

# Haptic Interaction in Human-Human and Human-Robot Dyads

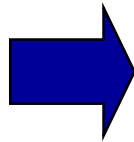
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# Towards Human-Centred Robotic Systems

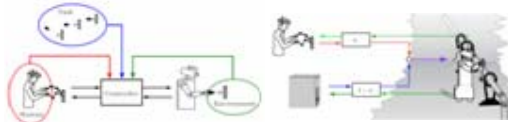


# Human-Centred Robotic Systems

## Telepresence and Teleaction Systems



Advanced teleoperation controllers



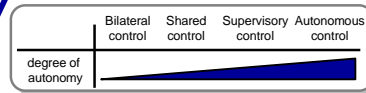
Single and multi-user scenarios



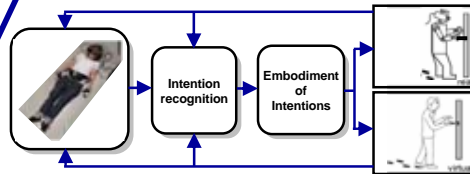
Intention recognition from physiological signals



Automatic shifting of robot autonomy



Embodiment of intentions and personalization of actions



Brain and Body Computer Interfaces

## Haptic Human-Robot Interaction



Modelling of human motor behavior



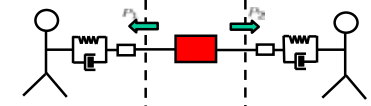
fMRI and TMS studies with Max-Planck Institute



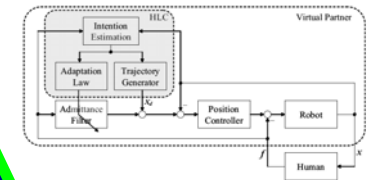
Human Motor Control



Analysis of human-human interaction



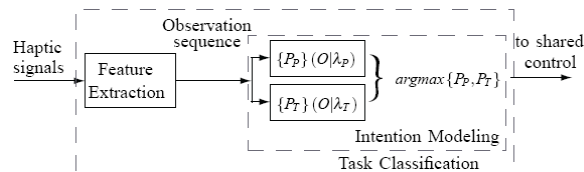
Synthesis of interactive robotic systems



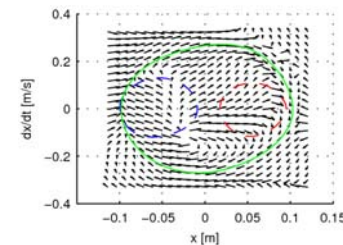
Design and control of human-system interfaces



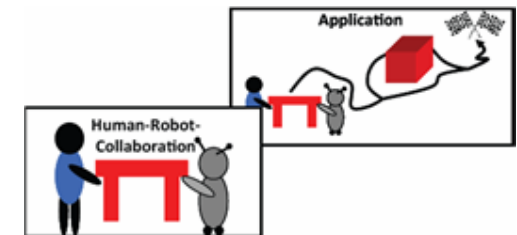
Intention recognition



Adaptation



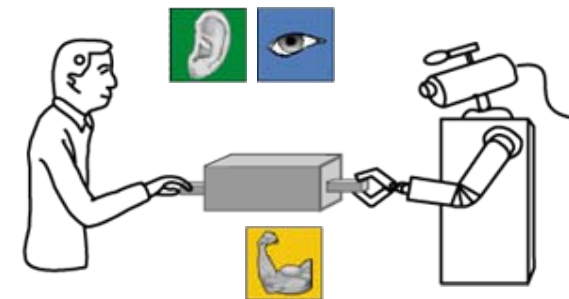
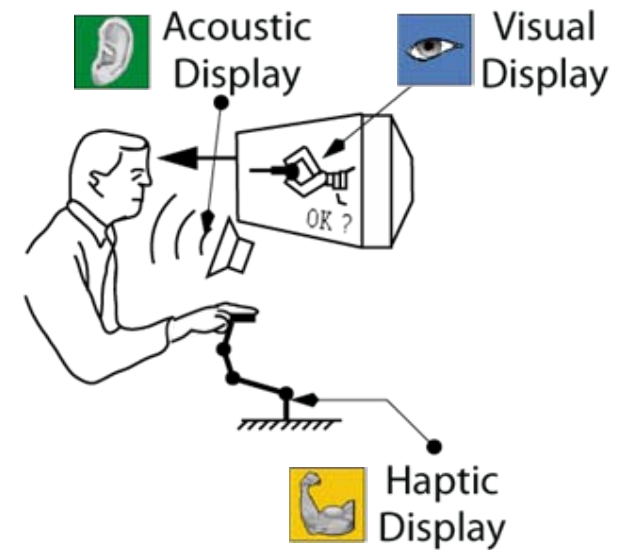
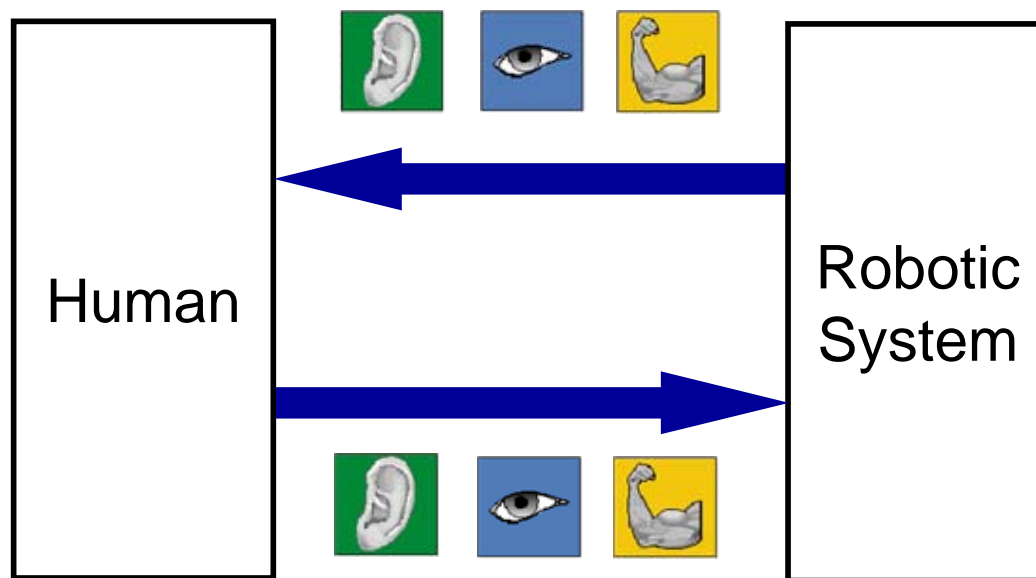
Joint decision making





# Haptics and other Modalities

**Haptics** refers to the sense of touch





# Haptic Interaction



Interaction with passive systems:

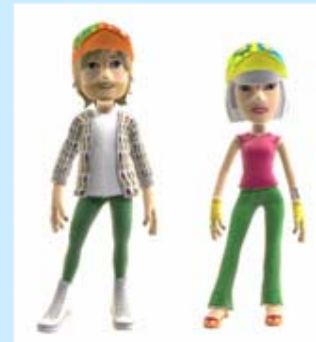


virtual



real

Interaction with active systems:



virtual



real

# Interaction Types

- Direct human-robot interaction

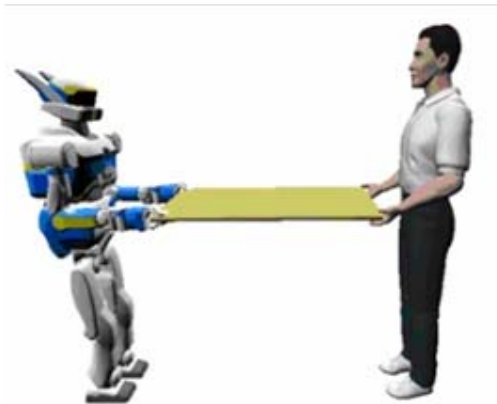


[Buss et al., 2009]



[Honda]

- Human-robot interaction mediated by an object



[AIST/JRL]



[Wojtara et al. 2009]

# Application Fields of Haptic Human-Robot Interaction

[Bidaud, 2006]



[Honda Walking Assist]

[Locomat, Hocoma]



Rehabilitation Devices

Walking Assistants

[Wojtara et al. 2009]



Assembling/Transporting Assistants

Haptic  
Human-Robot  
Interaction

[Kosuge 2005]



Social Interaction  
Partners



Sensorimotor  
Skill-Trainers

[Bergamasco et al. 2009]

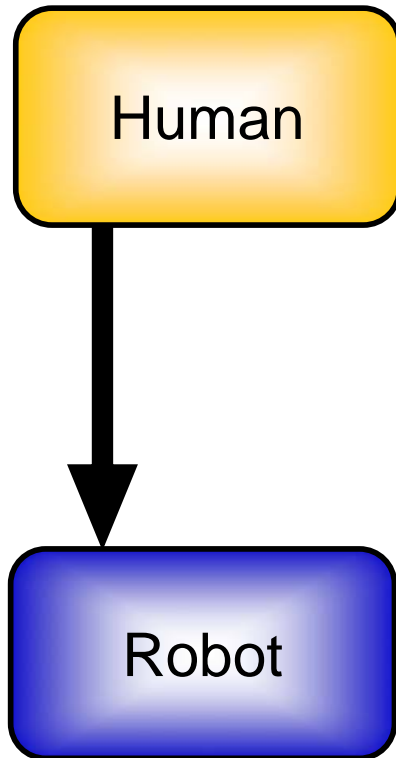


[Wang, Peer, Buss 2009]

