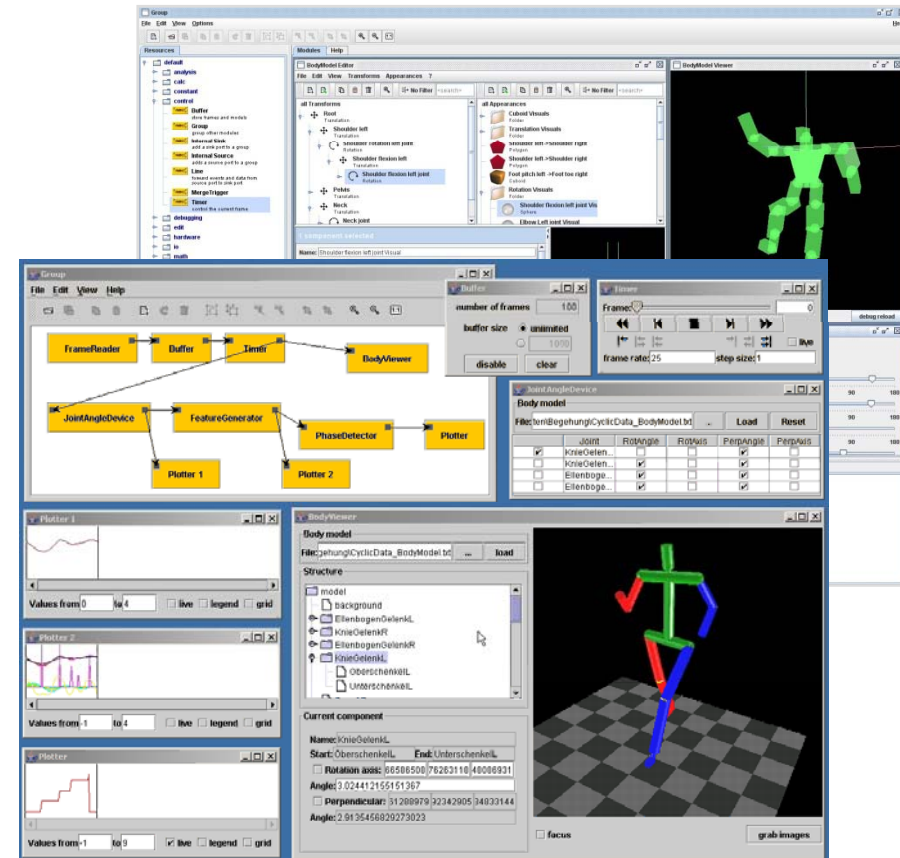
Visualisation, Analysis- and Modelling System



