

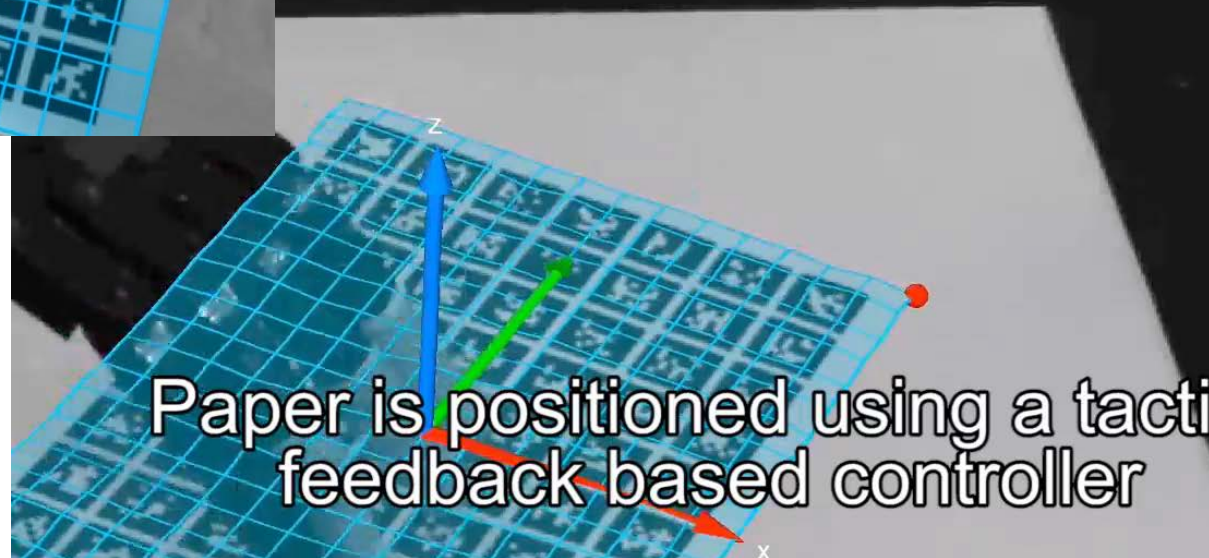
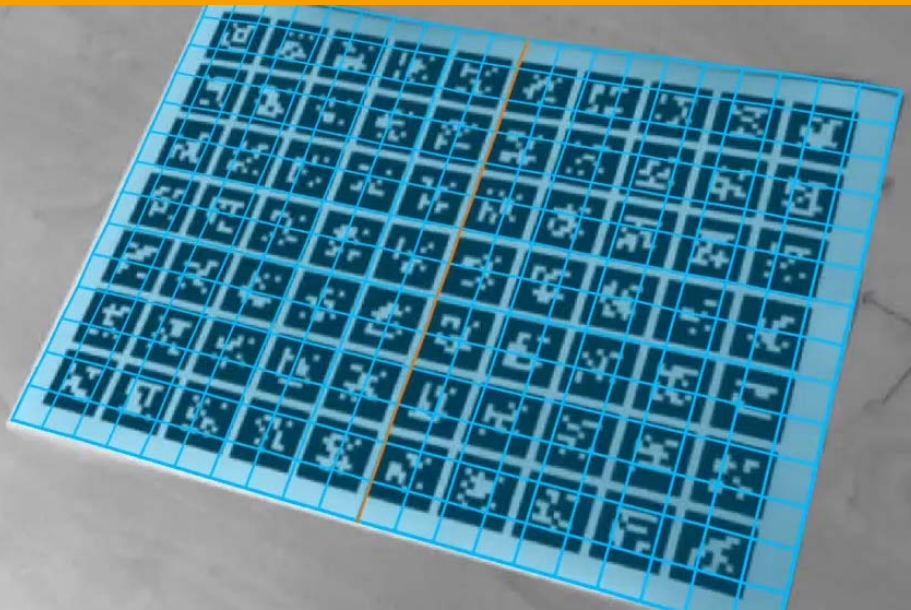
Tactile Perception for Autonomous Grasping and in Cars

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Background: Understand + Replicate Manual Intelligence

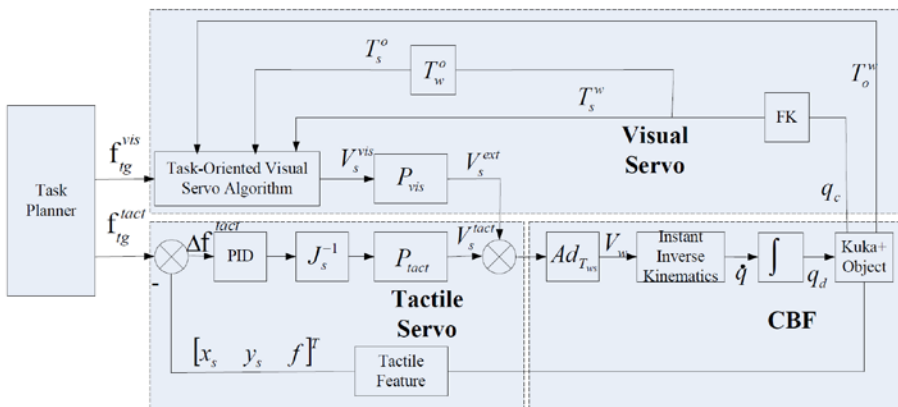


Paper is positioned using a tactile feedback based controller

Classical Engineering vs. Deep-Learning

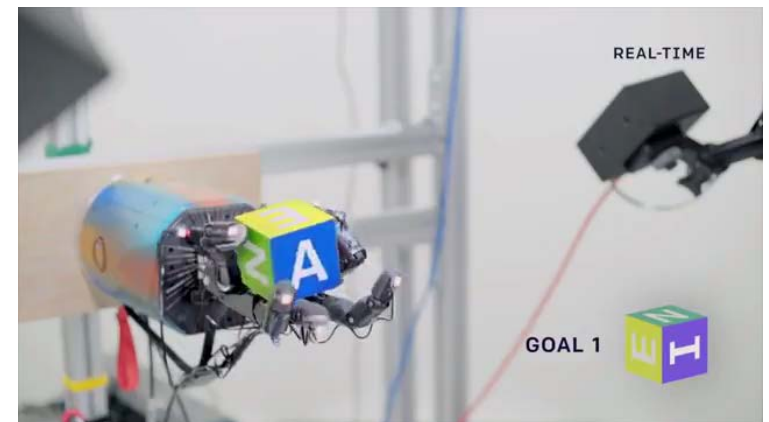
Classical Control Engineering

- Great results when world is correctly modelled
- No generalization beyond model
- Human-time intensive



End-to-End Learning

- Indispensable when modelling is hard or infeasible
- Generalization within training domain
- Machine-time intensive



Classical Engineering vs. Deep-Learning

Classical Control Engineering

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Combine both approaches!