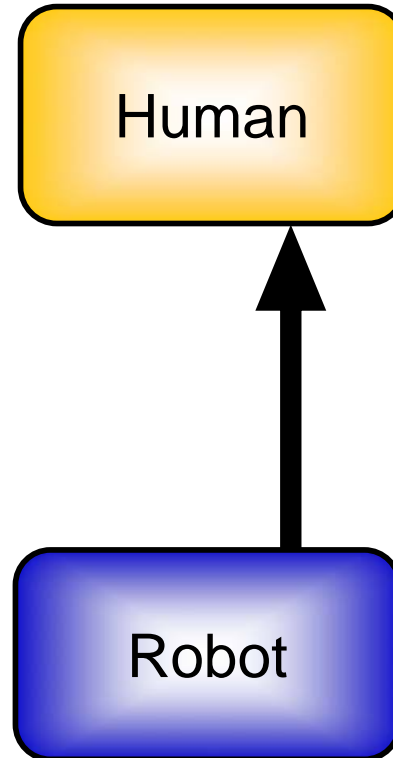


# Interactive / Proactive Robot vs. Fully Passive/Active Robot

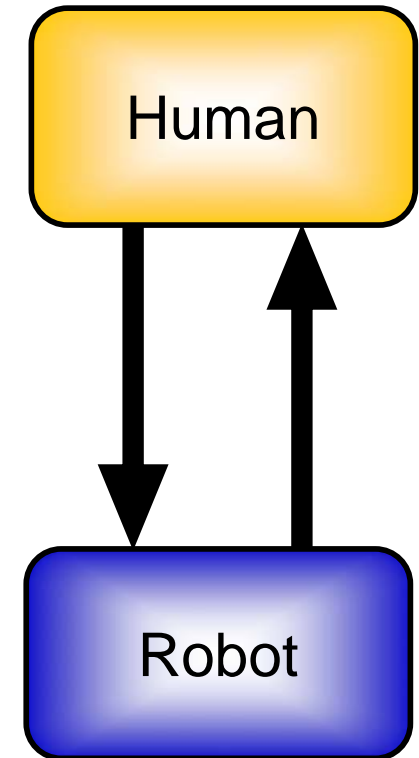
Fully passive robot



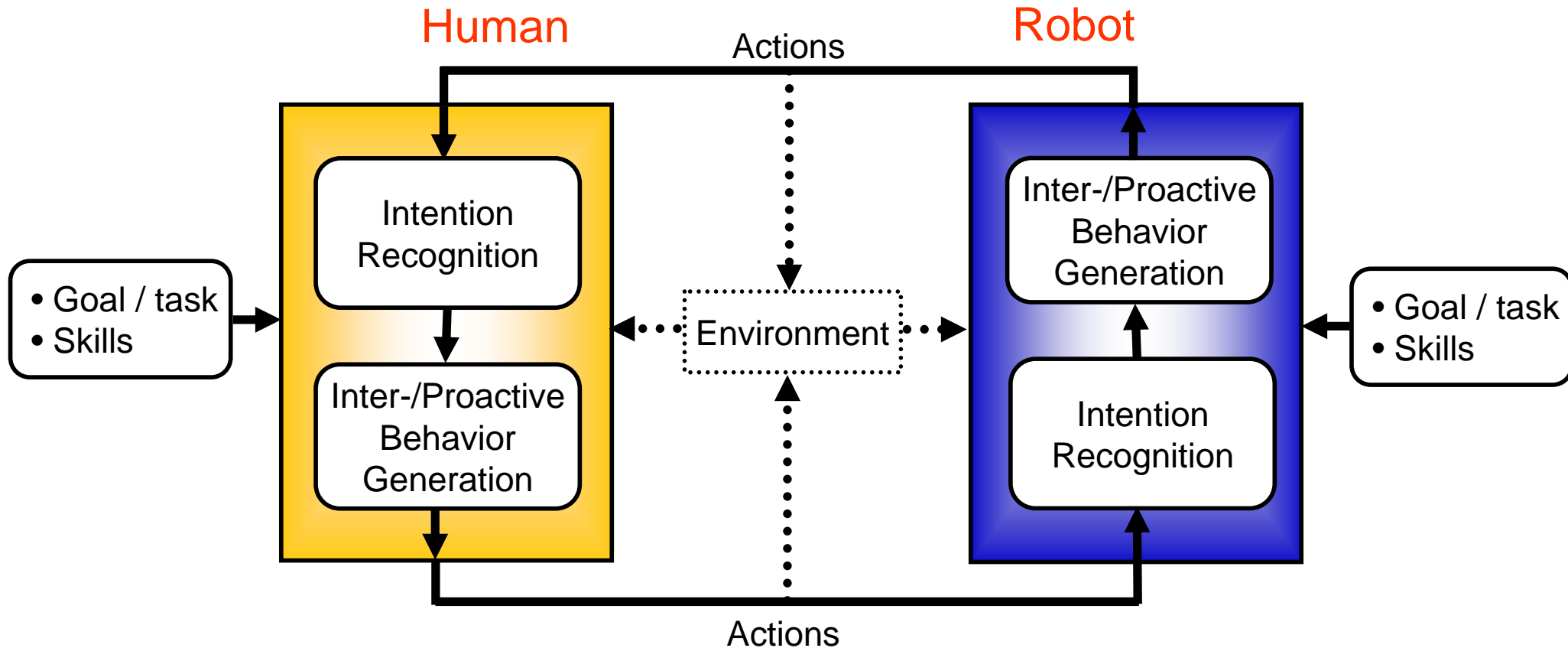
Fully active robot



Interactive/proactive robot



# Haptic Human-Robot Interaction



# State of the Art

## Human intention recognition:

- recognition of human intention to accelerate/decelerate by analyzing energy [Tsumigawa:2001] and force derivative [Duchaine:2007]
- recognition of task phase from motion/force data using production rules, NN, HMM [e.g. Kosuge:2007; Stefanov, Peer, Buss: 2009, 2010]

## Fully passive robots:

- Robots with constant impedance parameters [Kosuge:2000]
- Robots with human-like pre-recorded impedance characteristics [Rahman:1999]
- Robots with varying impedance parameters depending on human intention [Tsumigawa:2001, Duchaine:2007], human arm stiffness [Tsumigawa:2002], optimality functions [Ikeura:2002]
- Robots implementing virtual constraints [Arai:2000]

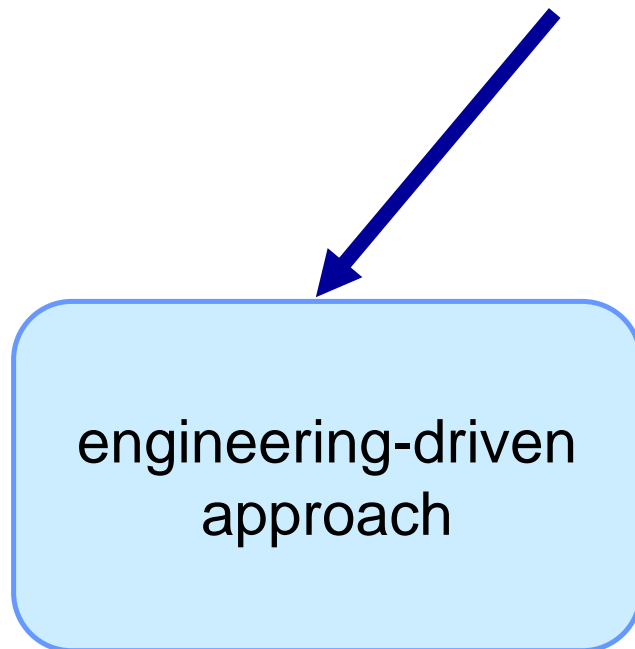
## Interactive/proactive robot behavior generation:

- Robots implementing minimum jerk model [Maeda:2001, Corteville:2007]
- Robots implementing role sharing strategies: accelerator/decelerator [Reed:2007], active/passive [Wang, Peer, Buss:2009, Evrard:2009], weight and DOF separation [Wojtara:2009]

