

# Haptic Sensing for Assistive Robots

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EMORY  
UNIVERSITY

# Haptic Sensing for Assistive Robots

- Quick overview of assistive robotics
- Data-driven models of forces
- Whole-arm tactile sensing
- Thermal tactile sensing

**Healthcare Robotics Lab**

<http://healthcare-robotics.com>

# Health-Related Physical Assistance

- People with motor impairments, injuries and illnesses would often benefit from physical assistance.
- Today in the US alone
  - >15,000 with ALS <sup>[1]</sup>
  - ~290,000 with a spinal cord injury <sup>[2]</sup>
  - ~14,000,000 aged 65 years and older have a severe disability <sup>[3]</sup>



[1] Paul Mehta, M. D. "Prevalence of Amyotrophic Lateral Sclerosis—United States, 2012–2013." MMWR. Surveillance Summaries 65 (2016).

[2] Singh, Anoushka, et al. "Global prevalence and incidence of traumatic spinal cord injury." Clin Epidemiol 6 (2014): 309-331.

[3] Brault, Matthew W. "Americans with disabilities: 2010." Current population reports 7 (2012): 0-131.

[image] <http://www.nurseuncut.com.au/how-stressed-are-you/>

# Types of Tasks

- **Activities of Daily Living (ADLs)**
  - Feeding, toileting, transferring, dressing, and hygiene
  - Predictive of ability to live independently
  - Manipulation near the person's body
- **Instrumental Activities of Daily Living (IADLs)**
  - Housework, food preparation, taking medications, ...
  - Manipulation of objects in the environment



*In 2010 in the US alone, ~12,000,000 “aged 6 years and older ... needed assistance with one or more ADLs or IADLs” [1]*

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# Robotic Opportunities



- Provide **independence**
- Robots preferred for some tasks [1]
- 24/7 personalized assistance

[1] Cory-Ann Smarr, Tracy L. Mitzner, Jenay M. Beer, Akanksha Prakash, Tiffany L. Chen, Charles C. Kemp, and Wendy A. Rogers. *Domestic robots for older adults: Attitudes, preferences, and potential*. *International Journal of Social Robotics*, 6(2):229–247, 2014.

[image] from Willow Garage

# Assistive Robots

- Robotic Prostheses
- Robotic Orthoses / Exoskeletons
- Wheelchair Mounted Robot Arms
- Desktop Robots
- Mobile Manipulators



**DEKA Arm by DEKA**



**JACO by Kinova**



**HAL by Cyberdine**



**MySpoon by SECOM**

# Potential Benefits of Mobile Manipulators

- Operate independently from user
- No don/doff
- Assist diverse end users
- Mass market product



[image] from Willow Garage



# Our Data Suggest that Many People will be Open to Assistance from Mobile Manipulators

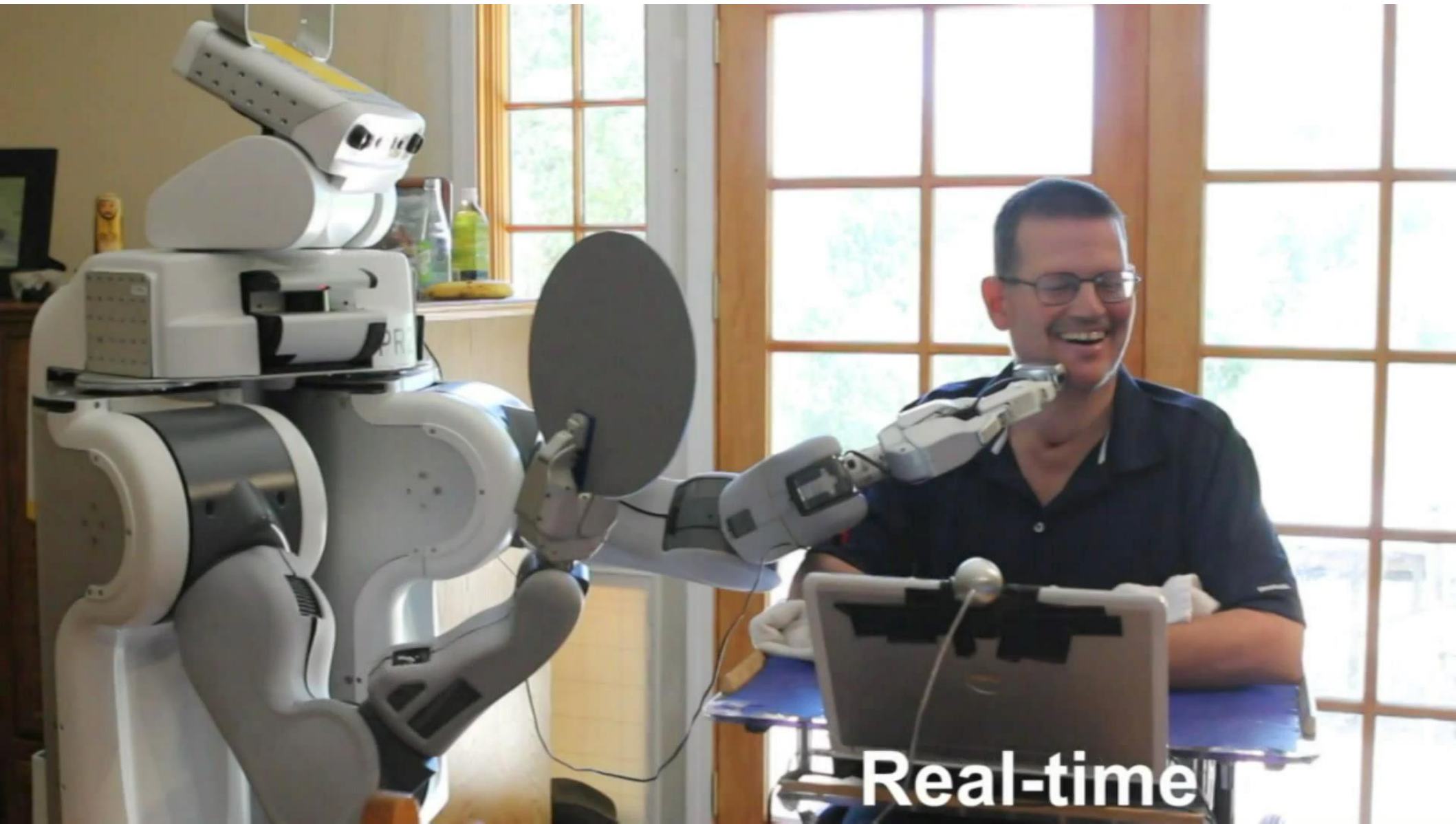
- Over 200 participants in studies about assistive mobile manipulation
- Most studies with representative end-users
  - Older adults
  - Nurses
  - Able-bodied participants (mock generic patients)
  - People with severe motor impairments



[Image] Chih-Hung King, Tiffany L. Chen, Zhengqin Fan, Jonathan D. Glass, and Charles C. Kemp, *Dusty: An Assistive Mobile Manipulator that Retrieves Dropped Objects for People with Motor Impairments*, *Disability and Rehabilitation: Assistive Technology*, 2011.



Hai Nguyen, Matei Ciocarlie, Kaijen Hsiao, and Charles C. Kemp, *ROS Commander (ROSCo): Behavior Creation for Home Robots*, IEEE International Conference on Robotics and Automation, 2013.