- Exploit prior world knowledge / models.
- Focus learning on aspects that are hard to model

Importance of tactile and visual feedback



Tactile Sensors for Manipulation

- How can we utilize tactile sensors
 - for handover?
 - to detect dropping of objects?
 - to detect incipient slippage?
 - to detect unintended contact?
 - explore object surfaces?
- What are required specs w.r.t.
 - spatial resolution?
 - temporal resolution?
 - sensitivity?
 - force range?



Incipient Slip Detection + Grasp Force Adaptation

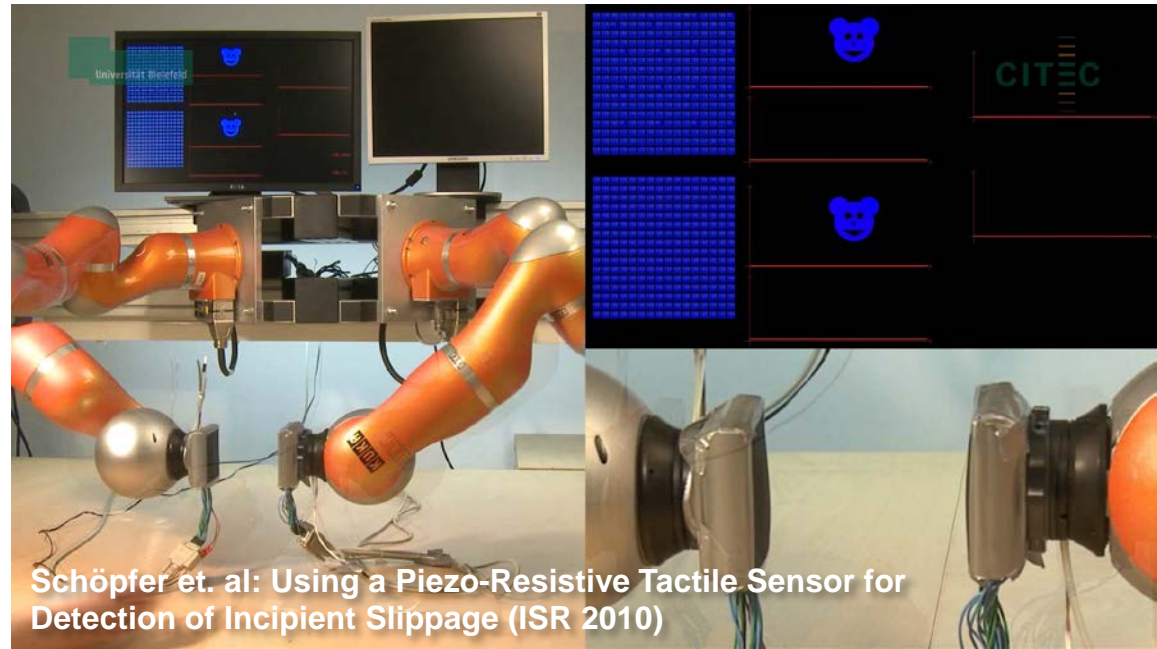
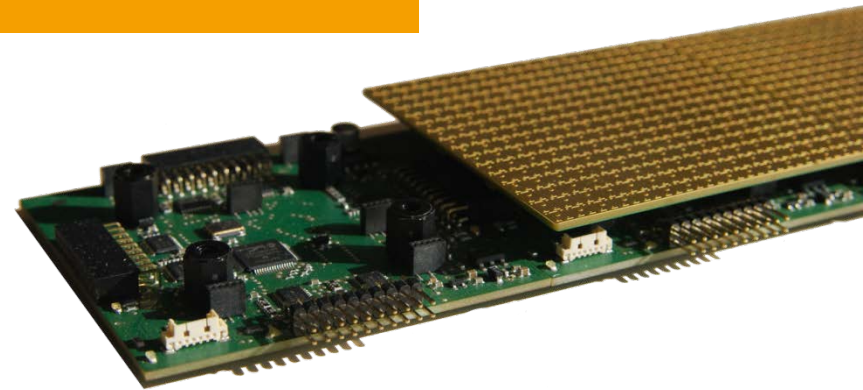
- Incipient slippage generates micro vibrations in ~ 400 Hz range
- ... amplified by human fingerprints
- calls for fast tactile sensors

Incipient Slip Detection + Grasp Force Adaptation

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- calls for fast tactile sensors

- 16x16 tactile array
- ≈ 2 kHz frame rate

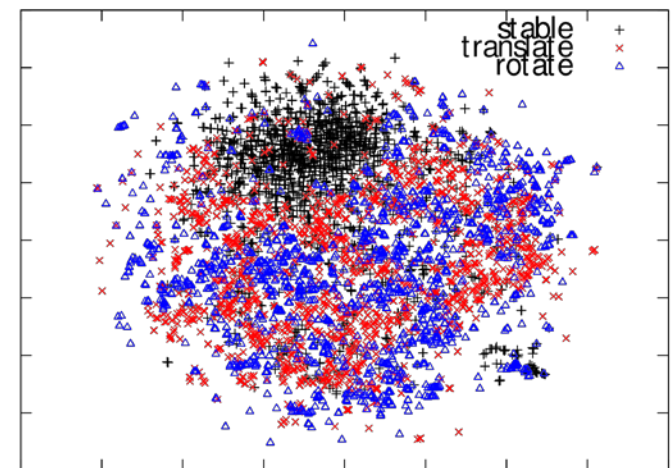
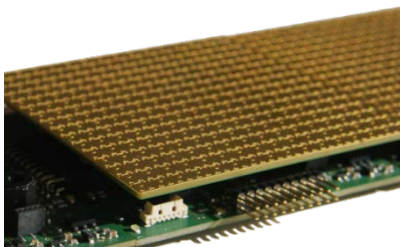
- train a neural network to classify slippage
- increase grasping force on slippage



Schöpfer et. al: Using a Piezo-Resistive Tactile Sensor for Detection of Incipient Slippage (ISR 2010)