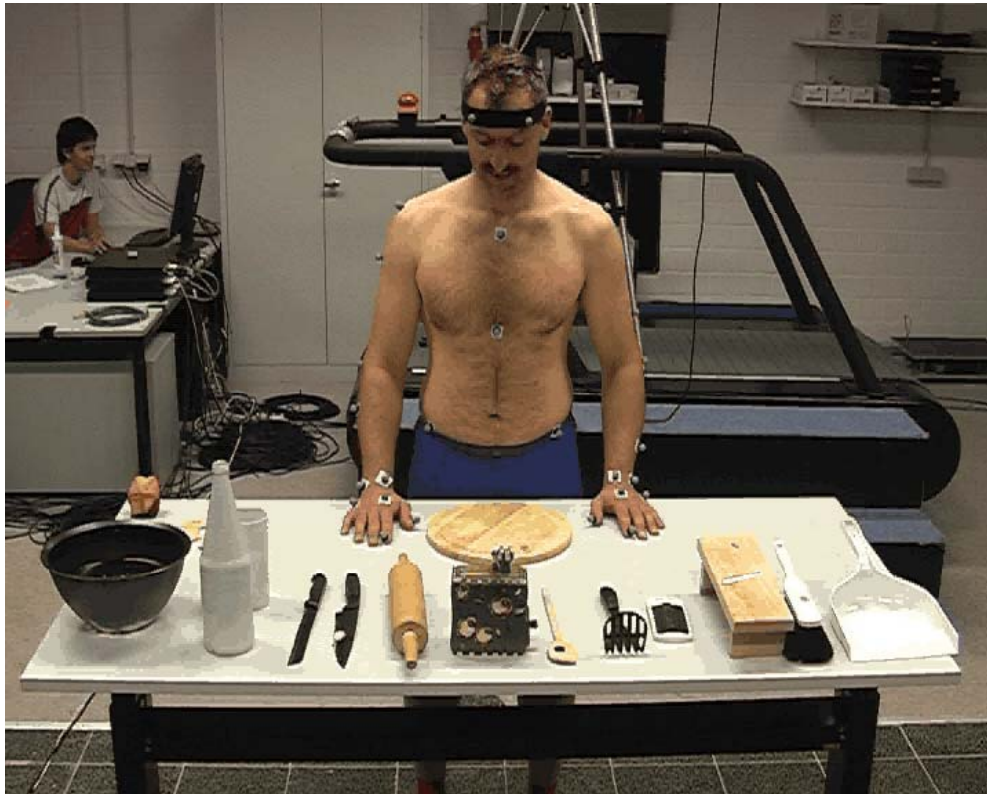
Online Motion Recognition

Visualisation, Analysis- and Modelling System

- Application
 - Signal processing
 - Data pre-/post-processing (Filtering, smoothing, signal transformationen)
 - Visualisation
 - Acquisition and Analysis of human motion



Motion data for CRC 588



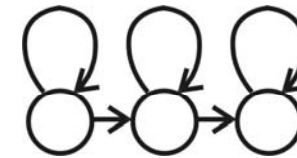
Training data

- 3 subjects
- 10 typical kitchen moves
 - rolling pastry
 - pouring water
 - planing apple
 - grinding coffee
 - sweeping
 - grating apple
 - stirring
 - cutting cake
 - cutting apple
 - pitching



Motion recognition for CRC588

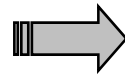
- Online motion recognizer
 - Development framework: Janus Recognition Toolkit
 - HMMs for 49 motion phases (idle_position, picking_bottle, etc. ...)
 - Each HMM has 3 states
 - HMM-Topology: left-to-right
 - Motion representation by a context free grammar
- Motion data
 - 1 subject
 - 600 motion sequences (10 different types) (500 training, 50 development, 50 test data)
 - 24 normalized joint angles



Connection Vicon - Vamos



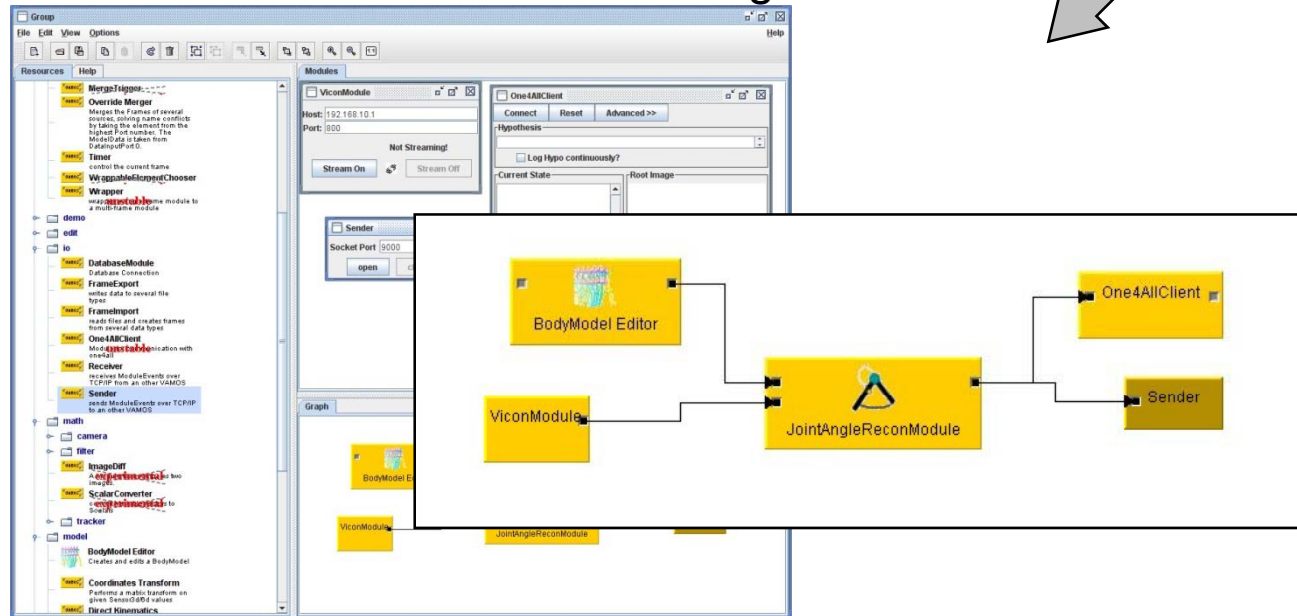
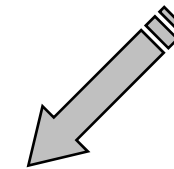
Marker