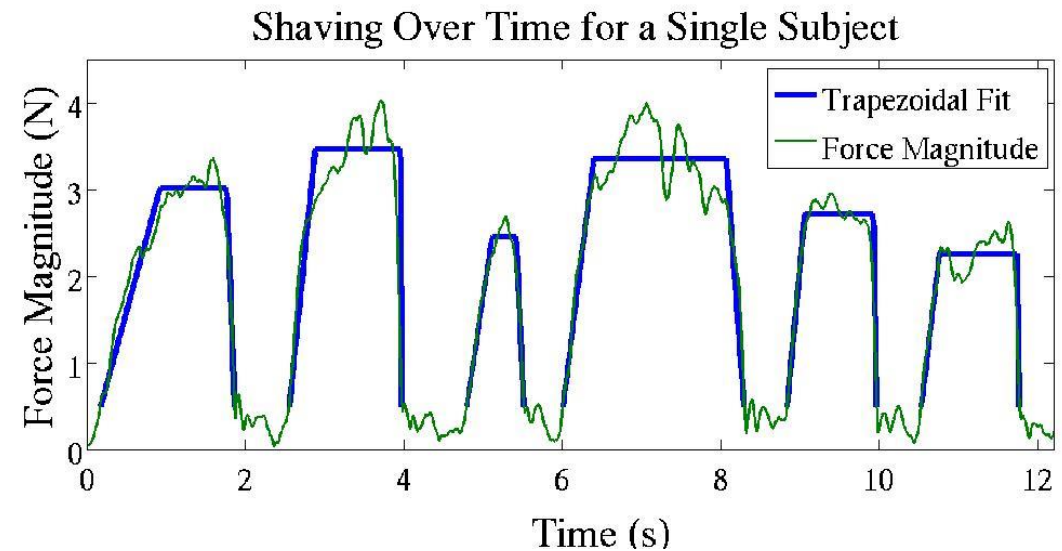
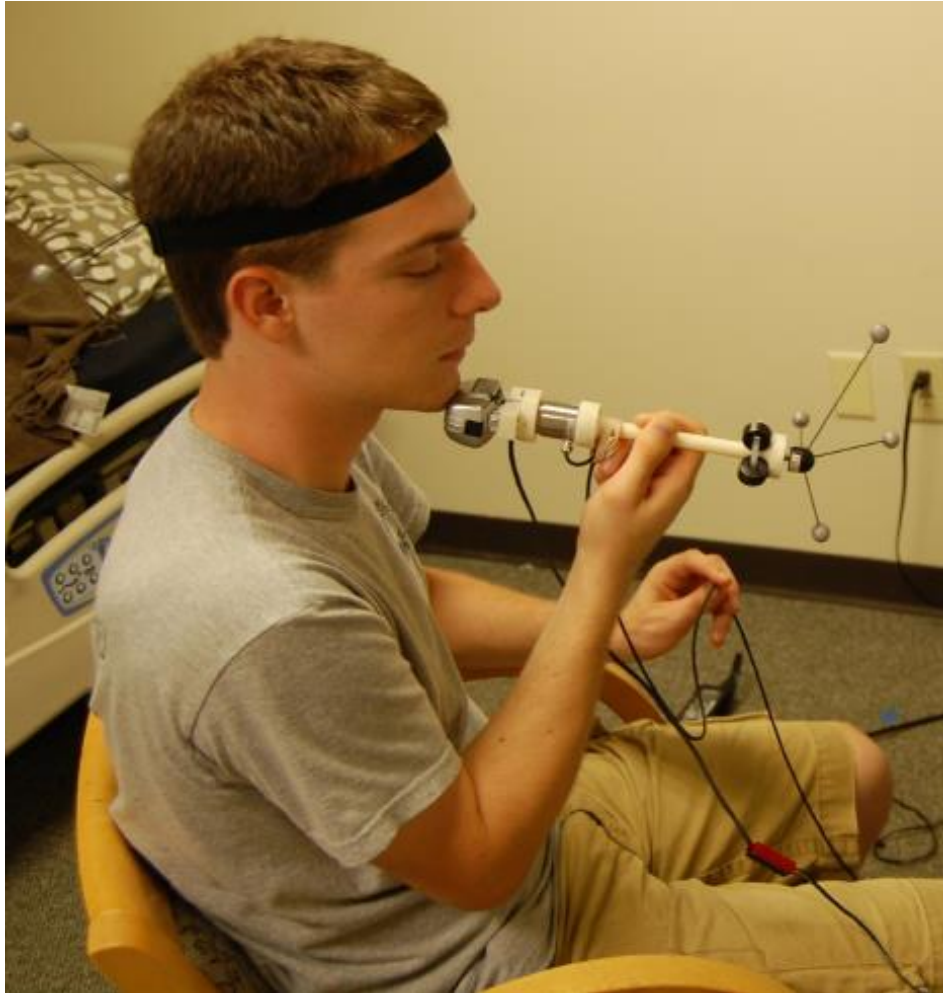
**Real-time**

Kelsey Hawkins, Phillip M. Grice, Tiffany L. Chen, Chih-Hung King, and Charles C. Kemp, *Assistive Mobile Manipulation for Self-Care Tasks Around the Head*, 2014 IEEE Symposium on Computational Intelligence in Robotic Rehabilitation and Assistive Technologies, 2014.

**How can haptic sensing help?**

# **Data-Driven Models of Forces for Robot-Assisted Tasks**

# Data-driven Models for Robot-Assisted Shaving

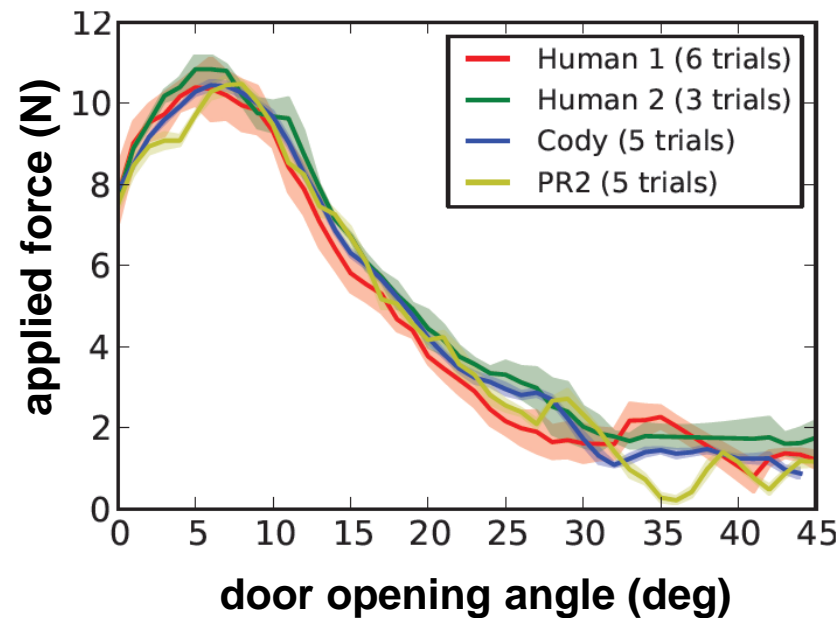


## Statistics for the Target Force of Shaving Strokes

Maximum	8.2 N
First Quartile	2.7 N
Median	1.9 N

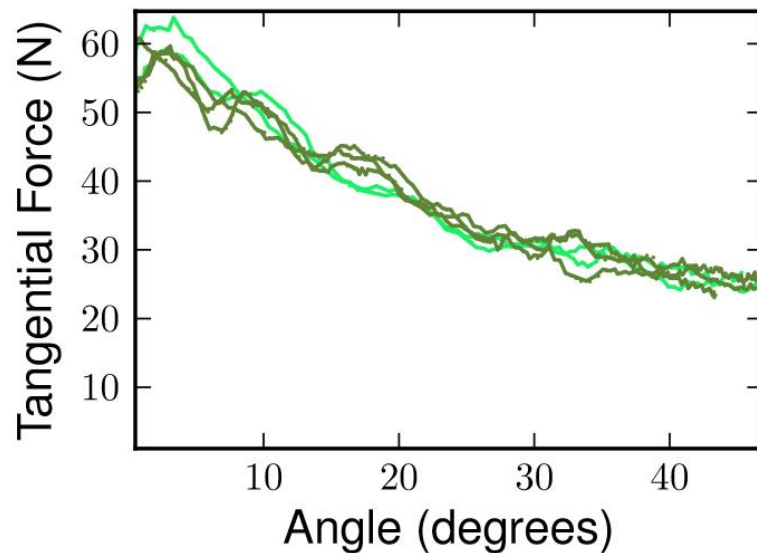
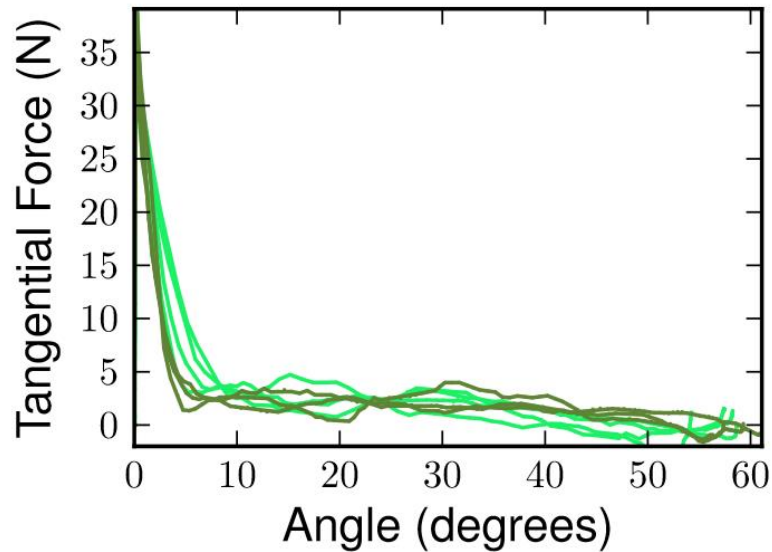
Kelsey Hawkins, Chih-Hung King, Tiffany L. Chen, and Charles C. Kemp, *Informing Assistive Robots with Models of Contact Forces from Able-Bodied Face Wiping and Shaving*, IEEE International Symposium on Robot and Human Interactive Communication (RO-MAN), 2012.