Incipient Slip Detection: Going more complex

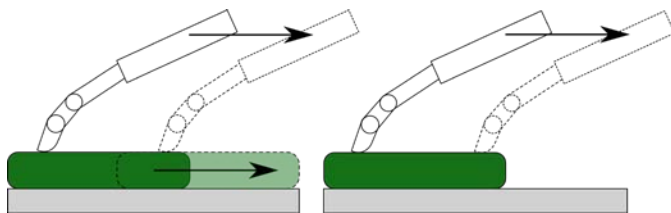
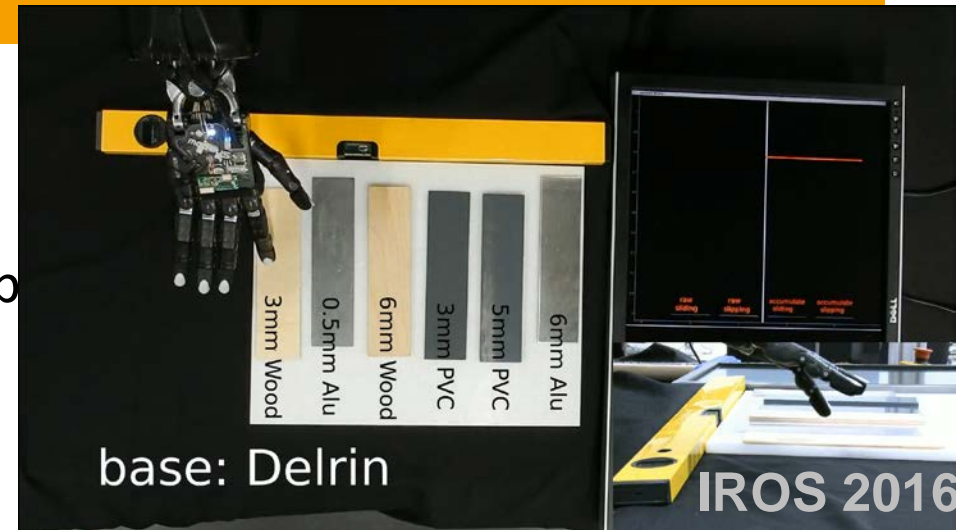
- Basic slip detection is simple
- What about different types of slippage?
 - stable – translational– rotational
- Sensor array vs. small finger tip?



t-SNE embeddings of input

Incipient Slip Detection: Going more complex

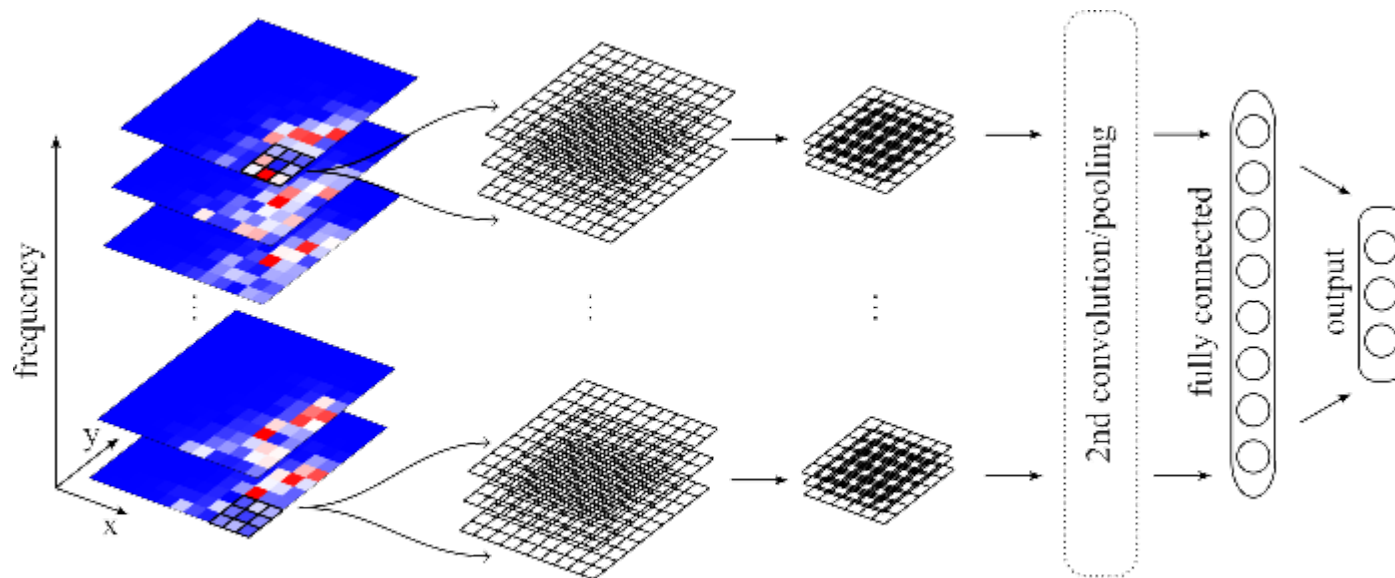
- Basic slip detection is simple
- What about different types of slipping
 - stable – linear – rotatory
- Sensor array vs. small finger tip?
- Object sliding vs. slipping



t-SNE embeddings of input

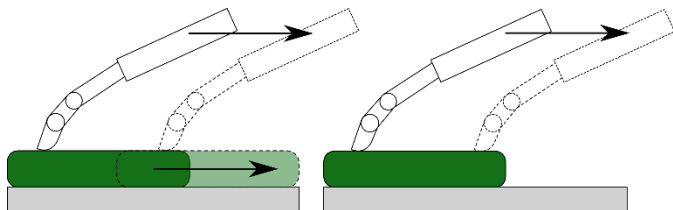
Deep Convolutional Neural Network for Slip Detection

- Fourier analysis for individual taxels
 - spatially arranged in taxel grid
- learn convolutional filter (3x3) per frequency layer
- 2x2 pooling
- fully-connected classification layer