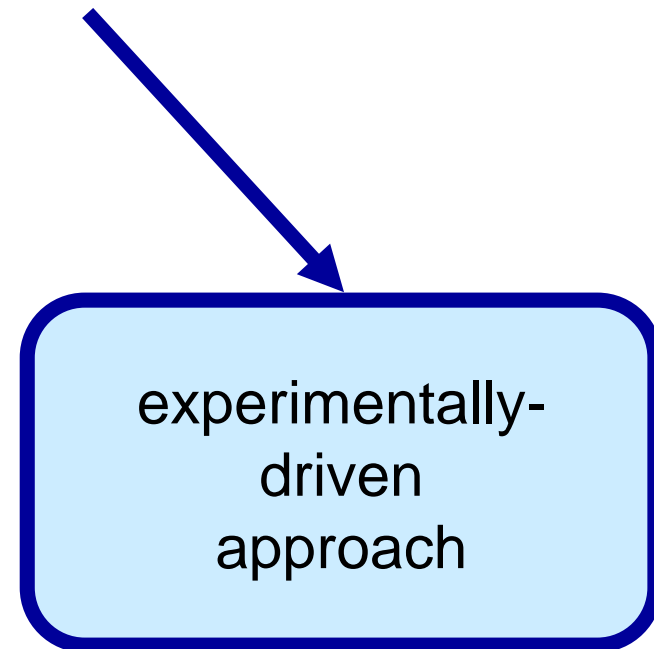


# Synthesis of Robot Behavior: Two Approaches

Synthesis of robot behavior



- performance-driven
- based on pre-knowledge or optimality criteria



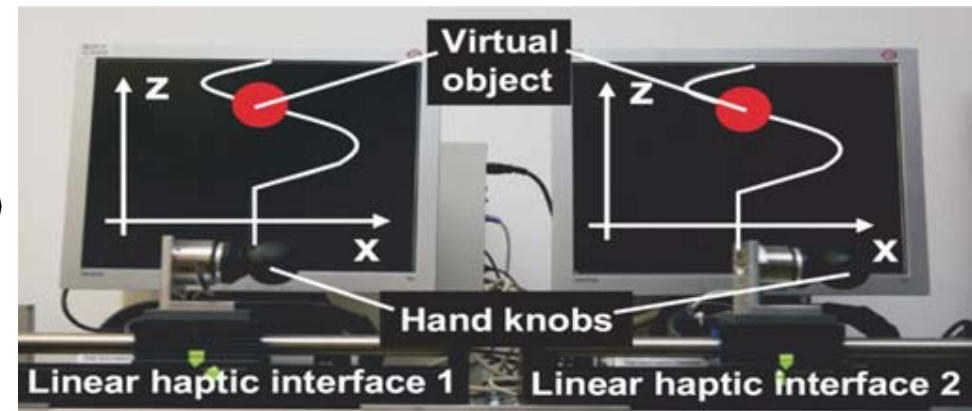
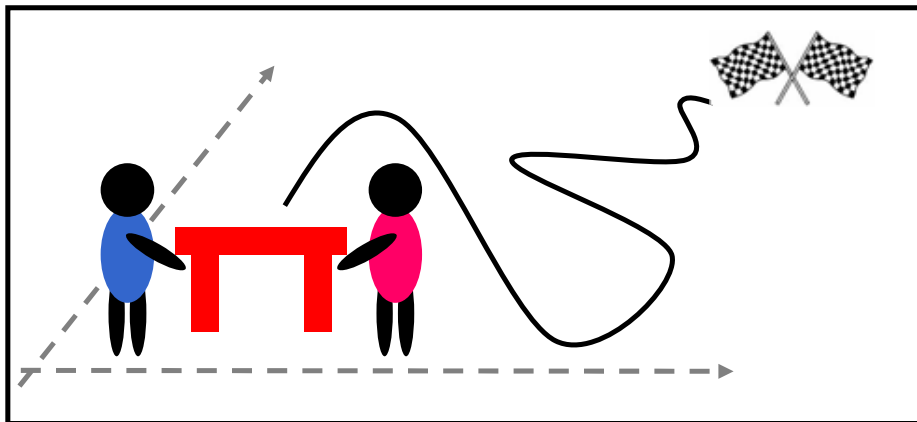
- human-centered
- based on experimental data

# Experimental Paradigm to Investigate Haptic Collaboration

## Research Questions:

- What influence has haptics on human-human collaboration?
- Do partners take over strategies?
- How do people adapt to each other?

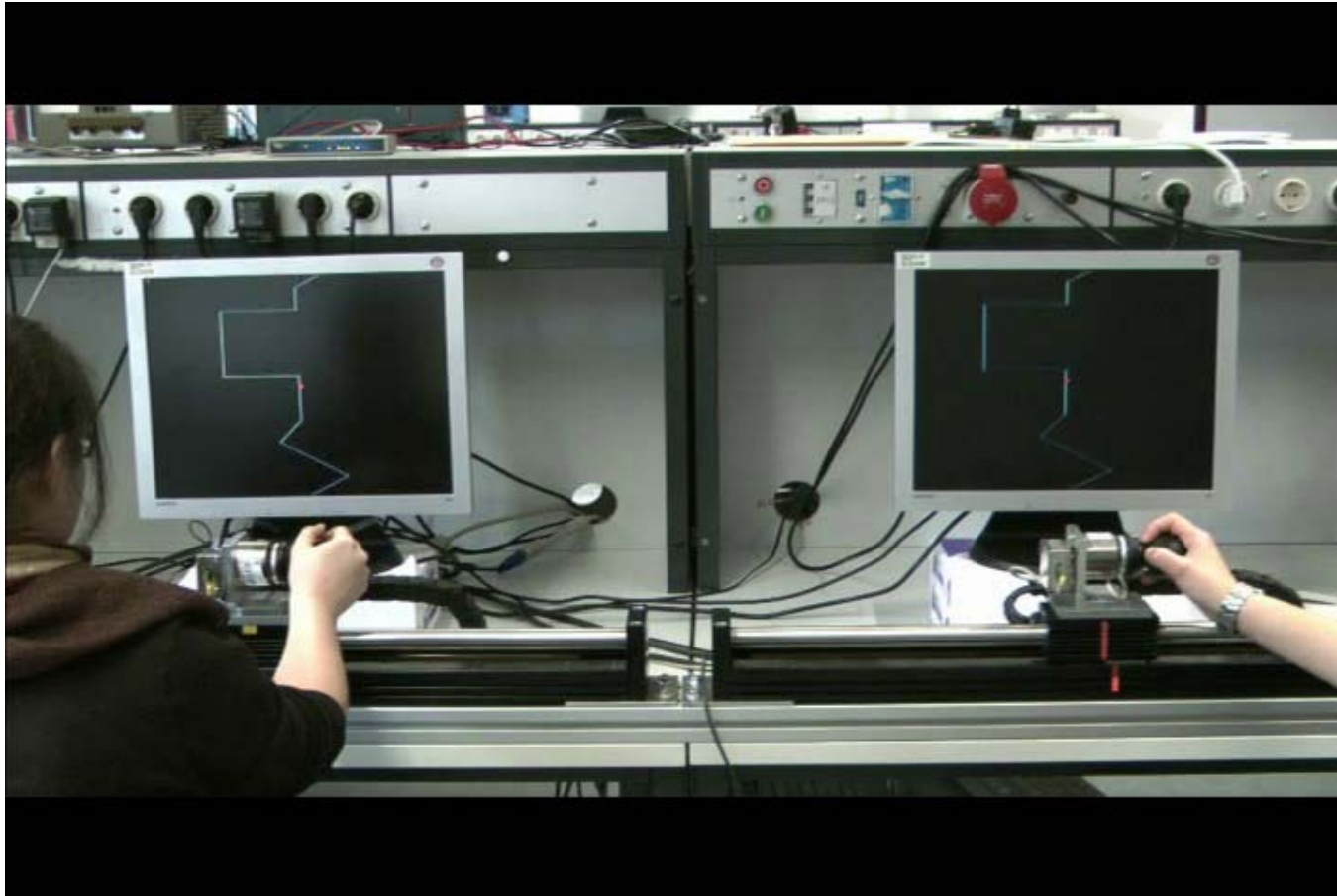
**Experimental paradigm:** Shared path following



## Investigated conditions:

- Individual and dyadic condition
- With and without haptic feedback

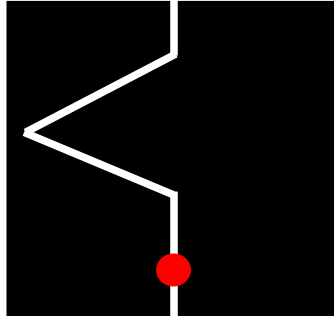
# Haptic Human-Human Collaboration





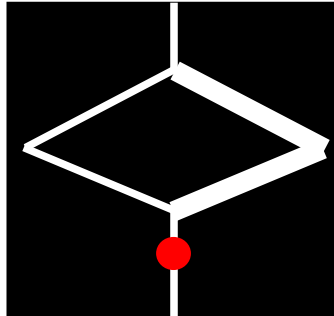
# Levels of Haptic Collaboration

**low-level of  
haptic  
collaboration**



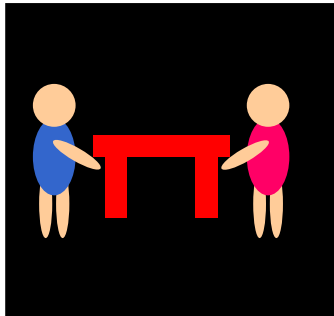
- 1DoF Pursuit Tracking Task
  - Goal: stay on the track
  - Action plan: follow the track
- **Control / adaptation / strategies**  
**Application: joint carrying / placing**

**high-level of  
haptic  
collaboration**



- 1DoF Pursuit Tracking Task
  - Goal: stay on the track
  - Action plan: follow one of the tracks
- **Decision making**  
**Application: obstacle avoidance**

**haptic  
collaboration in  
daily activities**



- More DoF
  - Different tasks
- **Generalization**

# Dominance in Haptic Interaction

[Groten/Feth/Peer/Buss, Ro-Man 2009]