# Data-driven Models for Robot-Assisted Door Opening



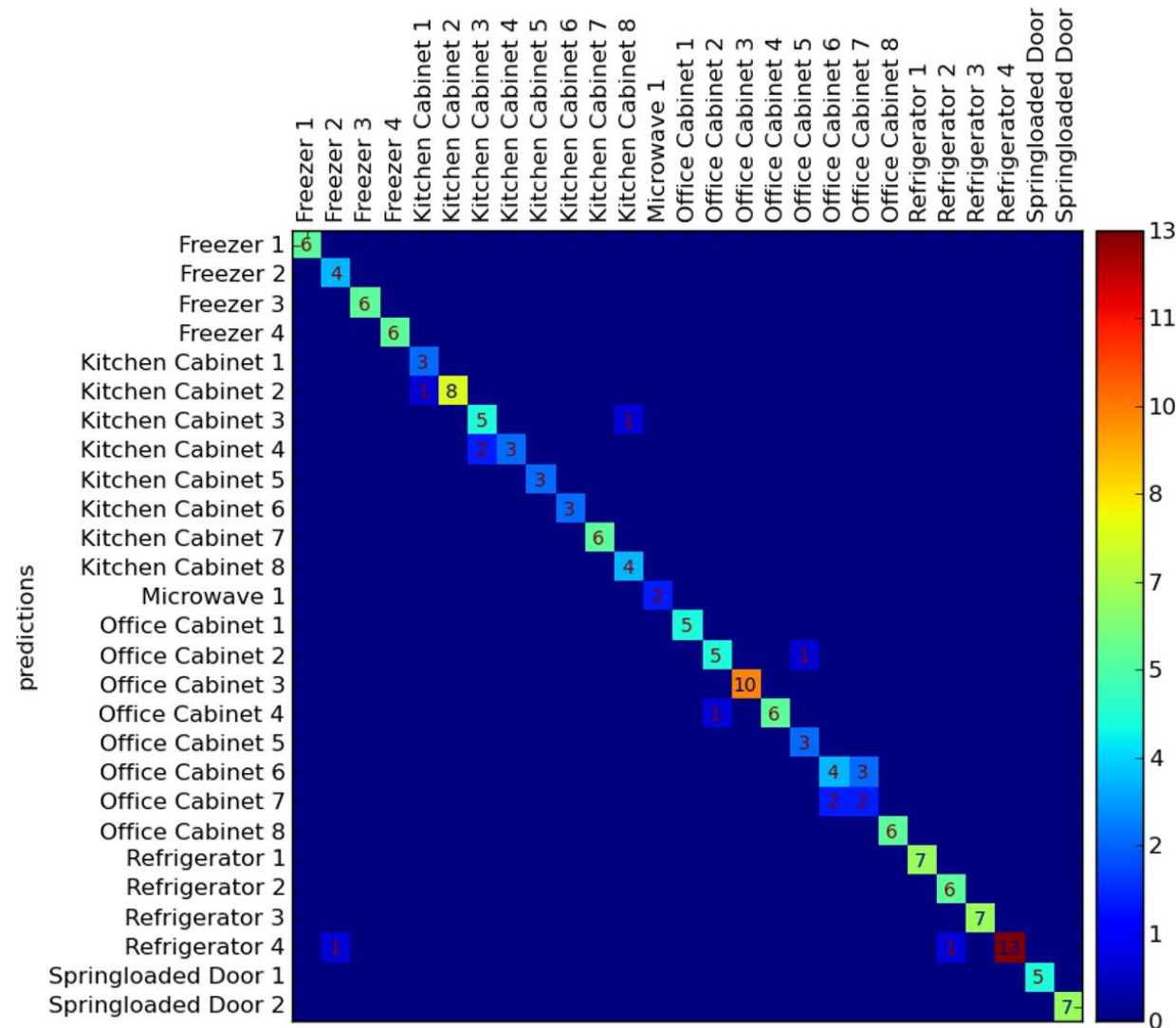
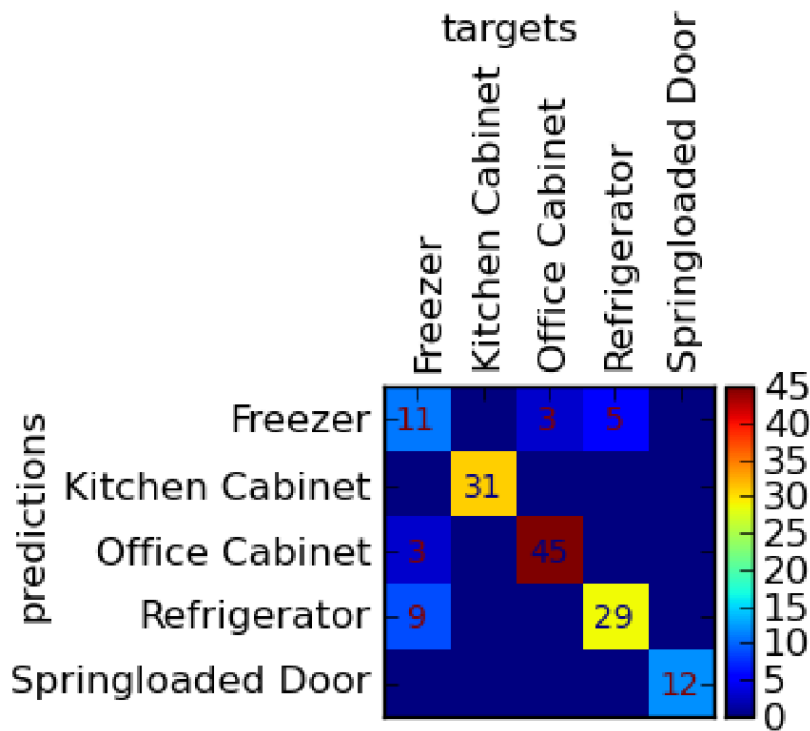
Advait Jain and Charles C. Kemp, *Improving Robot Manipulation with Data-Driven Object-Centric Models of Everyday Forces*, Autonomous Robots, 2013.

# Data-driven Models for Robot-Assisted Door Opening





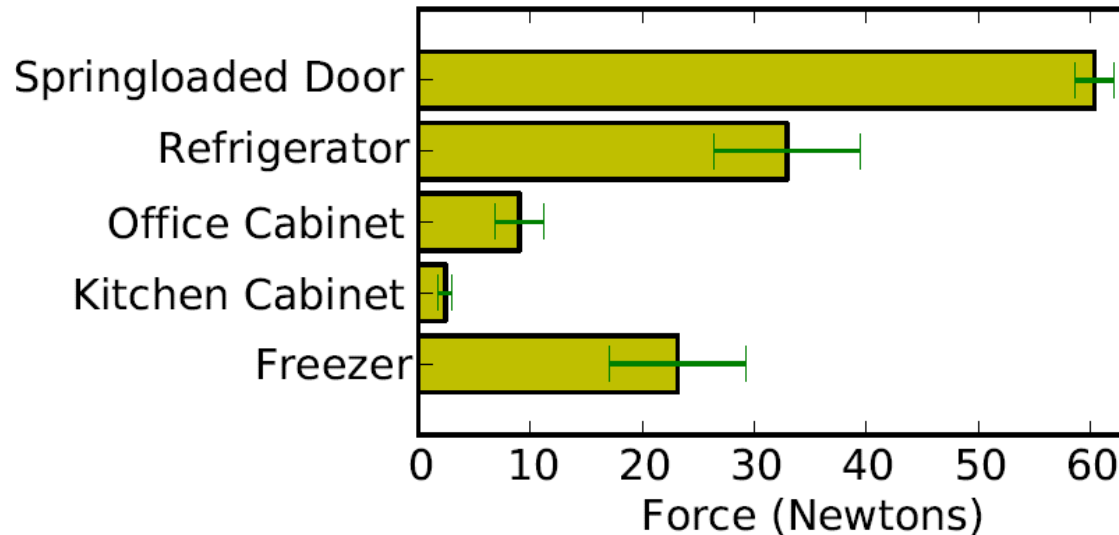
# Recognition of Categories and Instances



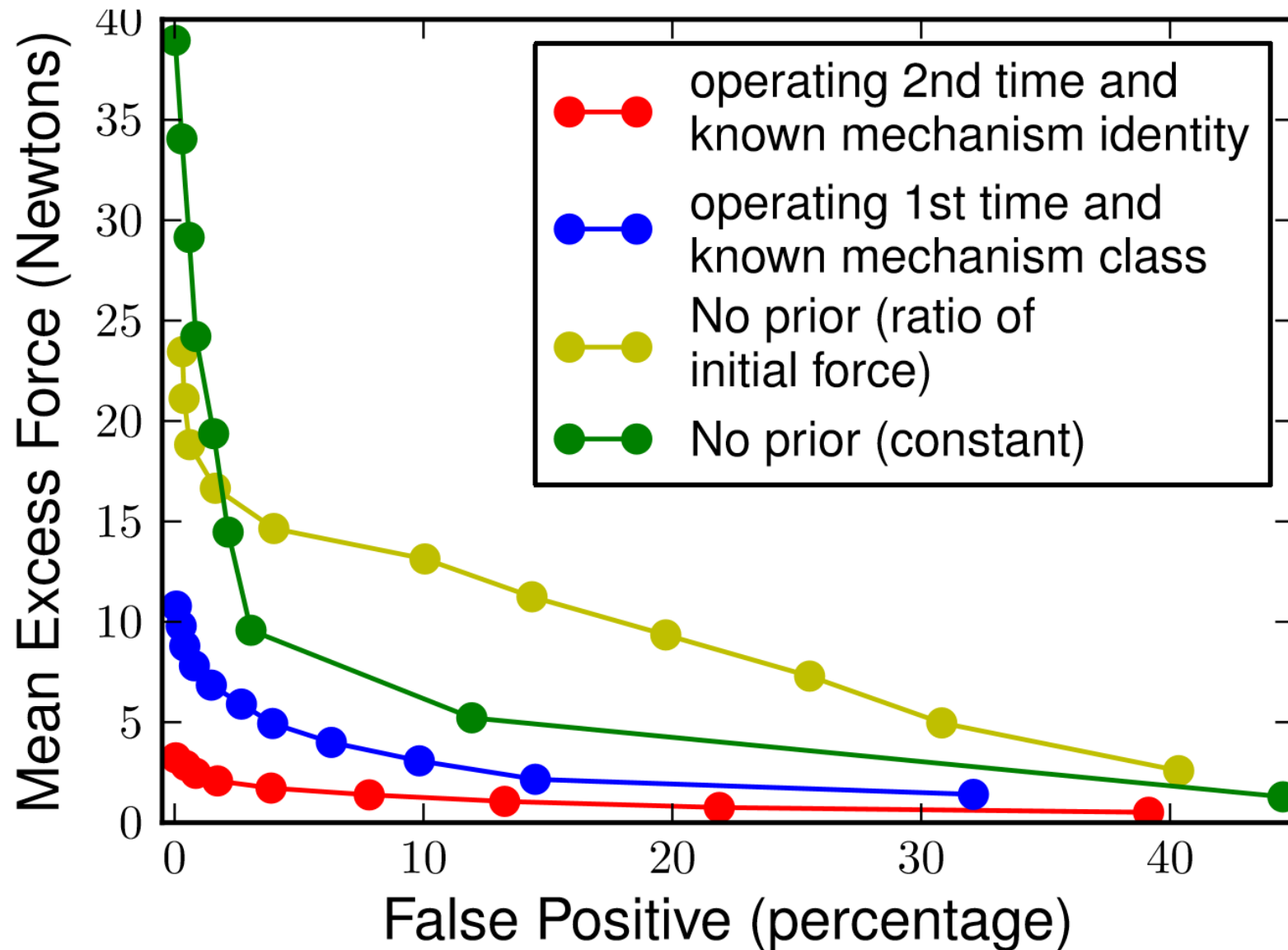
# Aggressive, Timid, or Smart?