Tracking



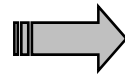
Vicon Server



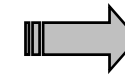
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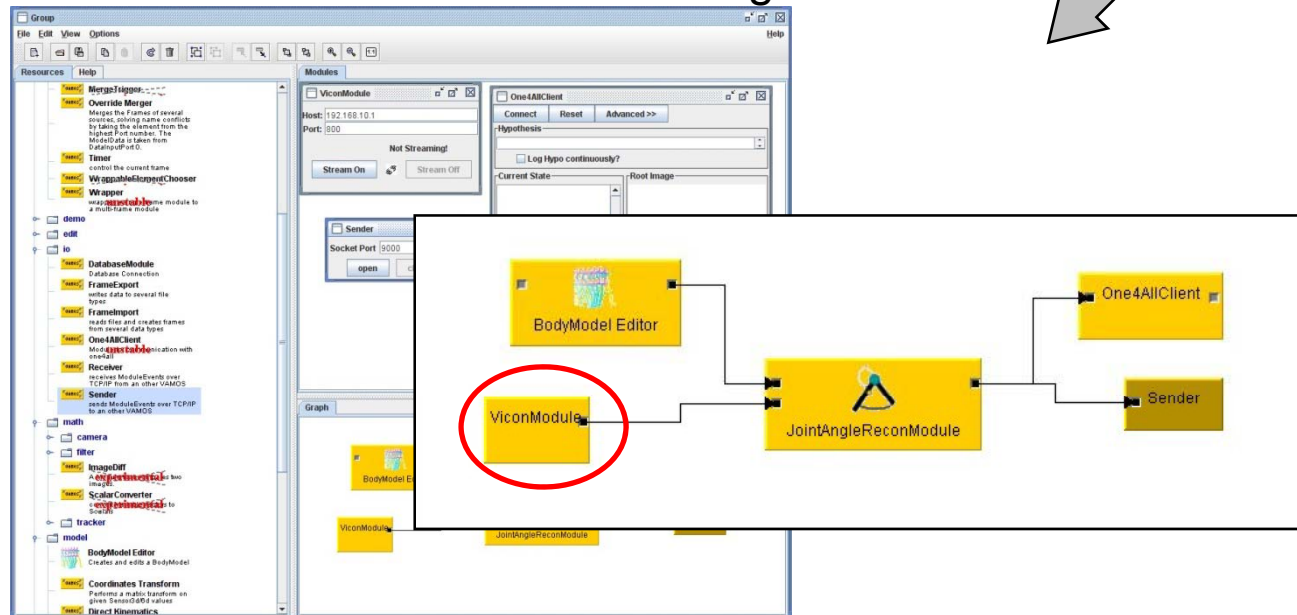
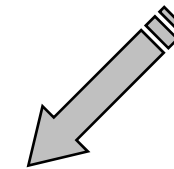
Marker