Results

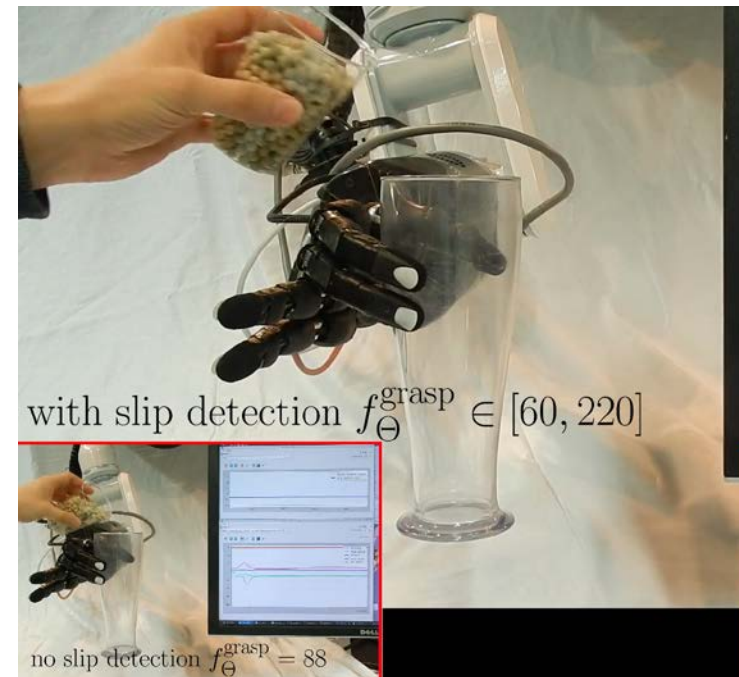
- Different *types* of slip
 - static
 - translational
 - rotational
- Slip Detection on Fingertips
- „Sliding with“ vs. „Sliding on“ an object



97.8%

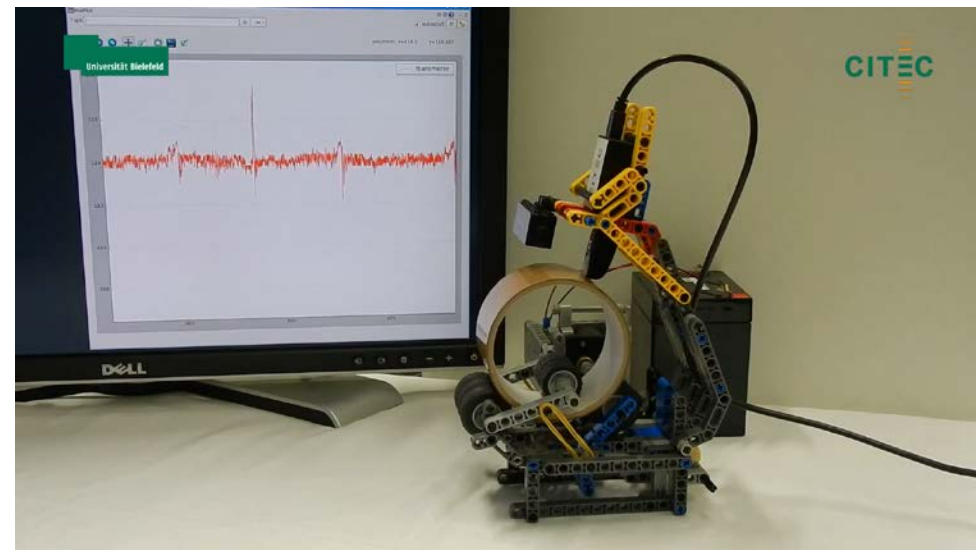
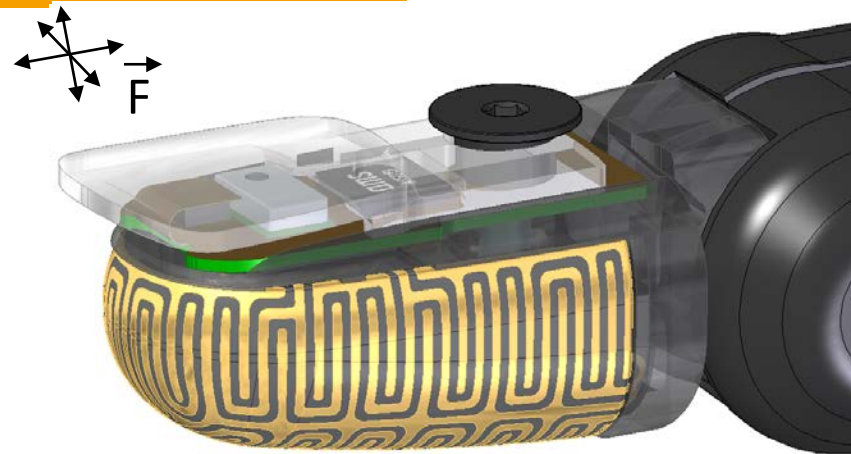
96.4%

81.9%
vs.
73.3%
(MLP)

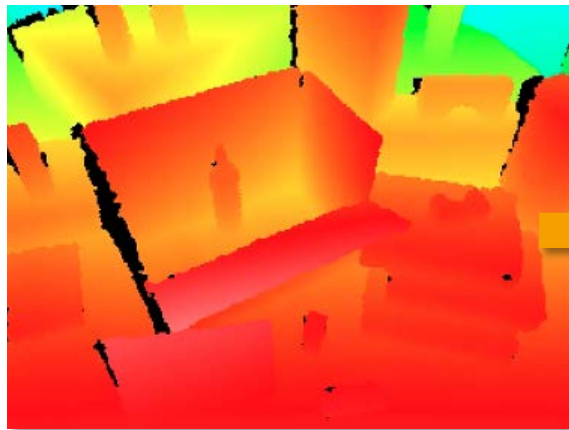


Sensorized Fingernail

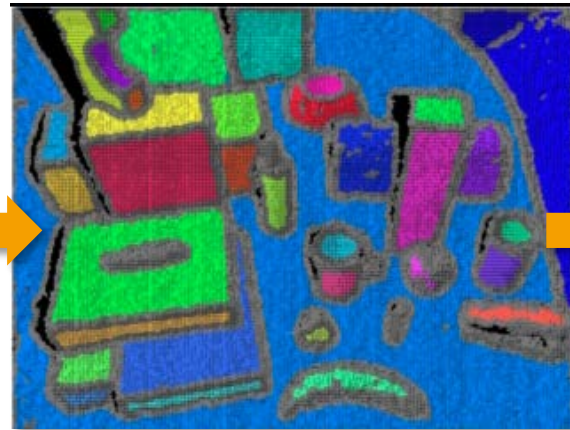
- nail important for:
 - grasping tiny objects
 - texture recognition
- 3-axes force sensor
 - lateral forces
 - hall sensors
- normal force
 - barometer sensor
 - sensitivity: $\approx 6\text{mN}$
- example application
 - detect end of tape