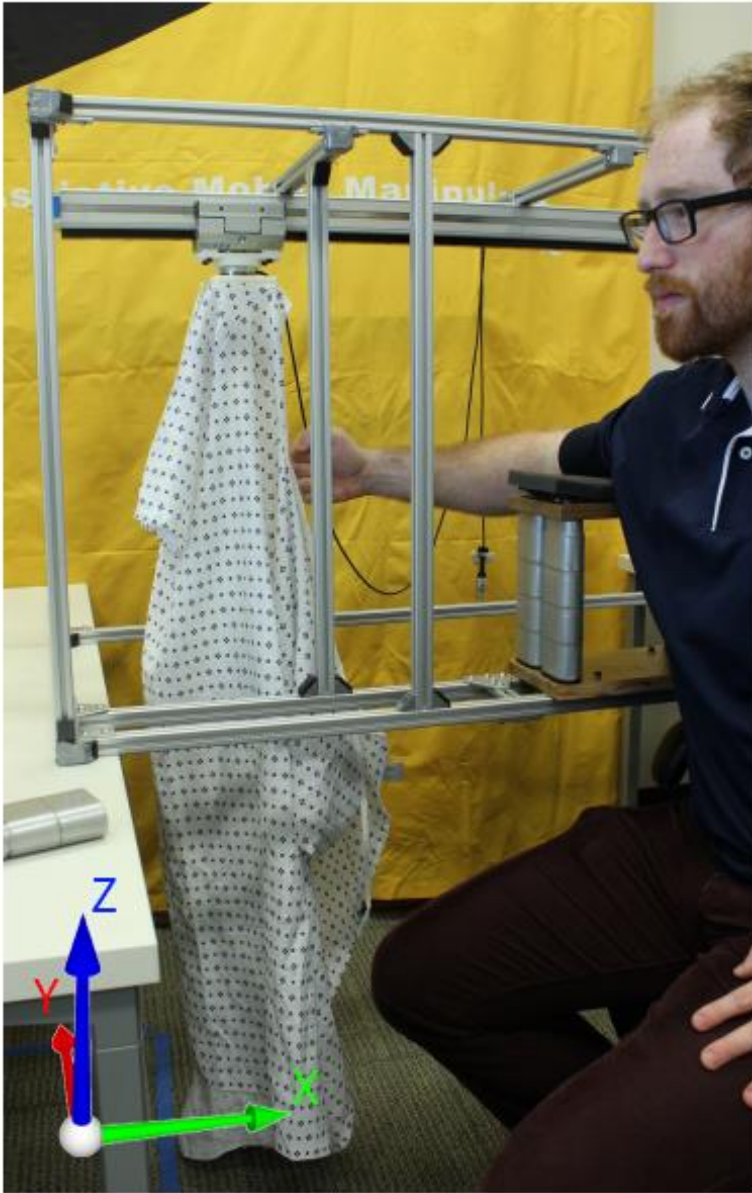
Initial Force to Open



# Data-driven Models for Robot-Assisted Door Opening



# Data-Driven Models for Robot-Assisted Dressing



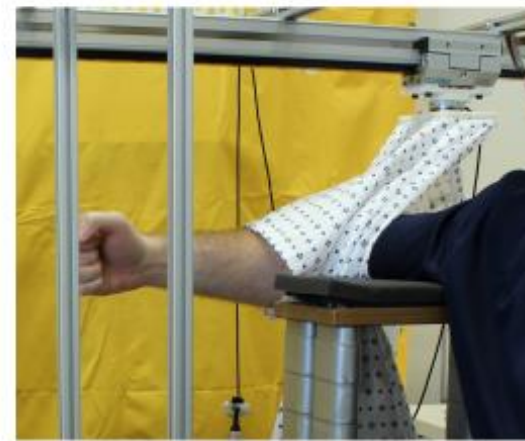
(a) Starting condition



(b) Missed outcome



(c) Caught outcome

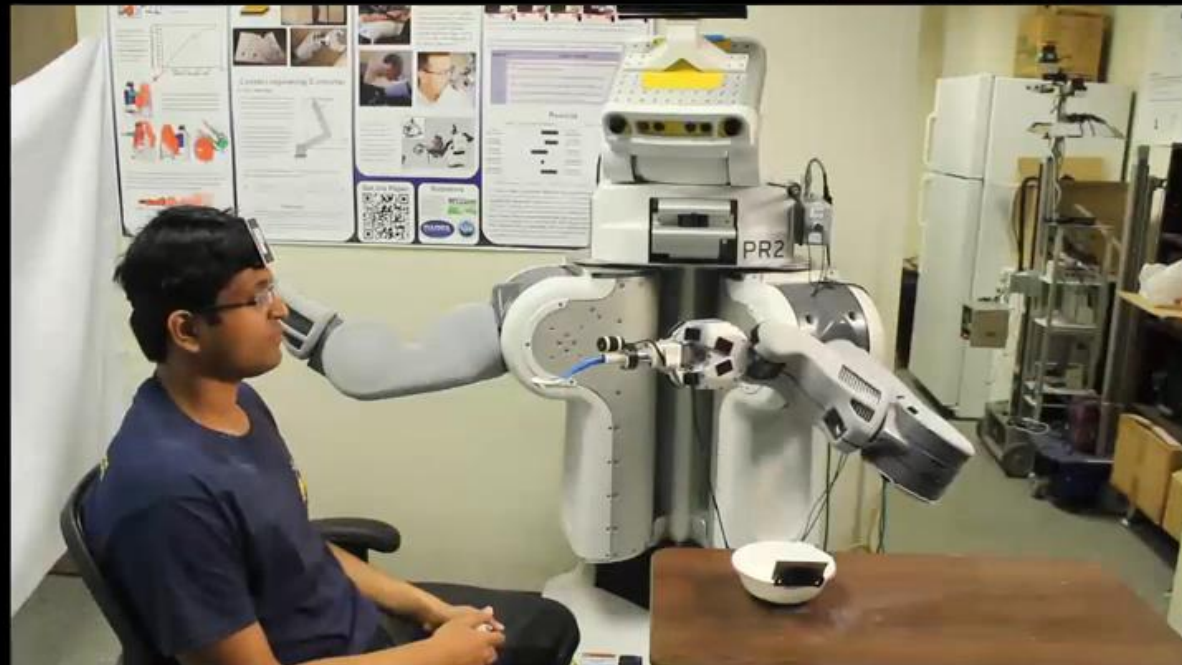
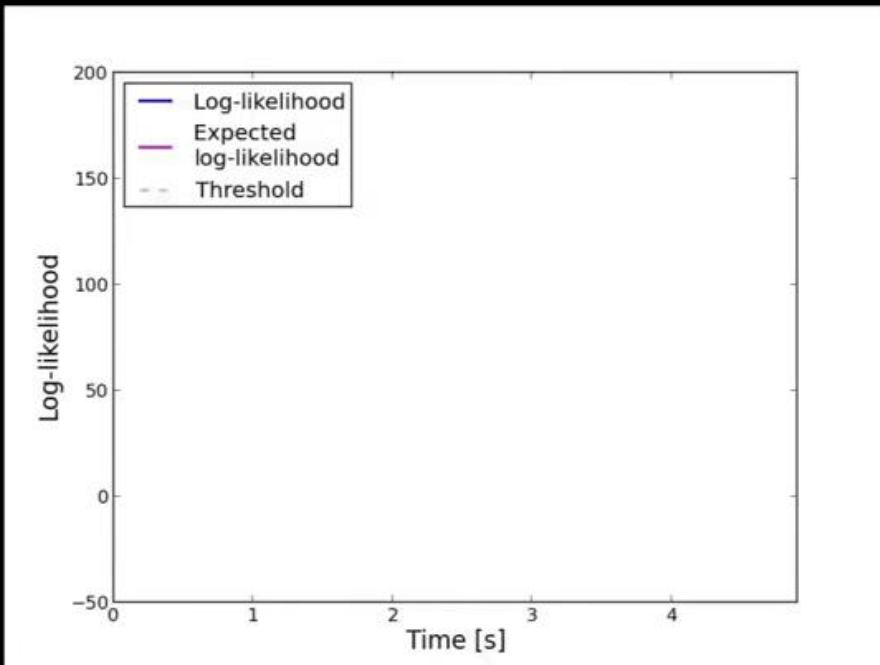
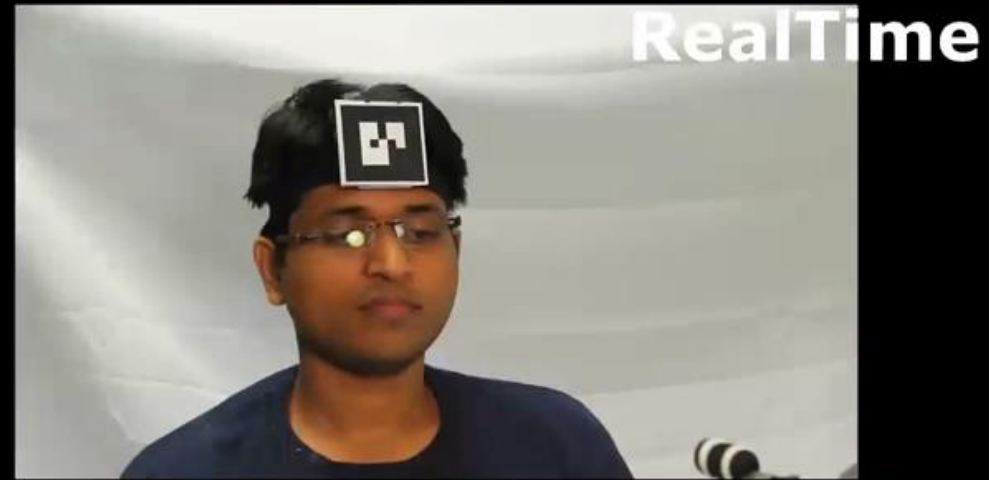