**Dominance measure** for haptic interaction:

Individual

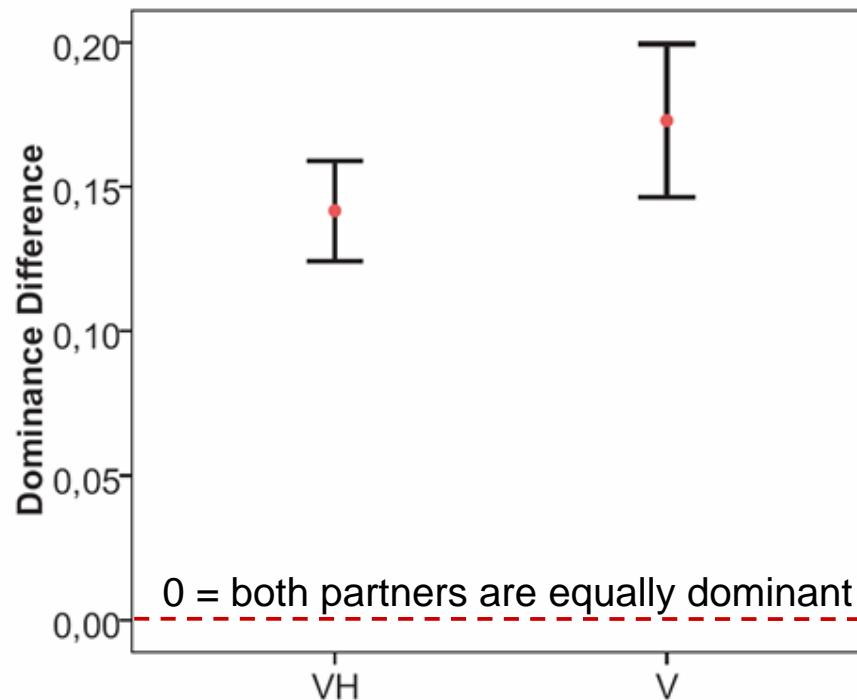
Dyadic

$$D_{12,i} = \frac{f_{1,i}^E}{f_{sum,i}} \quad \bar{D}_{diff} = |\bar{D}_{12} - \bar{D}_{21}|$$

- Time independent dominance: [.39, .61]

- One partner always domineers

- In both feedback conditions



## Results:

Analysis of behavioral consistency

- Consistency **across partners** for V = 49% and VH = 64%
- 32% in V and 24% in VH depend on **interaction**  
→ Adaptation
- Dyadic reciprocity: V= -.36; VH= -.52  
→ Dominance differences between partners varies in sub-trials



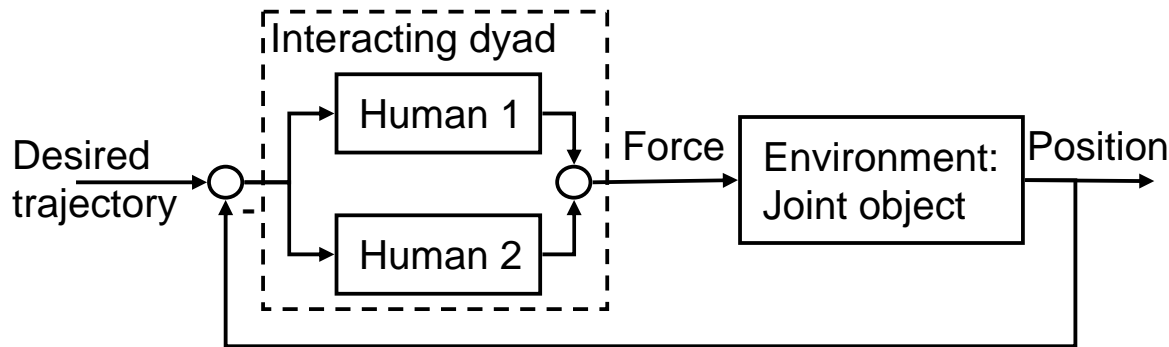
# Control-theoretic Model of Haptic Human-Human Interaction

[Feth/Groten/Peer/Buss, Ro-Man 2009]

Crossover model of single human's behavior [McRuer/Jex, 1967]:

$$G_h(s) = \underbrace{\frac{e^{-\tau s}}{(1-T_p s)}}_{\text{Perception/action loop}} \underbrace{[K(1 + T_z s)]}_{\text{Human control action}}$$

→ Extension to interacting dyad:



## Results:

- Single operator
- Interacting dyad → Mutual adaptation
- Individual within interacting dyad → Future work



# Lessons Learned

- Haptic feedback leads to higher **predictability** of behavior (consistency analysis)
- With haptic feedback the **communication** between partners is supported (performance in decision making)
- Control architectures for haptic collaboration should provide **unequal dominance** values for the involved partners (dominance analysis)
- Humans mutually adapt to each other → **adaptation capabilities** of technical partner required

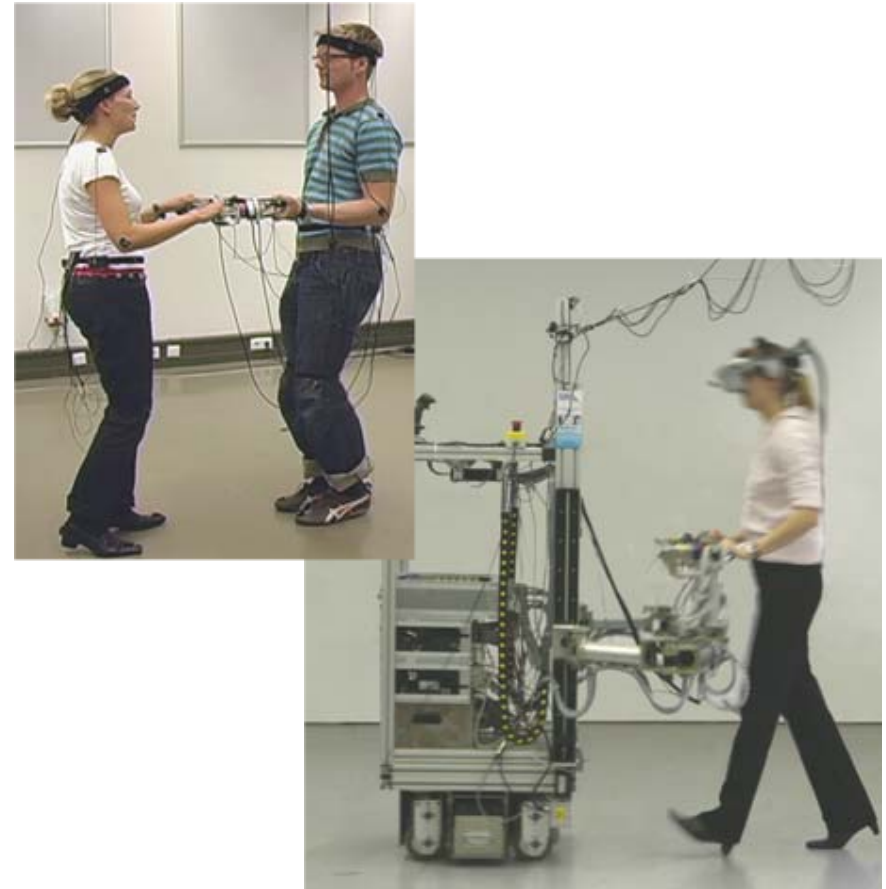


# Two Prototypical Scenarios

## Human-robot handshaking



## Human-robot dancing



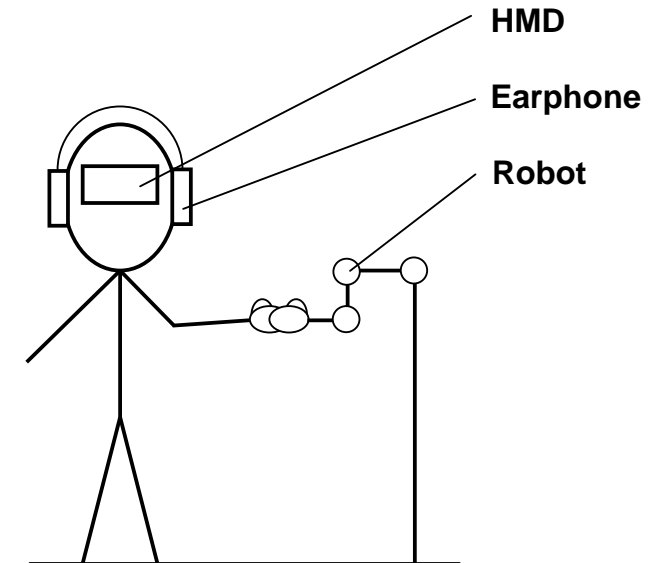
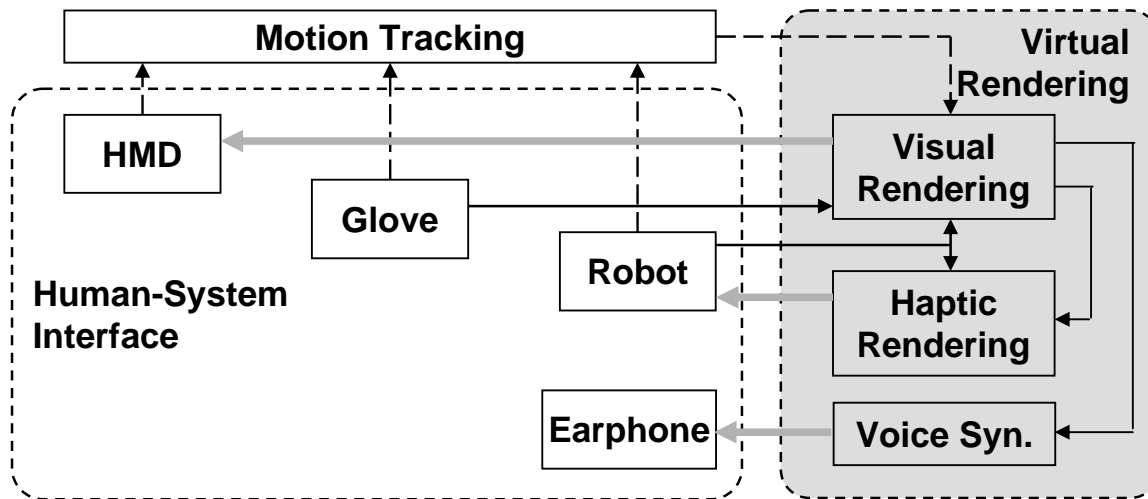
# Human-Robot Handshaking