Tracking



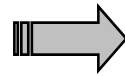
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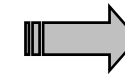
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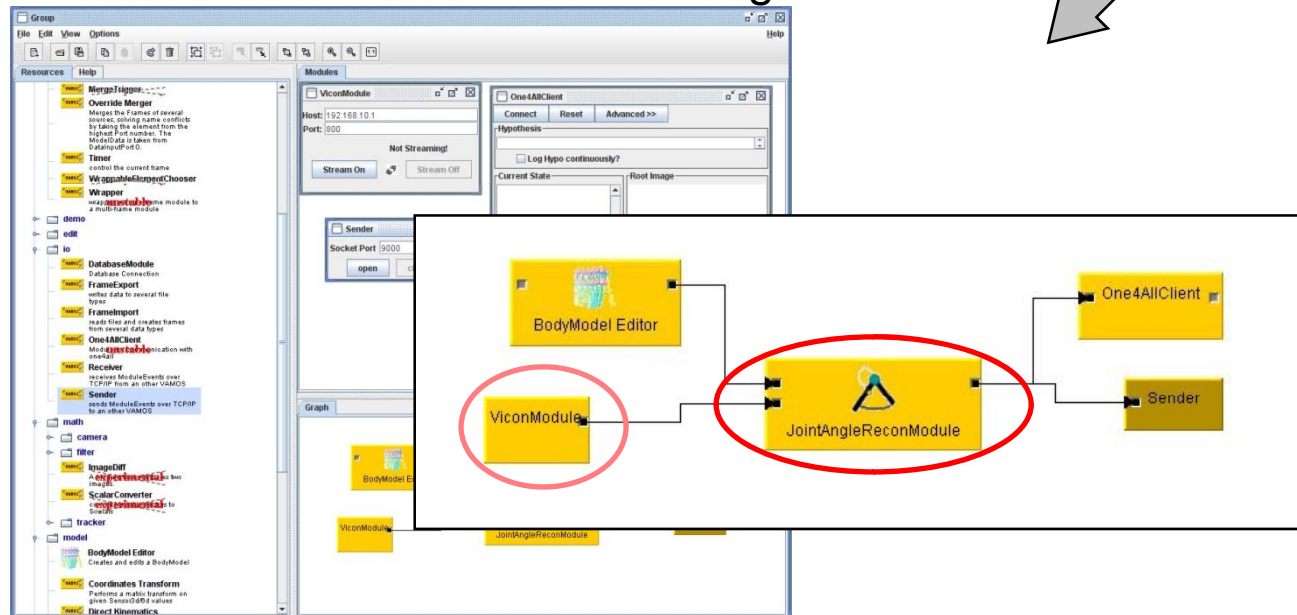
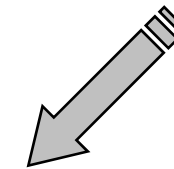
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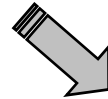
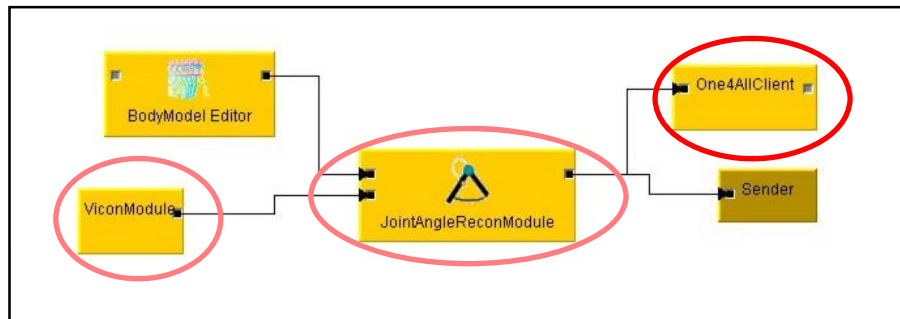
Tracking



Vicon Server



Example: Online Recognition of movements



The screenshot shows two windows: 'One4AllClient' and 'HMMBarGraph'. The 'One4AllClient' window has buttons for 'Start', 'Reset', and 'Advanced >>'. It contains a 'Hypothesis' field, a checkbox for 'Log Hypo continuously?', a 'Current State' field, and a 'Root Image' field. The 'Root Image' field displays the text 'Nothing recognized'. The 'HMMBarGraph' window displays a list of movements: 'Einschenken', 'Teig_auswellen', 'Kuchen_schneiden', 'Apfel_hobeln', 'Apfel_schneiden', 'Apfel_reiben', 'Kehren', and 'Stampfen'. At the bottom, there are controls for 'grid' (min value: 0.0, max value: 1.0), 'data' (checkboxes for 'show bars', 'show means', 'show bounds'), and 'control' (checkbox for '#frames to take short Means from:', '#frames to take long Means from: 10', and a 'Clear' button).