(d) Good outcome

# An anomalous feeding task

-- Current log-likelihood

-- Dynamic Threshold



Daehyung Park, Zackory Erickson, Tapomayukh Bhattacharjee, and Charles C. Kemp, *Multimodal Execution Monitoring for Anomaly Detection During Robot Manipulation*, IEEE International Conference on Robotics and Automation (ICRA), 2016.

# Data-Driven Models of Forces for Robot-Assisted Tasks

- Common sense about forces during tasks
  - Detect anomalies and unsafe situations
  - Recognize object instances and classes
  - Infer task-relevant state
- Forces are useful for sharing
- Data collection challenge
  - Handheld device
  - Human participants

# **Whole-Arm Tactile Sensing**

# Dominant Strategy for Robotic Manipulation has been to **Avoid Contact**

- Between the robot's arm and **the world**
- Between the robot's arm and **other parts of its body**
- Between the robot's arm and **people**



# Contact with the World is Common



Reaching a high shelf



Tying a rope to a pole



Installing a car seat



Cleaning a car trunk



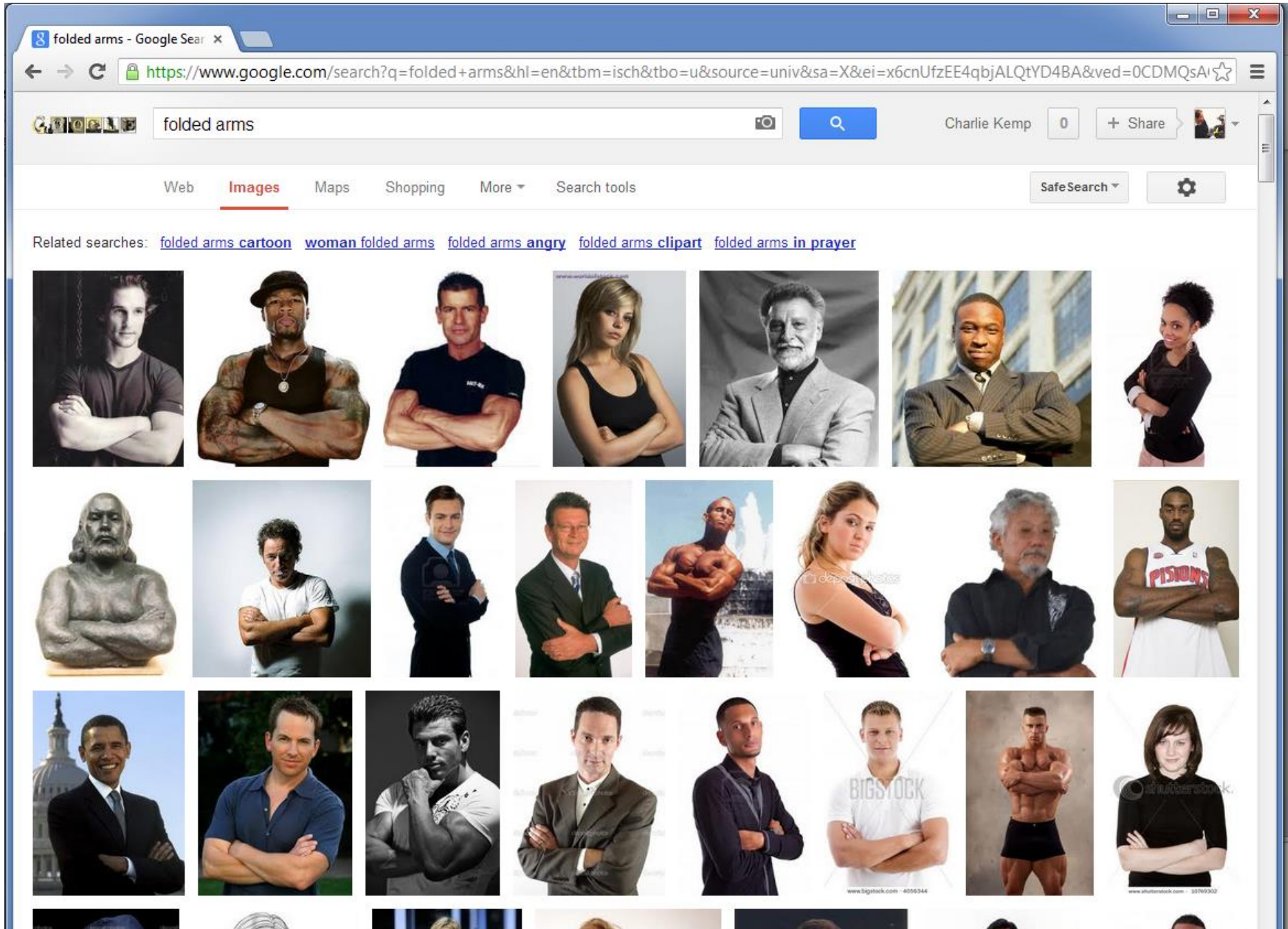
Plumbing



Carrying boxes

[images] found on the internet and used without permission

# Self-contact is Common



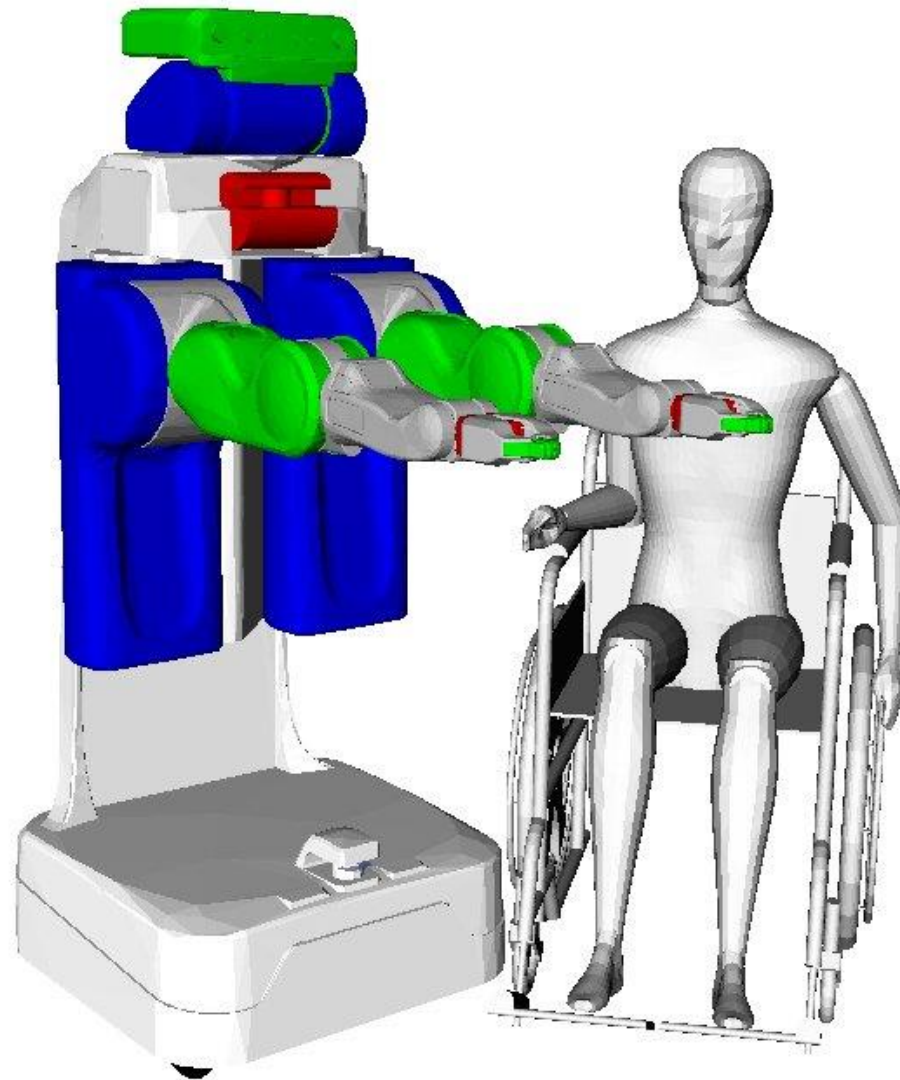
[image] Results of Google search performed by Charles C. Kemp