## Overall goal:

- Replace one human in handshaking with a robot
- Multi-modality: haptics, vision, audio
- Recreate handshake arm dynamics
- Provide the human participant with a realistic handshaking experience

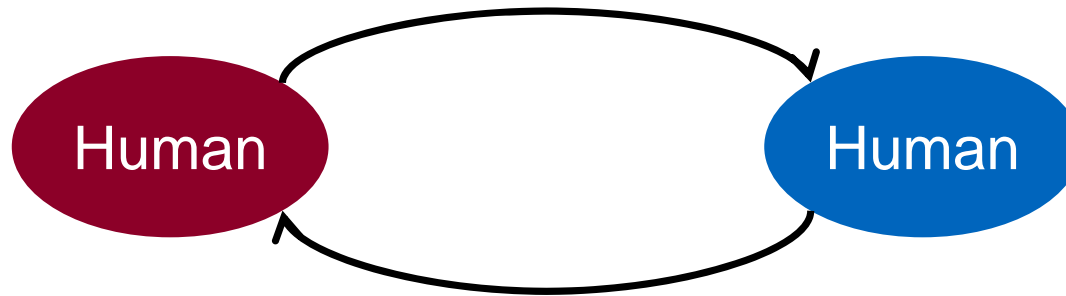


[Wang, Lu, Peer, Buss, EH2010]

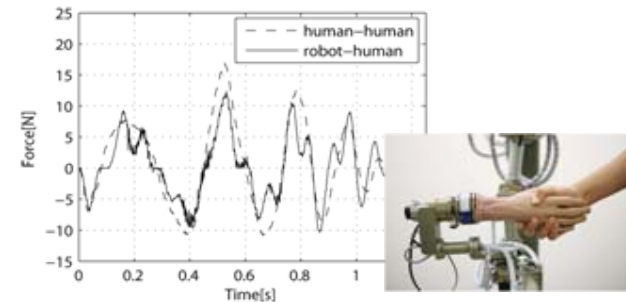
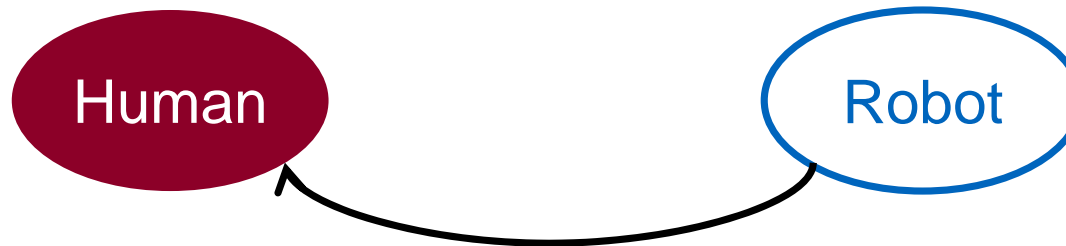


# Our Approach: Record – Replay – Recreate

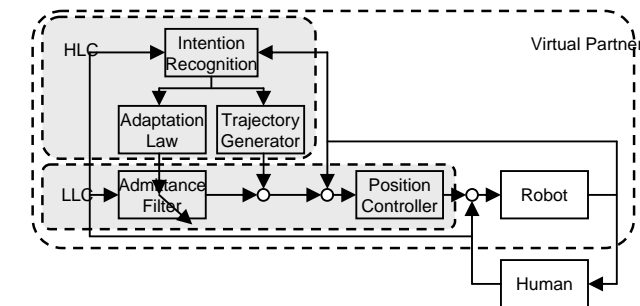
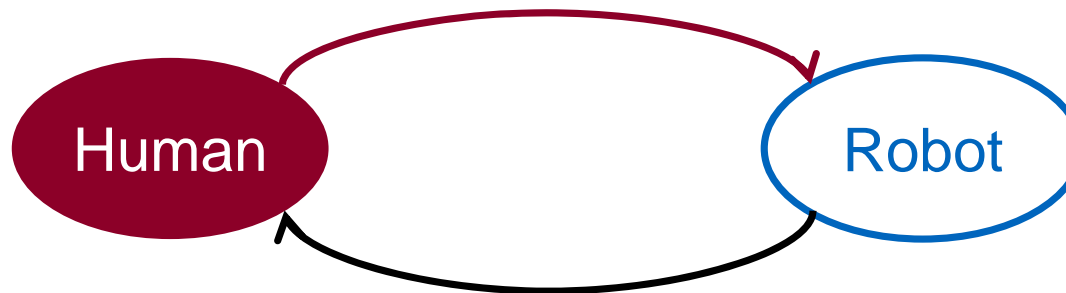
Record



Replay

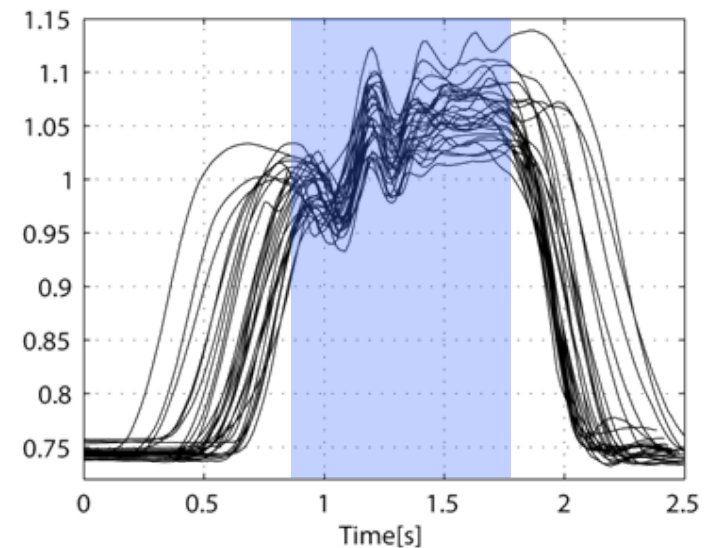
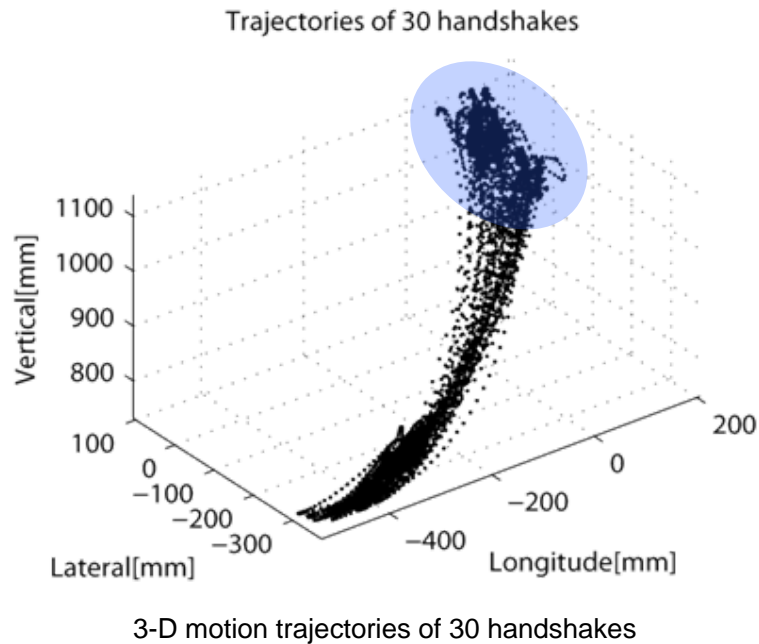
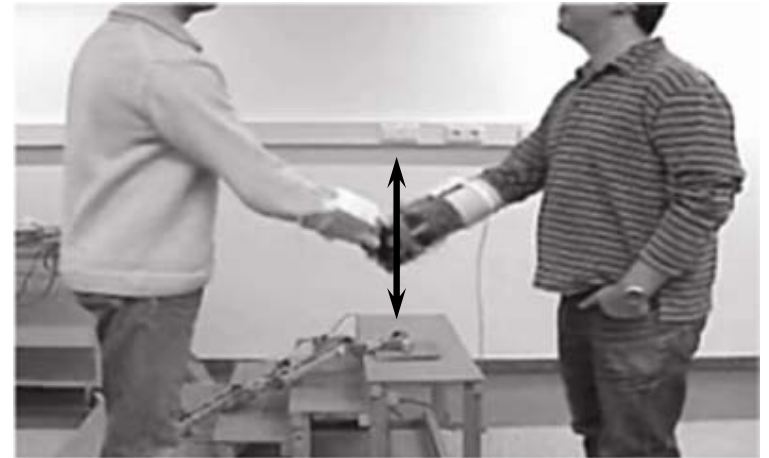