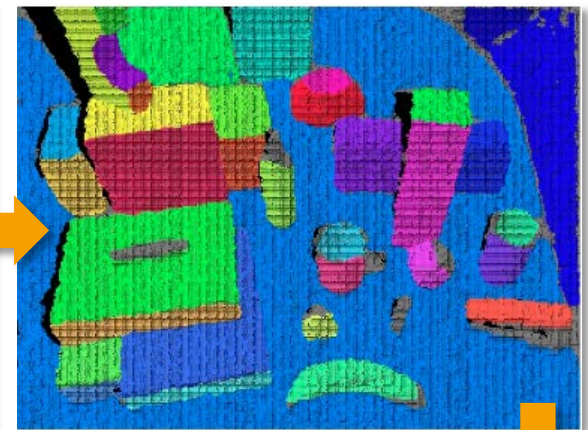
Learning Grasps from Depth Images



RGB-D input



Surface segmentation



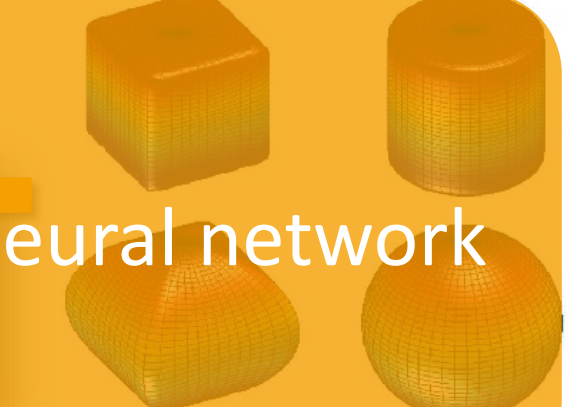
Object segmentation



Grasp execution



Grasp selection



Superquadric fitting

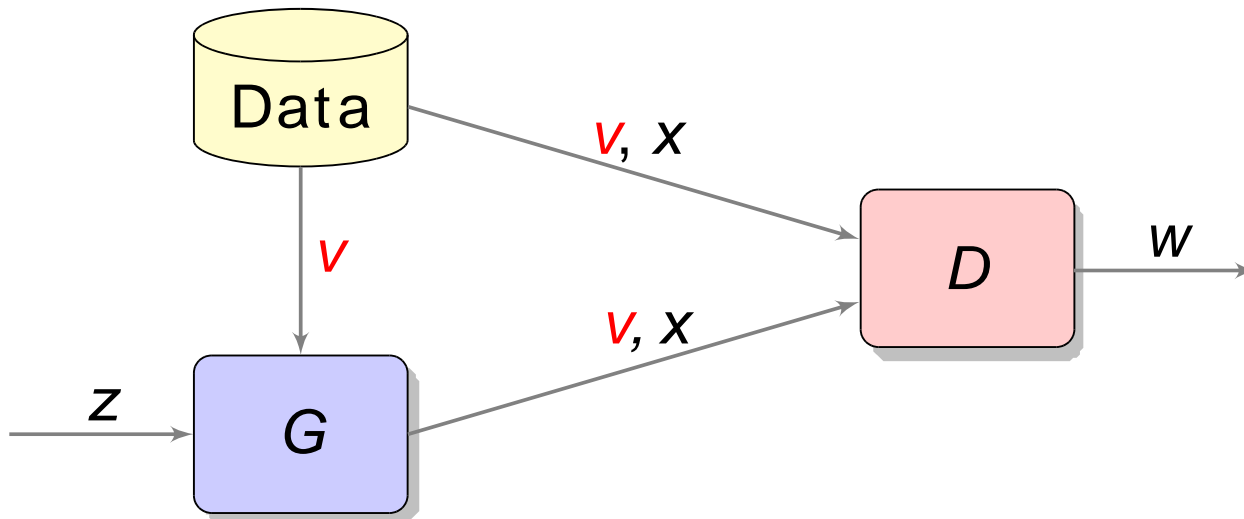
Replace with neural network



Learning Grasps from Depth Images

- Conditional Wasserstein GAN
- Generative model

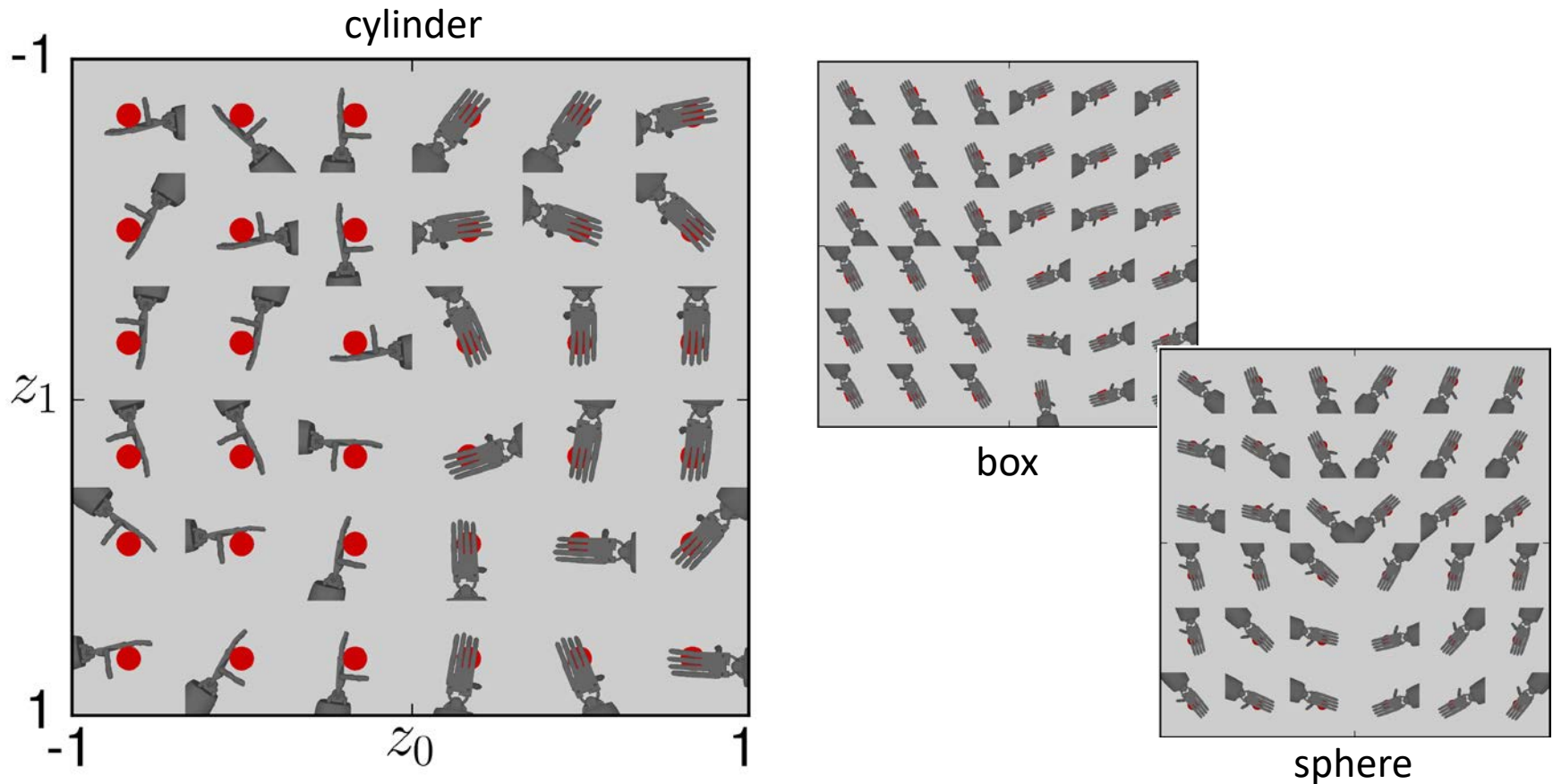
G: grasp generator
D: discriminator
v: depth image
x: grasp
z: random variable