# Contact with People is Common (e.g., when providing assistance)



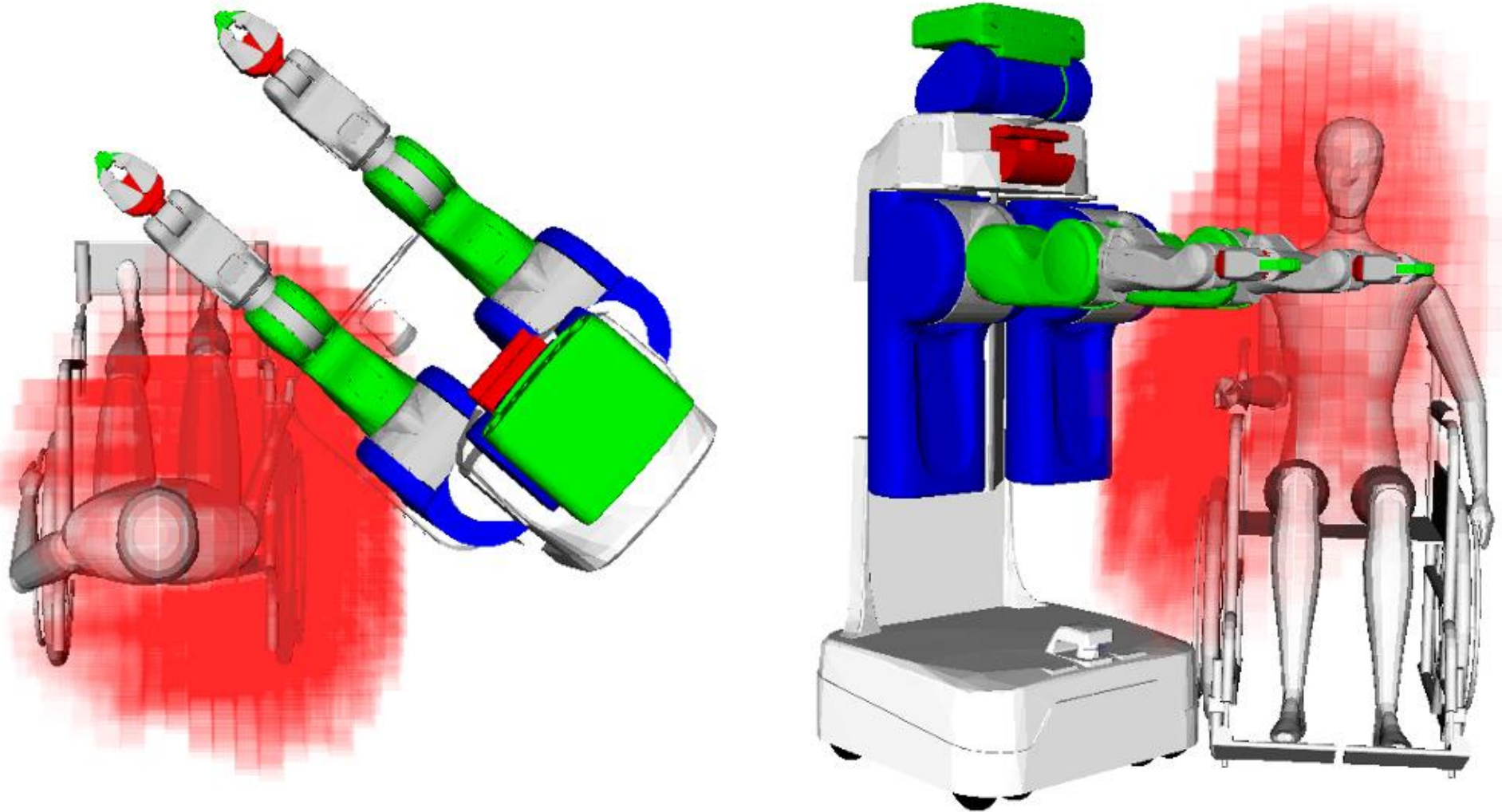
[images] found on the internet and used without permission

# Geometric Simulation to Investigate Value of Contact

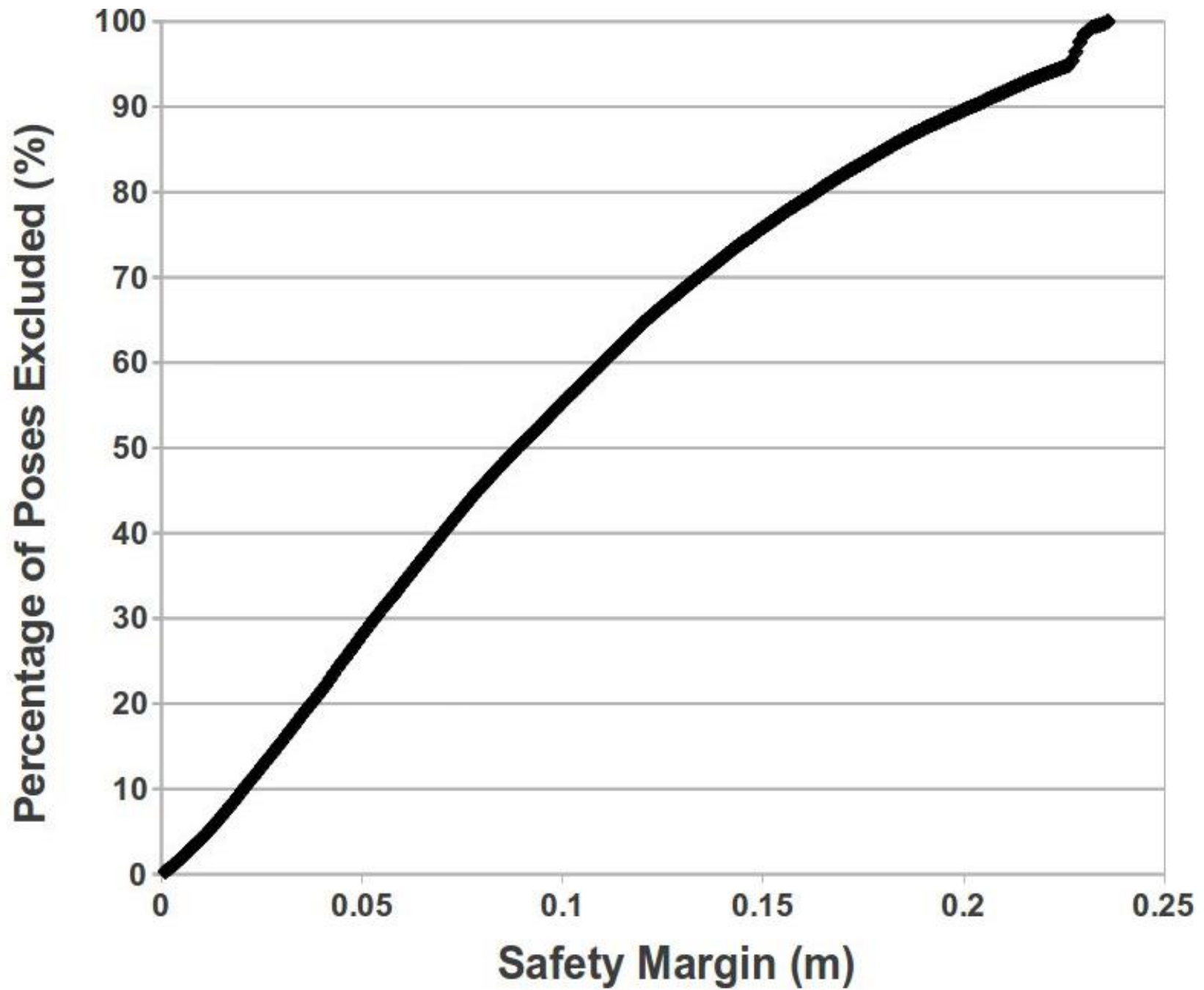


Phillip M. Grice, Marc D. Killpack, Advait Jain, Sarvagya Vaish, Jeffrey Hawke, and Charles C. Kemp, *Whole-arm Tactile Sensing for Beneficial and Acceptable Contact During Robotic Assistance*, 13th International Conference on Rehabilitation Robotics (ICORR), 2013.

# Unreachable End Effector Poses with Safety Margin of 4cm



Phillip M. Grice, Marc D. Killpack, Advait Jain, Sarvagya Vaish, Jeffrey Hawke, and Charles C. Kemp, *Whole-arm Tactile Sensing for Beneficial and Acceptable Contact During Robotic Assistance*, 13th International Conference on Rehabilitation Robotics (ICORR), 2013.