Recreate



# Recording Handshakes

- 900 human-human handshakes
- similarity and variance
- 3 stages: approach, **shake**, release



# Replaying Handshakes

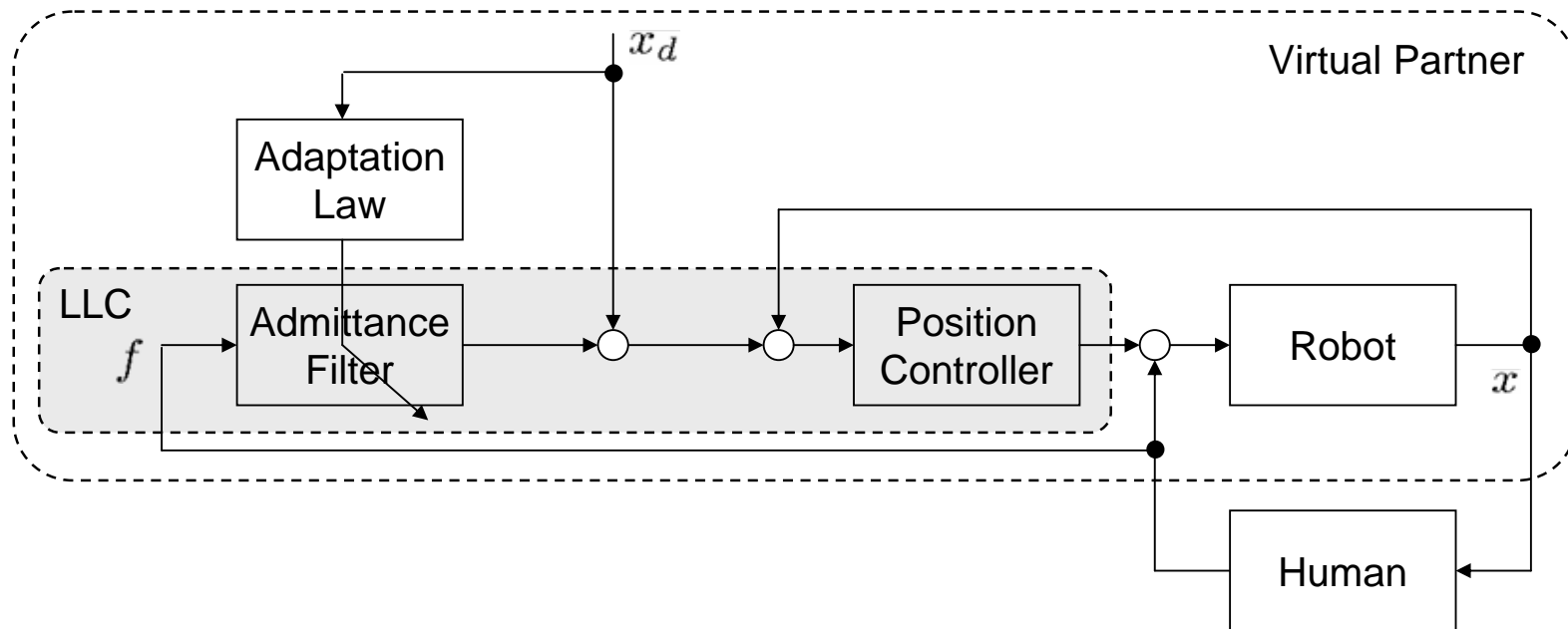
[Wang, Peer, Buss, WHC2009]

## Replay strategy:

- Replay of pre-recorded trajectories  $x_d$
- Admittance filter to provide compliance, parameters adjustable with respect to the reference trajectory

$$f(t) = M(t)\ddot{x}(t) + B(t)\dot{x}(t) + K(x_d)(x_d(t) - x_0)$$

$$K(x_d) = K_0 + \lambda(x_d(t) - x_0)$$

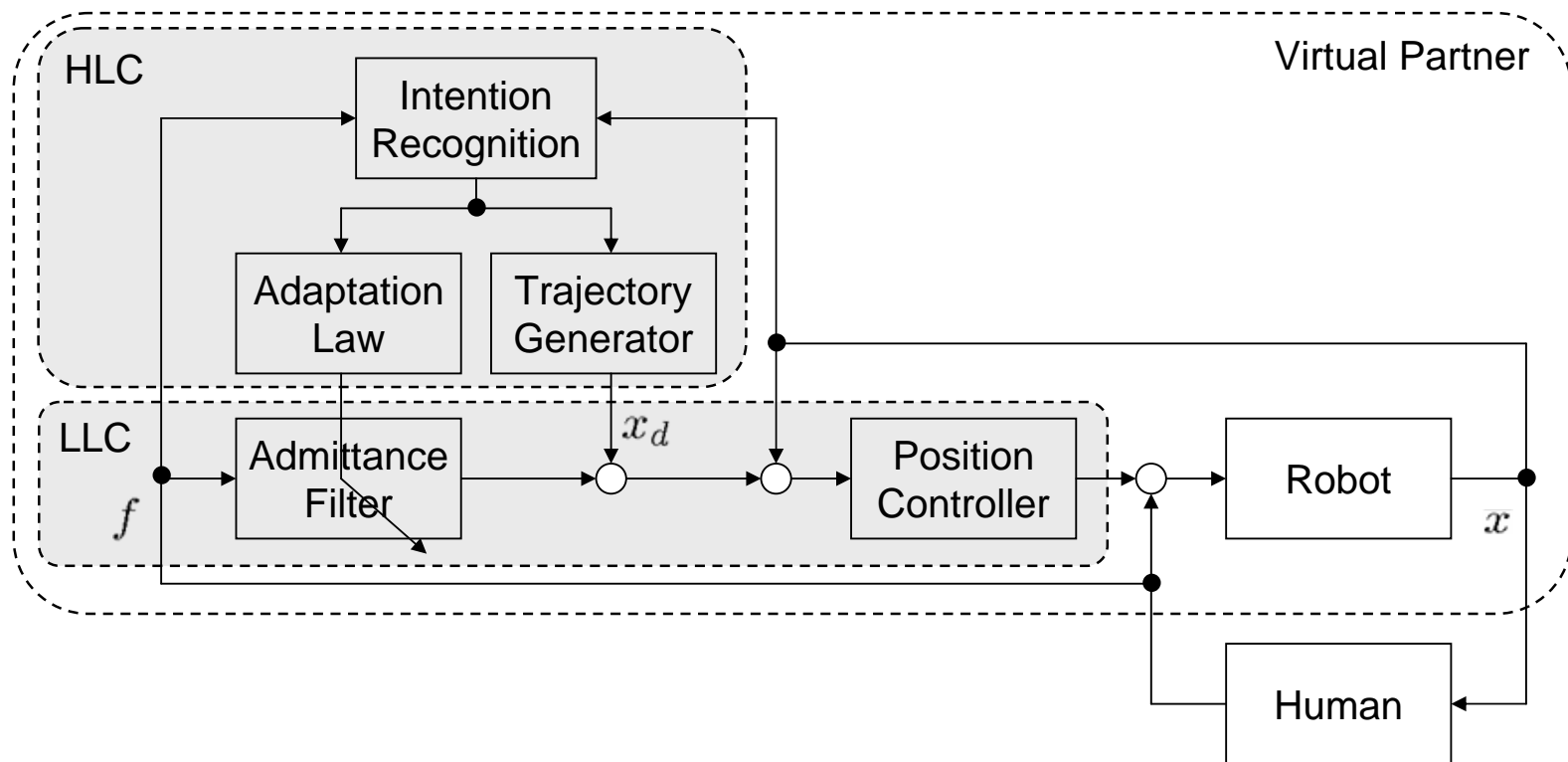




# Recreating an Inter-/Proactive Handshake Partner

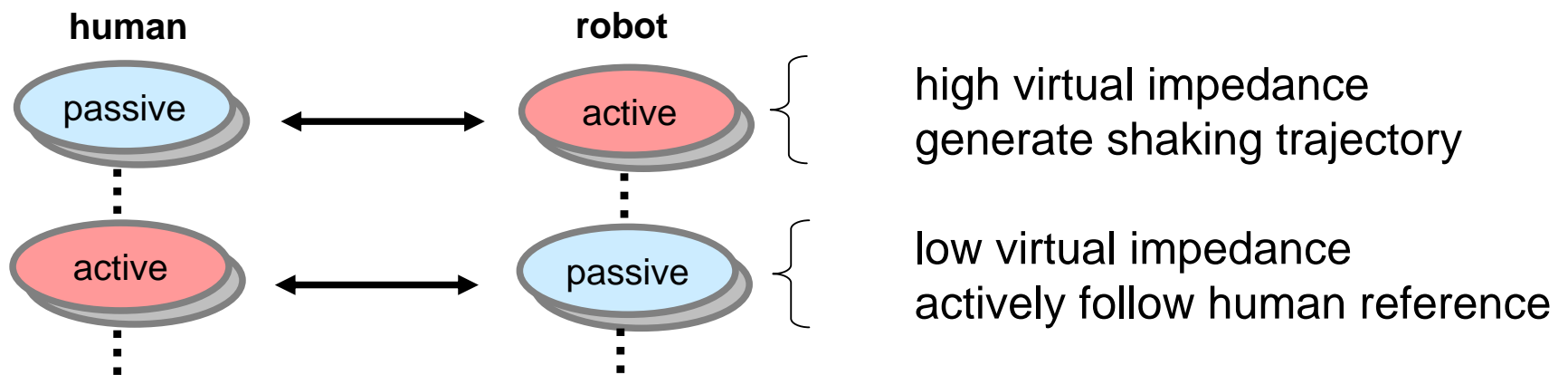
[Wang, Peer, Buss, ICM2009; WHC2009]