$$\min_D \max_G \mathbb{E}_{\mathbf{v} \sim P_v} \left(\mathbb{E}_{\tilde{\mathbf{x}} \in p_{\tilde{\mathbf{x}}}(\mathbf{v})} [D(\tilde{\mathbf{x}}, \mathbf{v})] - \mathbb{E}_{\mathbf{x} \in p_{\mathbf{x}}(\mathbf{v})} [D(\mathbf{x}, \mathbf{v})] \right)$$

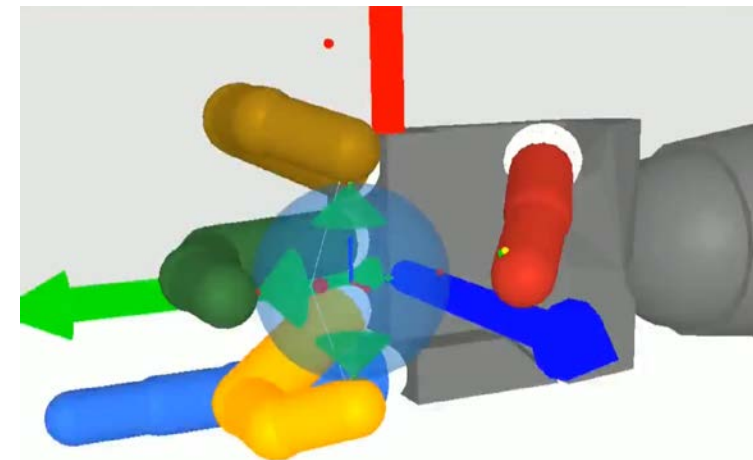
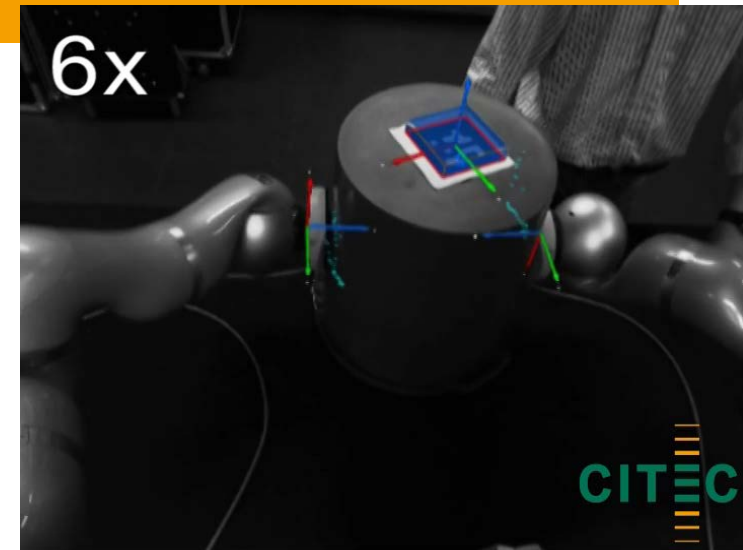
Learning Grasps from Depth Images: Result

- Latent space exhibits nice, self-emerging structure



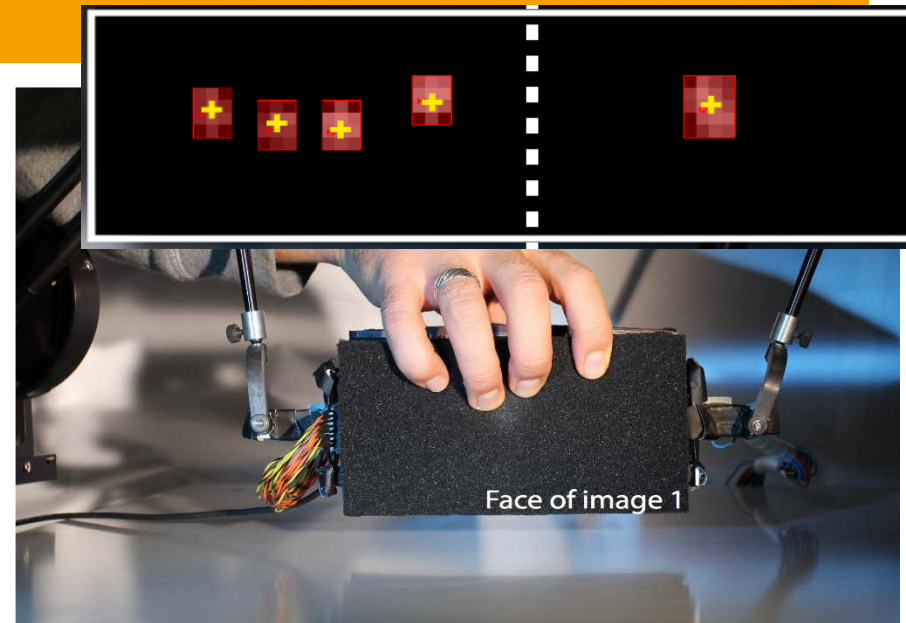
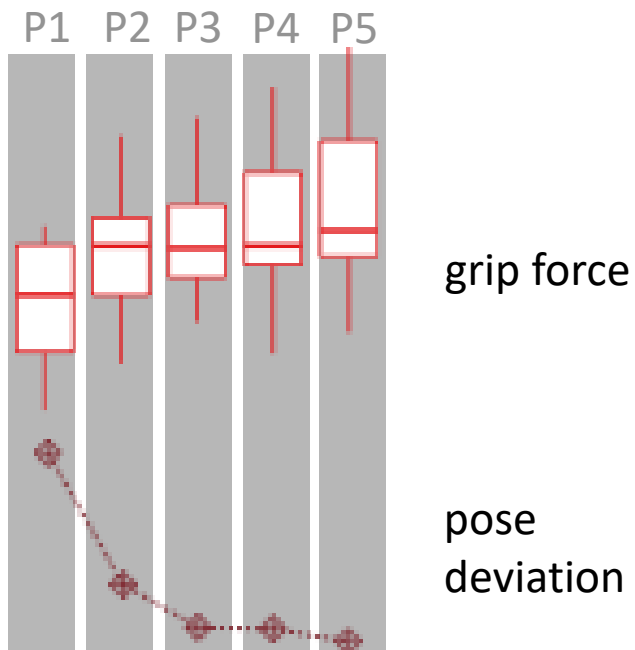
Tactile Servoing