Results

- Error rate for motion phases: 5,5%
→ With CFG over 10 different motion sequences and 50 data sets
- Correct recogniton of all motion sequences
- Runtime ~ 20 fps



Future work

- Motion recognition independent of the individual person
→ Training with more than one person



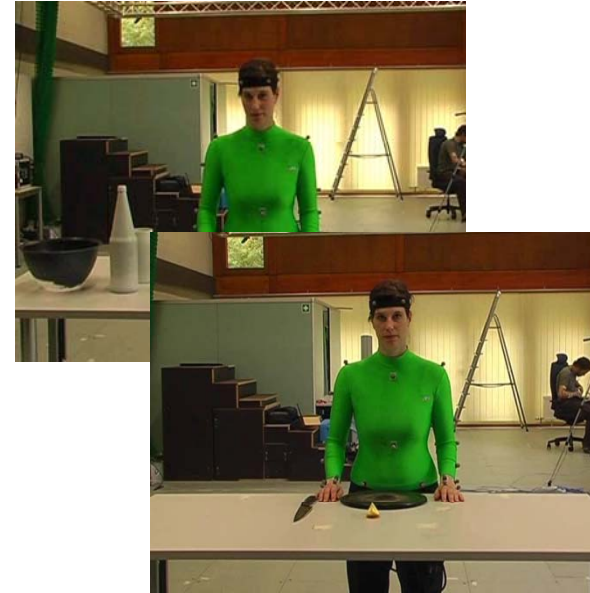
Future work

- Motion recognition independent of the individual person
→ Training with more than one person
- Natural movements:
 - Flexible object positionen
 - Fluent motion and motion variations
- Coordinated and concurrent movements



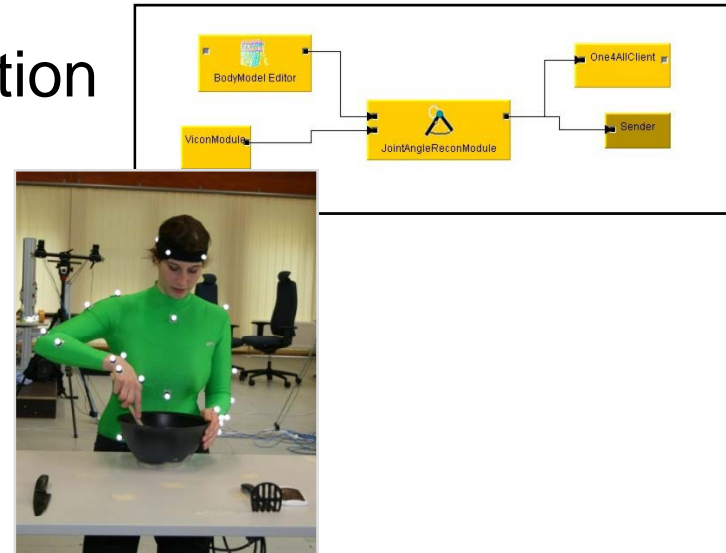
Future work

- Motion recognition independent of the individual person
→ Training with more than one person
- Natural movements:
 - Flexible object positionen
 - Fluent motion and motion variations
- Coordinated and concurrent movements
- Adaption of motion recognition to the robots visual system