- Implementation of additional **high-level controller**  
→ Trajectory generation / impedance adaptation based on human intentions



# Recreation Strategy

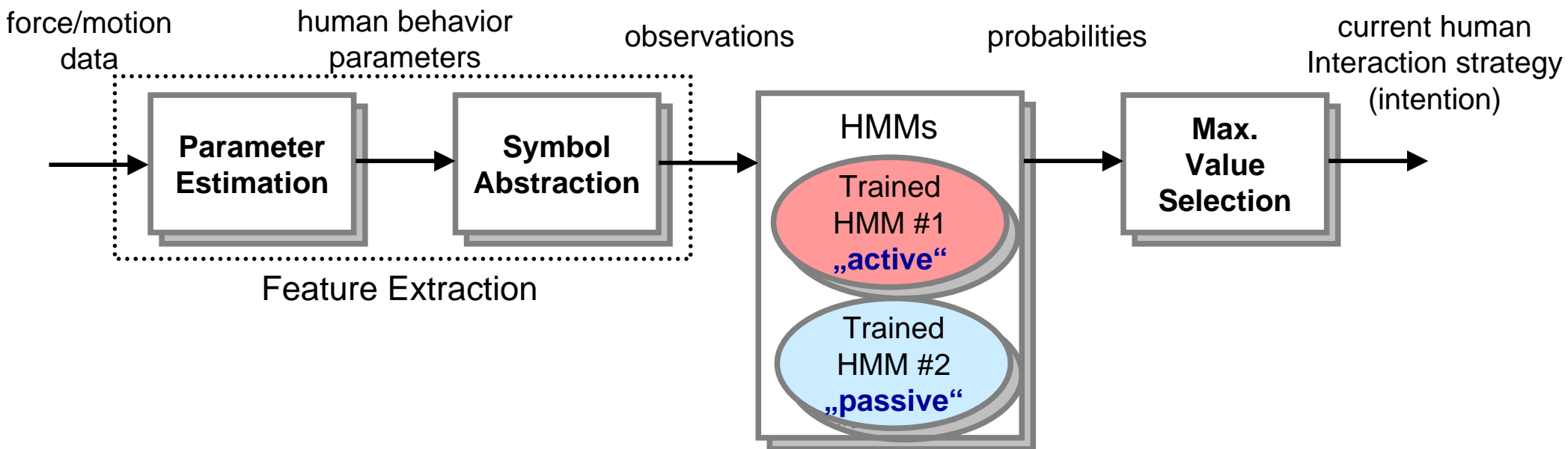
- Partners can select between **two different strategies** when performing a handshake:
  - 1 „active“ → take over the lead and decide on the joint trajectory
  - 2 „passive“ → follow as best as possible the lead of the partner
- Interacting partners take over **opposite roles**



- Timely combination of both strategies** allows to create arbitrary types of handshakes

# Intention Recognition Module

- Classifier based on Hidden Markov Models (HMMs)
- Two trained **HMMs “Active” and “Passive”**
- Symbolized human behavior parameters as HMM observations
- Select the HMM with maximum likelihood to determine the current human interaction strategy (intention)



# Trajectory Generator and Adaptation Law: Passive Human

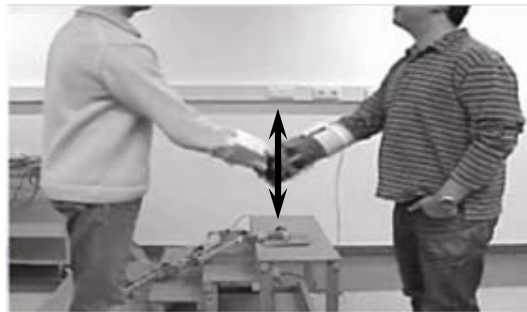
## Trajectory Generator:

- Damped sinusoidal function for typical handshake
- Parametrization according to results obtained in human experiments

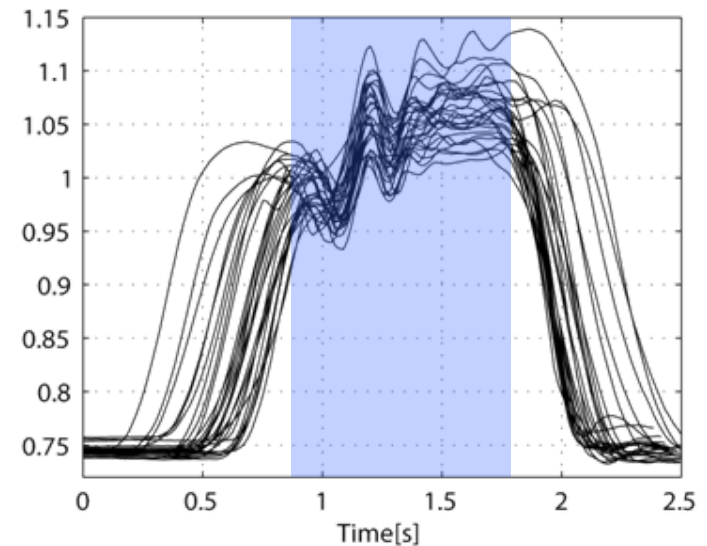
Position	Min	Max	Mean
Duration/s	0.712	1.816	1.01
Amplitude/m	-	0.161	0.098

## Adaptation Law :

- no adaptation



Vertical-time motion trajectories of 30 handshakes



# Trajectory Generator and Adaptation Law: Active Human

## Trajectory Generator:

- Trajectory defined by turning points  $A$  and  $E$ , and segment lengths  $D$

## Adaptation Law (minimization of interaction force):

- Online adaptation of trajectory to human partner depending on interaction force  $f$

$$A_{n+1} = dir \cdot \beta_1(\beta_2 + f_t)(1 - \sin \frac{\pi t}{2D_n}) + x_t$$

$$E_{n+1} = -dir \cdot \beta_1(\beta_2 + f_t) \sin(\frac{\pi}{2}(\frac{t}{D_n} + 3)) + x_t$$

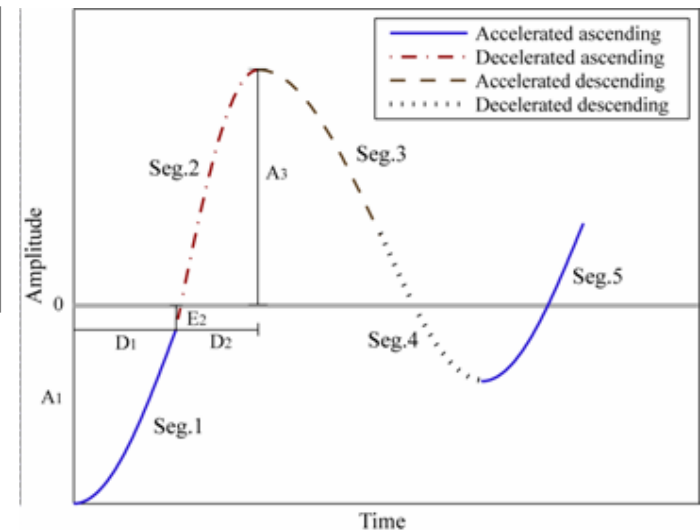
$D_{n+1}$  starts if force changes direction

$D_n$  ... segment length, initialized with 250ms

$\beta_1, \beta_2$  ... tunable parameters

$f_t, x_t$  ... currently measured interaction force and position

$dir$  ... direction of motion

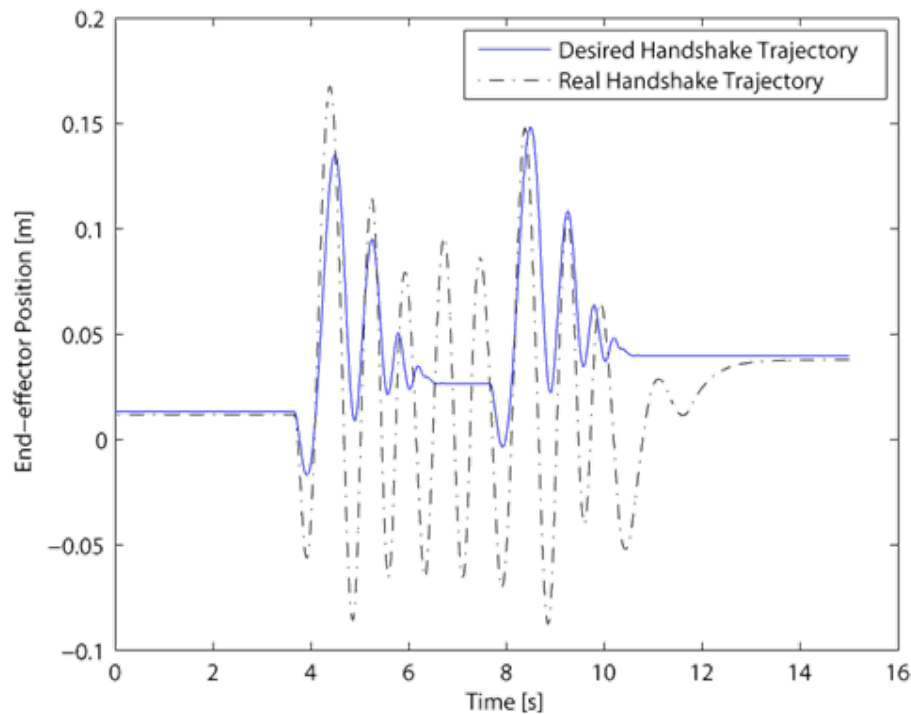


- Interpolation using 5th order polynomial connecting current and desired positions