- use tactile features directly for EE control
- **contact force** error mapped onto
⇒ normal motion
- **contact position** error mapped onto
⇒ tangential motion
⇒ rolling/tilting motion
- **axis orientation** error mapped onto
⇒ rotation around normal axis
- external motion component
generates exploratory behaviour



Sensorized Objects for Human Manipulation Research

- „Tactile Object (TACO)“
 - book-like object
 - 2x2 sensor arrays á 16x16

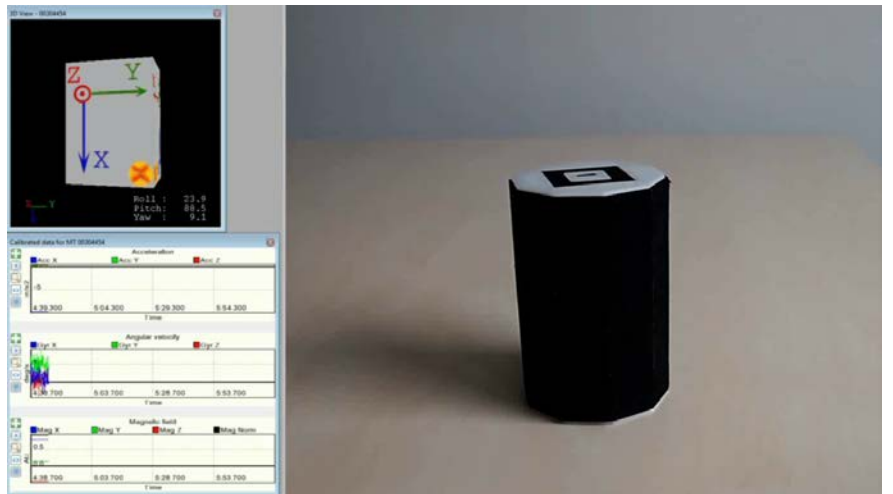
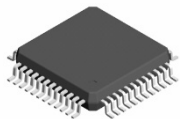
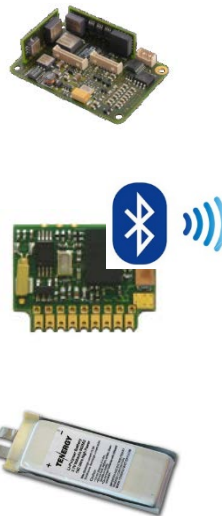
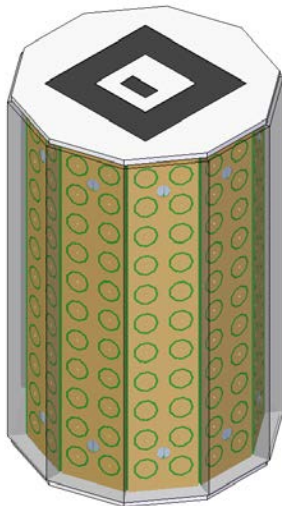
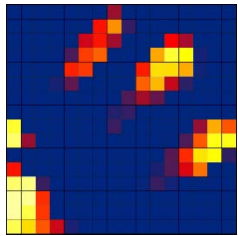
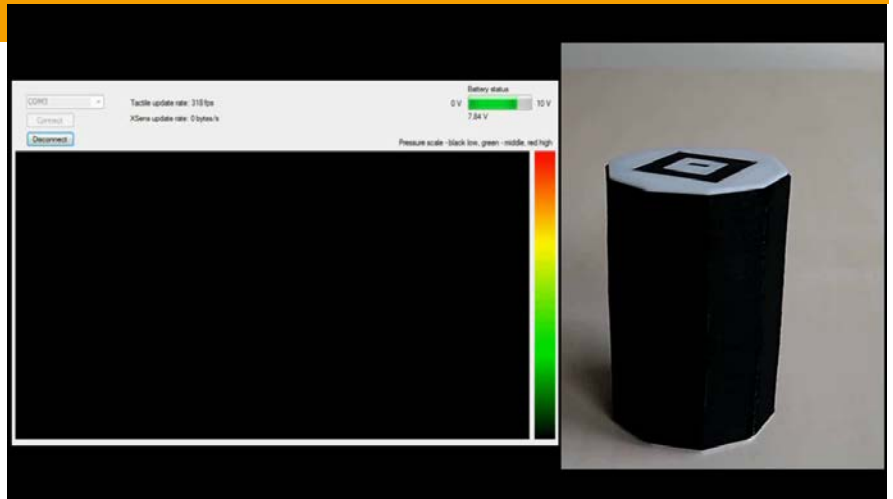


When perturbed, humans quickly adapt and stiffen the grasp in a synergistic fashion.