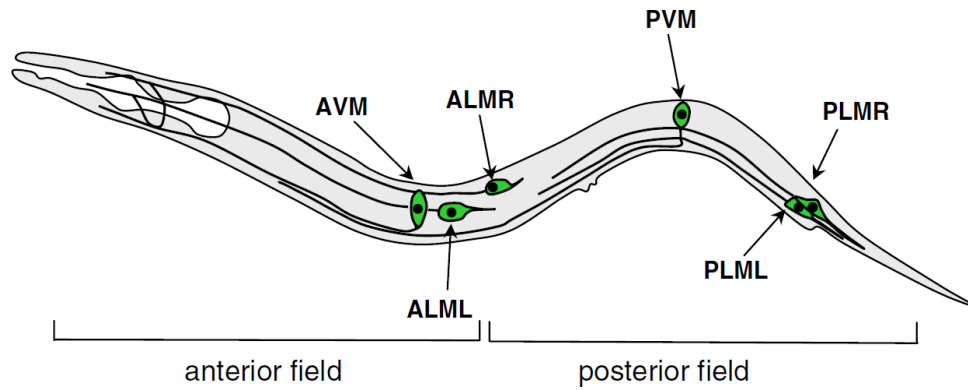


Phillip M. Grice, Marc D. Killpack, Advait Jain, Sarvagya Vaish, Jeffrey Hawke, and Charles C. Kemp, *Whole-arm Tactile Sensing for Beneficial and Acceptable Contact During Robotic Assistance*, 13th International Conference on Rehabilitation Robotics (ICORR), 2013.

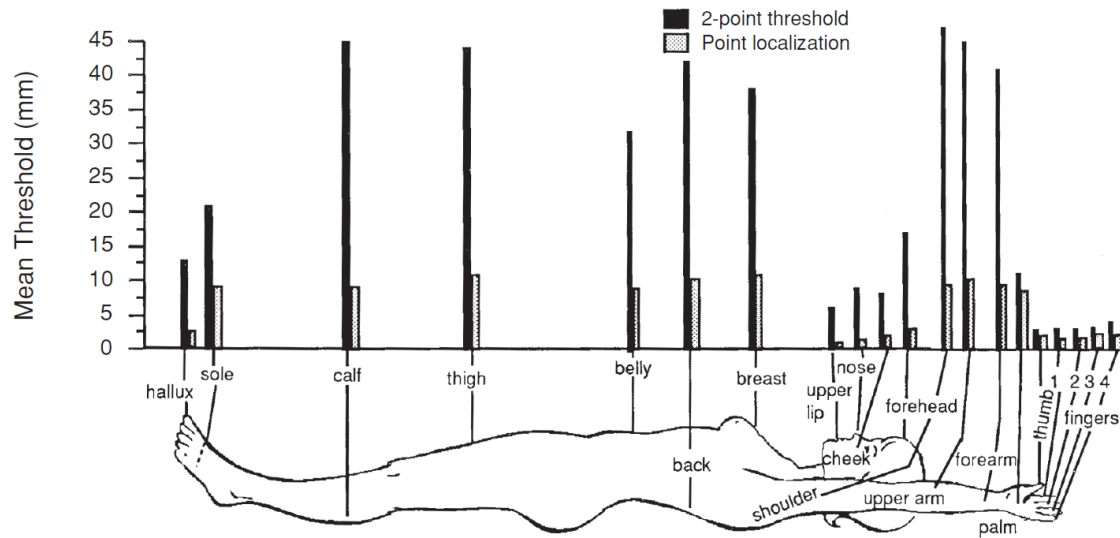
# Controllers that Allow Contact

- Assume
  - Low contact forces have no associated penalty
  - The robot has
    - Low-stiffness compliant joints
    - Whole-arm tactile sensing

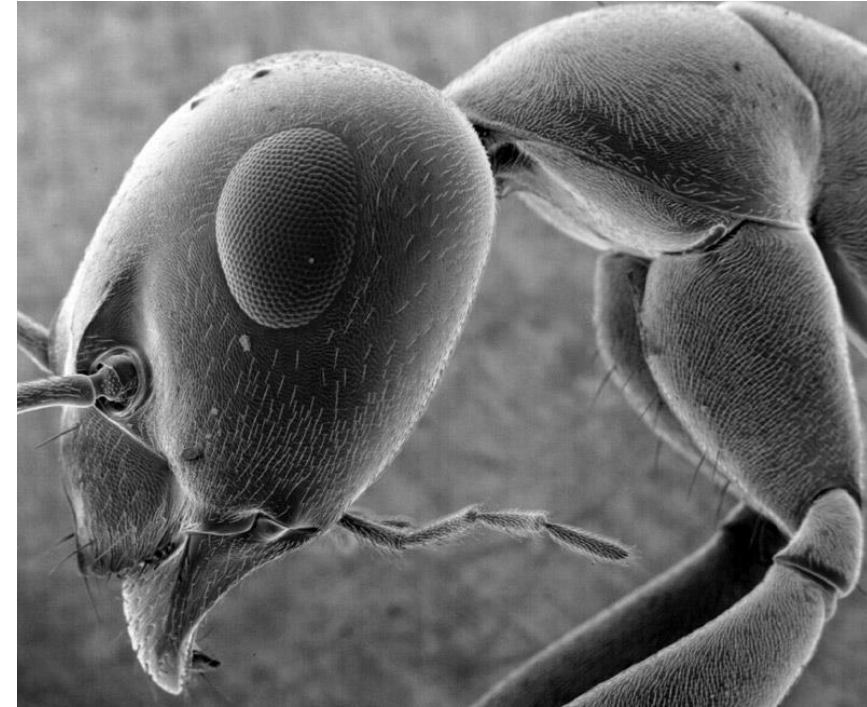
# Whole-body tactile sensing is everywhere.



Nematode (~mm)



Human (~m)



Ant (~cm)

[image of nematode] Bianchi L, *Mechanotransduction: Touch and feel at the molecular level as modeled in caenorhabditis elegans*. *Molecular Neurobiology* 36(3): 254–271, 2007.

[image of ant] from the Dartmouth College Electron Microscope Facility, <http://remf.dartmouth.edu/images/insectPart3SEM/source/31.html>

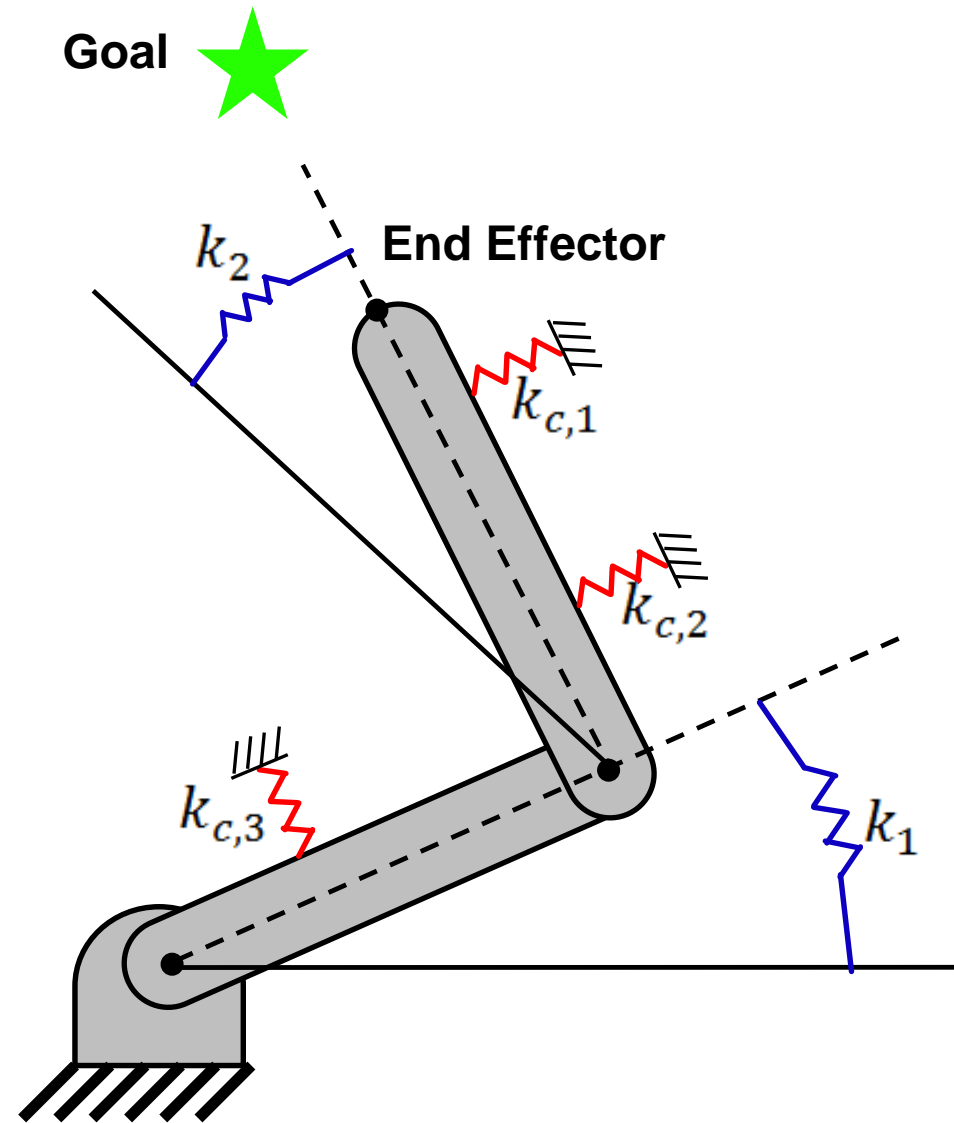
[image of human] Lederman, Susan J., and Roberta L. Klatzky. *Haptic perception: A tutorial*. *Attention, Perception, & Psychophysics* 71.7, 1439-1459, 2009.