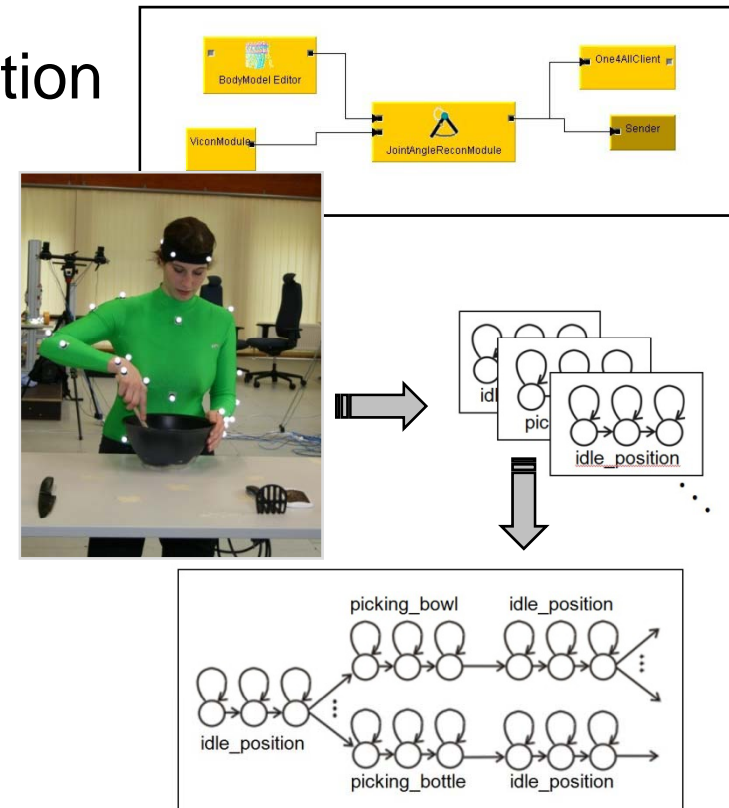
Conclusion

- Online motion recognition application
- Usage of marker-based tracking



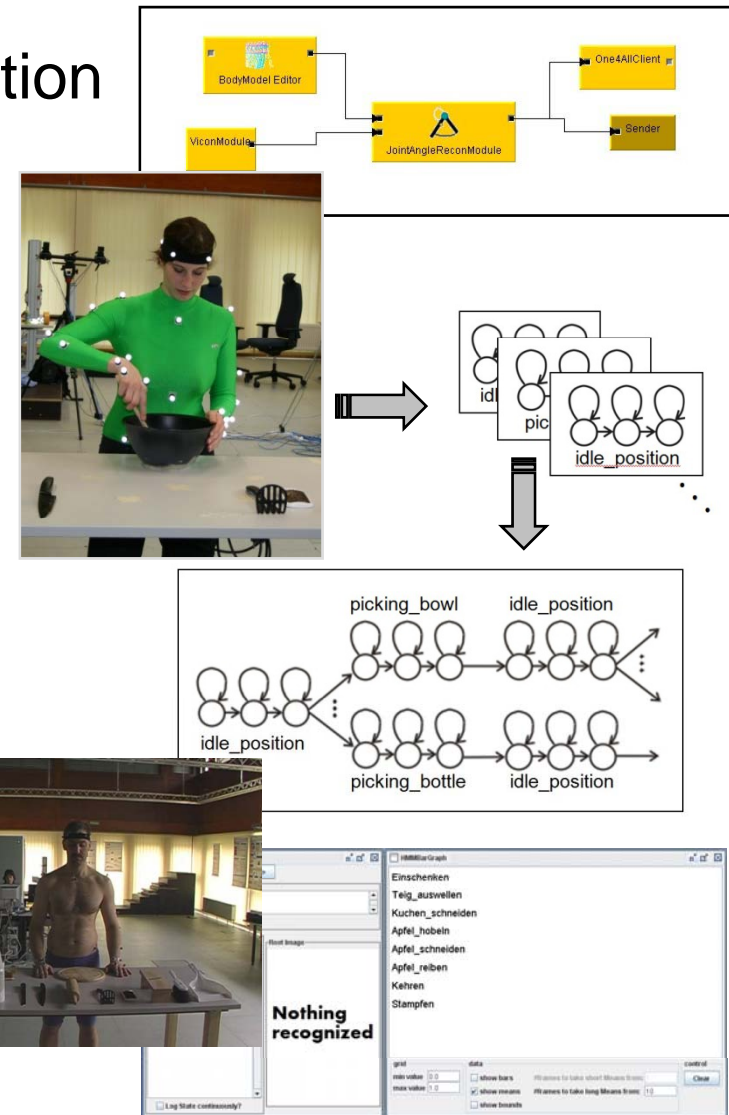
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- Motion recognition with HMMs
- Temporal motion segmentation
- CFG for motion sequences



Conclusion

- Online motion recognition application
- Usage of marker-based tracking
- Motion recognition with HMMs
- Temporal motion segmentation
- CFG for motion sequences
- Over 90% recognition
- Recognition in almost ,real-time‘



Project group



M3 – Motion and Action models



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D. Gehrig



T. Schultz



H. Kuehne



A. Woerner

Institute for Algorithms and Cognitive Systemes



Thank you!