Sensorized Objects for Human Manipulation Research

- **Tactile Can**

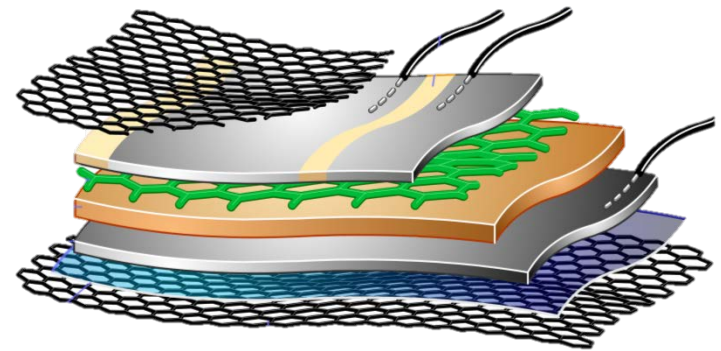
- 10 strips of sensor arrays á 2x10
- motion sensor
- various tools



Tactile Data-Glove for Human Manipulation Research



Büscher et al. Tactile dataglove with fabric-based sensors (Humanoids 2012)



Universität Bielefeld CITEC

Tactile dataglove with fabric-based sensors

54 tactile sensitive areas
Sensed pressure range 1 to 600 kPa

Gereon Büscher, Risto Kõiva, Carsten Schürmann,
Robert Haschke and Helge J. Ritter

Tactile Sensors for Hand Protheses

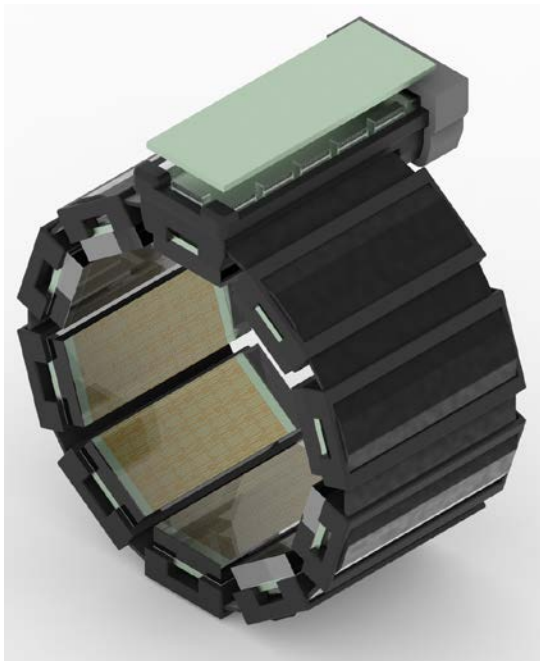
- Instrument phalanges of iLimb prosthetic hand
- sealed by silicon wrapping
- to be slipped on iLimb bones

- deep neural network applications
 - slip detection: LSTM, 92% acc.
 - stiffness classification: LSTM correlating hand-closing profile to tactile forces



Tactile-Based Prosthesis Control

- Replace EMG sensors with taxel array arranged as bracelet
- Measure muscle bulgings
 - 10x 4x8 taxel arrays



Tactile pen for therapy

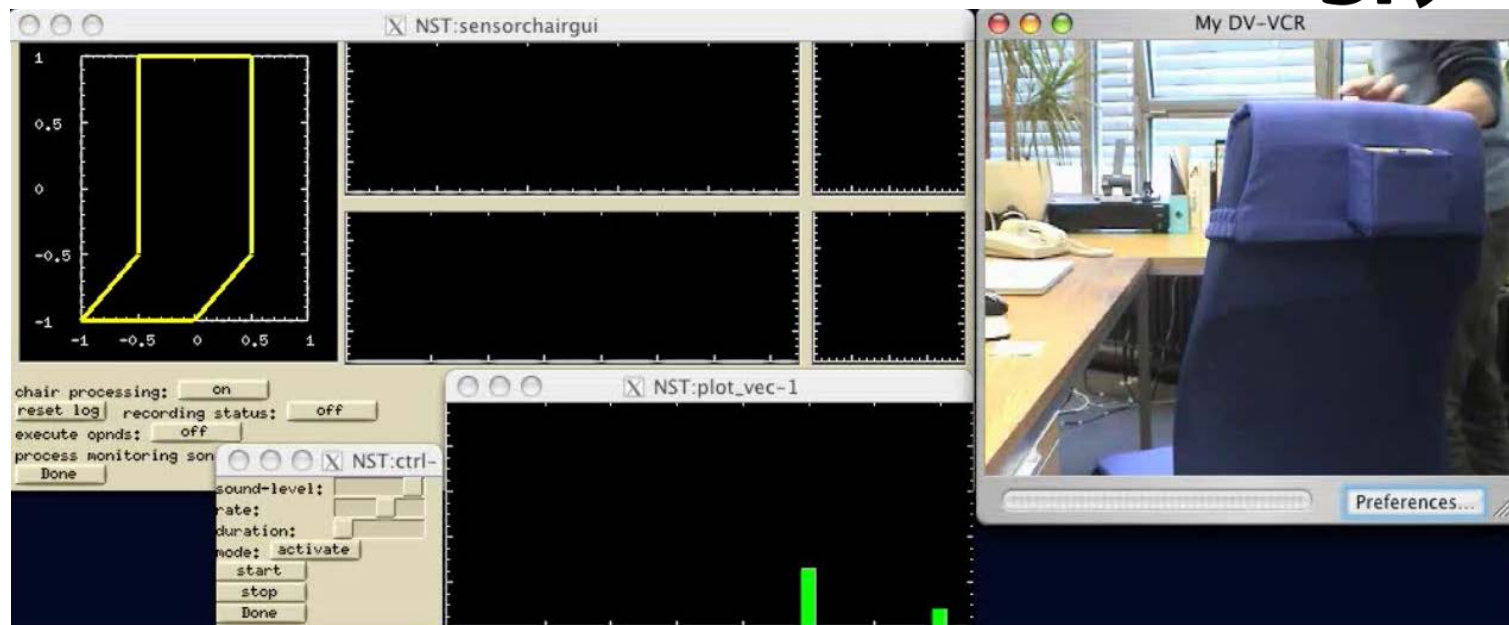
- Bad handwriting caused by bad hand posture and too high forces
- Measure applied forces
- and provide feedback



Tactile pen prototype using fabric-based sensors

Tactile Applications in Intelligent Cars?

- Monitor sitting pose of driver
 - adjust seat w/o knobs
 - monitor driver's attention
- „SonicChair“: sonified bio-feedback for sitting pose



Summary

- Combine classical engineering approaches with learning
- Exploit prior knowledge wherever possible to facilitate learning

- Tactile sensing is key for manipulation
- ... but has numerous applications beyond

Acknowledgements



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Büscher

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