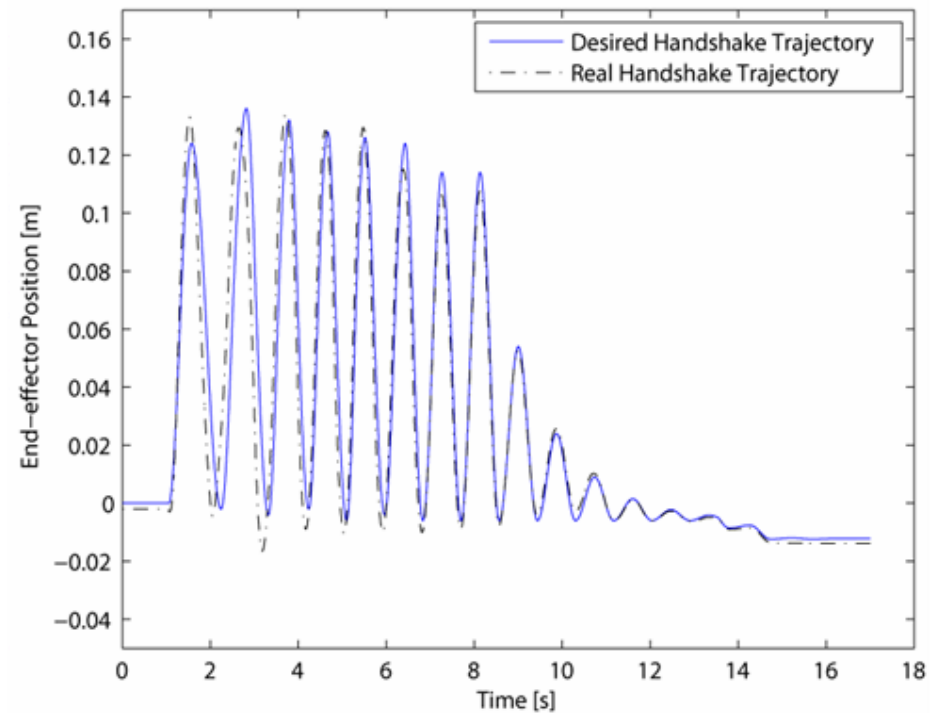
# Experimental Comparison: Replay vs. Interactive Controller

- Human is “active” and tries to lead the robot
- High level controller synchronizes with human trajectory

Compliance controller used for replay



Interactive controller used for recreation



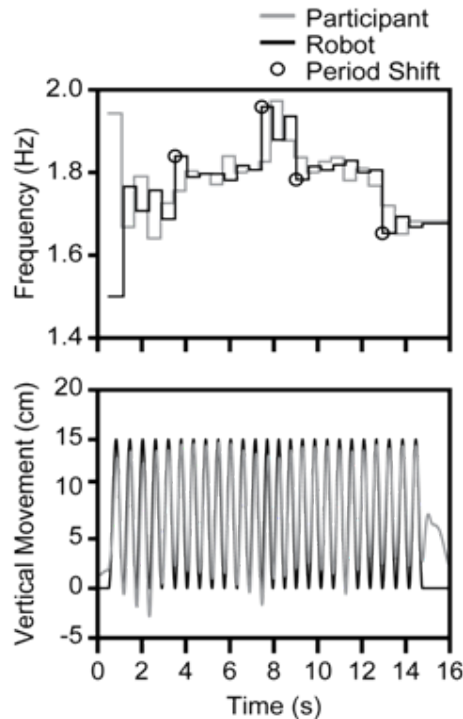
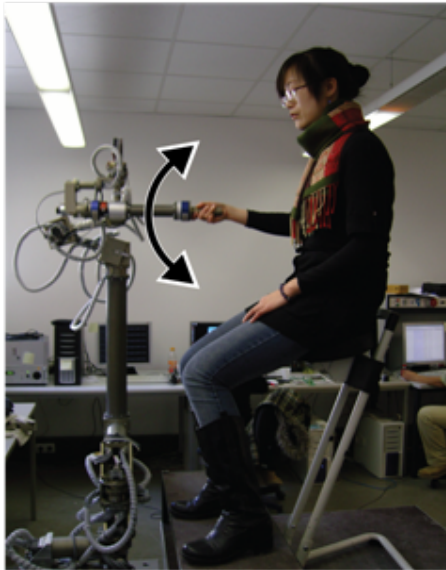


# Mutual Frequency Adaptation in Rhythmic Motor Tasks

*collaboration with Prof. Alan Wing from University of Birmingham*

## Research question:

Is evolution of movement in cooperative haptic manipulation tasks governed by error-based learning?

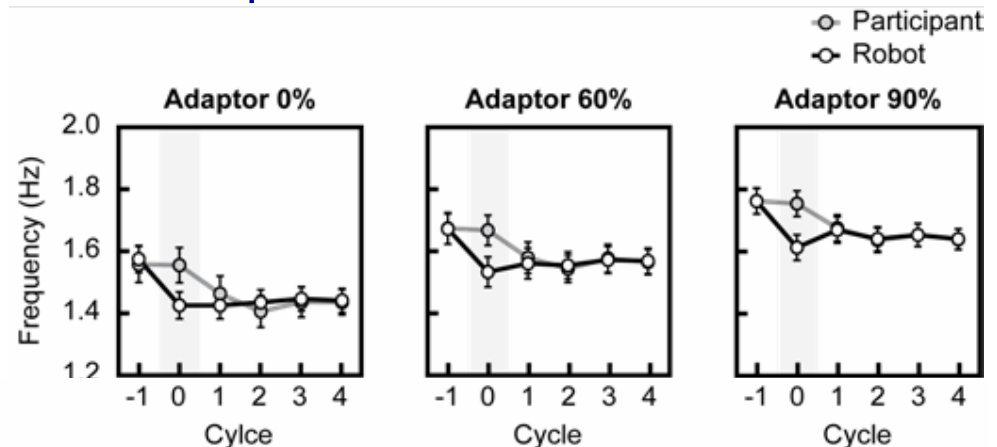


## Implemented adaptation law:

$$F_{n+1}^R = F_n^R + \alpha(F_n^P - F_n^R)$$

## Results:

- large parts of corrections occur within the period shift cycle
  - cooperative behaviour is very sensitive to partner behavior
- adaptation rate is flexibly adjusted to optimise net adaptation gain at cooperative level



# Video



# Evaluation Study: Virtual Cocktail Bar Scenario

[Giannopoulos/Wang/Peer/Buss/Slater, Brain Research Bulletin, submitted]



## Participants:

35 (29 male, 6 female)

## Haptic conditions:

- Basic robot
- Advanced robot
- Human

## Audio conditions:

- Stereo sound 
- Binaural sound 

## Dependent variable:

Presence feeling: plausibility, (place illusion)

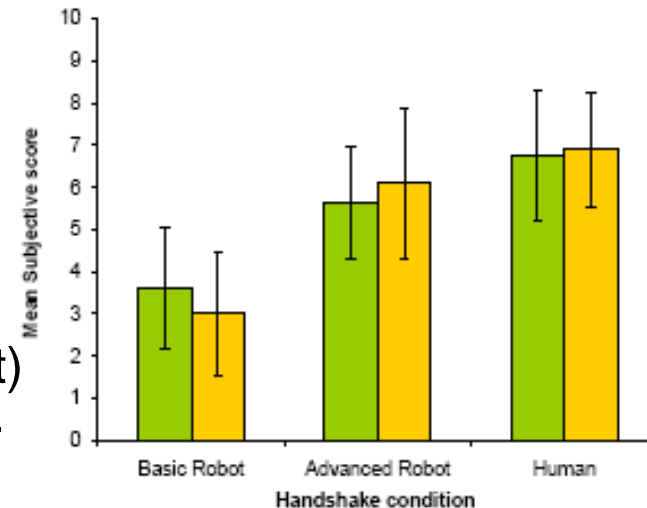
Questionnaire: (robot-like) 1 ... 5 ...10 (human-like)

## Results:

- $\text{mean}(\text{human}) > \text{mean}(\text{advanced robot}) > \text{mean}(\text{basic robot})$
- substantial confusion between advanced robot and human-operated handshakes (similar effect sizes)



## 2nd study: Visual condition



# Video



RECORD A DANCING COUPLE



# Summary and Conclusion

- Implementation of a haptic robot interaction partner is challenging as humans expect a partner showing **human-like interaction capabilities** in terms of **interactivity, proactivity, adaptivity**
- Thus, implementation of haptic interaction partners requires algorithms for
  - Human intention estimation
  - Interactive path planning
  - Adaptation of robot behavior
- **Synthesis of robot behavior** can follow different approaches
  - Engineering-driven or experimentally-driven approach
- **Human-human experiments** can inform implementation of haptic interaction partners



# Acknowledgement



Daniela  
Feth



Raphaella  
Groten



Zheng  
Wang



Jens  
Hölldampf