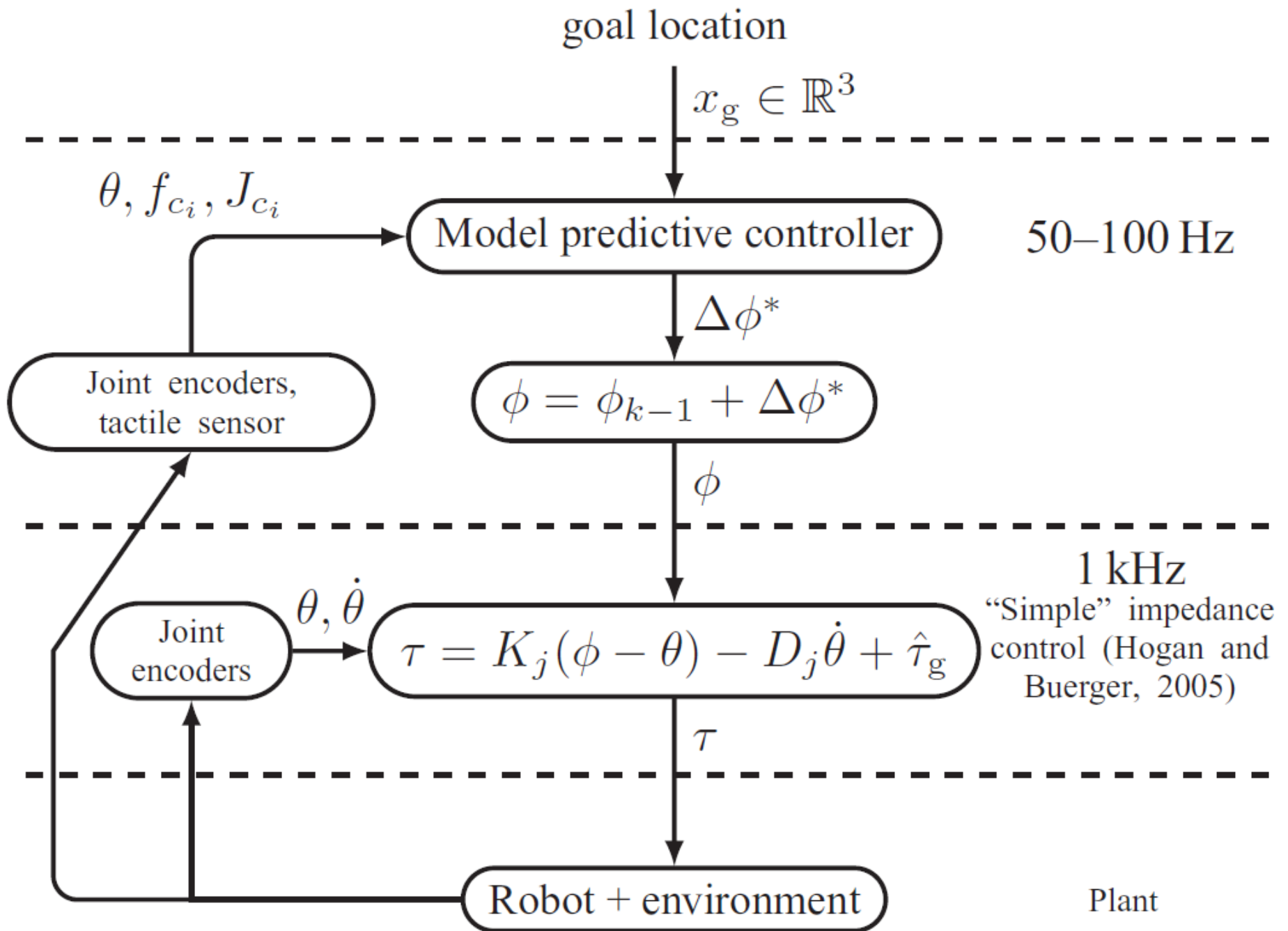


# Quasistatic Model Predictive Control

At each time step

- Generate quasi-static model
- Use quadratic programming to find a change to the equilibrium angles of the joints that
  - Minimizes the predicted distance to the goal
  - Subject to constraints on predicted contact forces

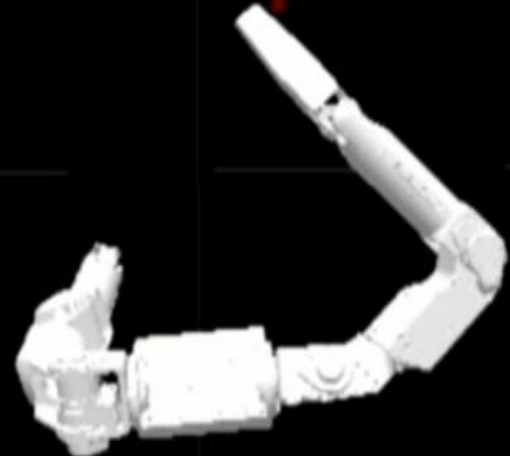




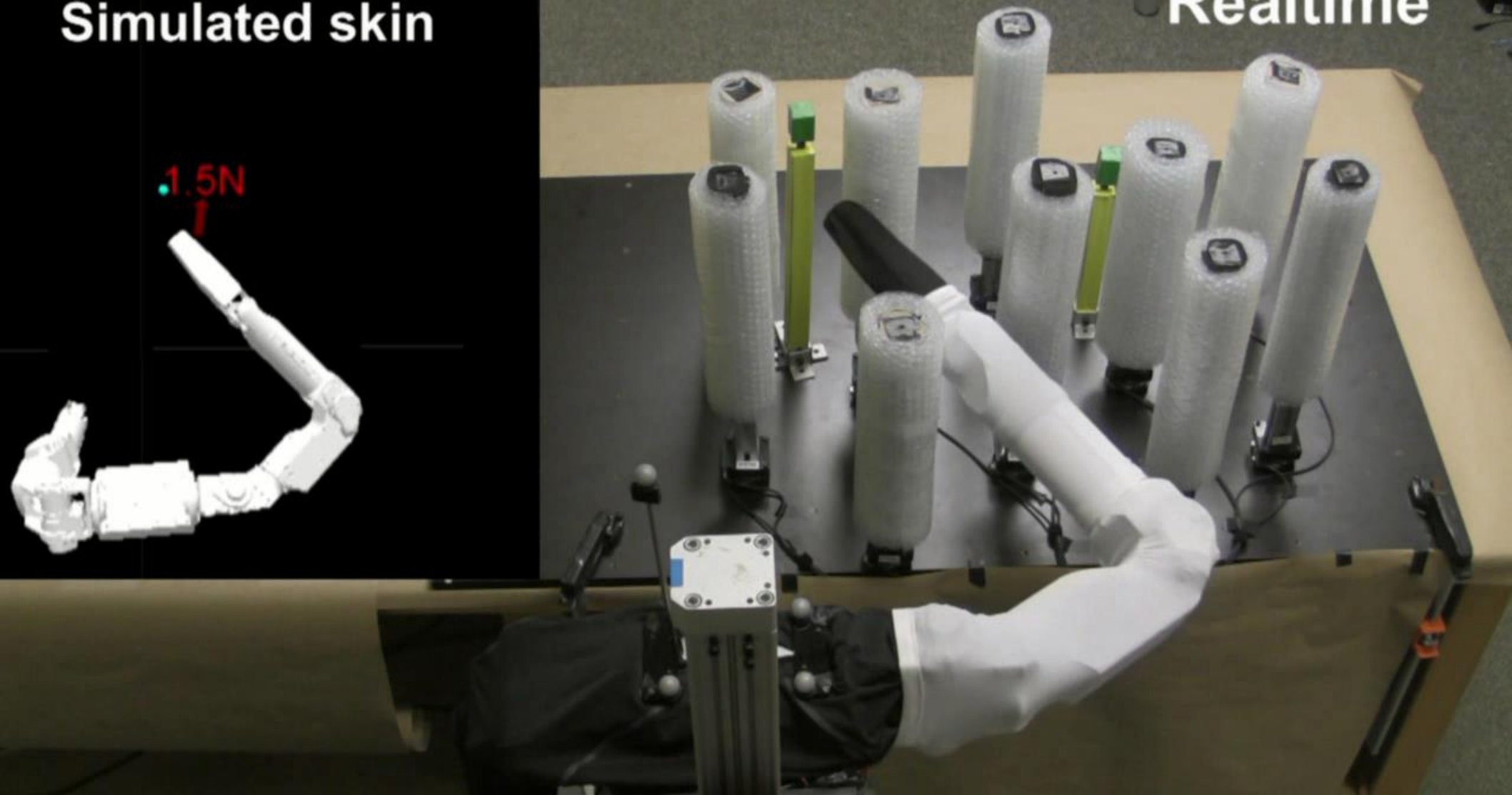
Advait Jain, Marc D. Killpack, Aaron Edsinger, and Charles C. Kemp, *Reaching in clutter with whole-arm tactile sensing*. The International Journal of Robotics Research, 32.4 (2013): 458-482.

Simulated skin

1.5N



Realtime

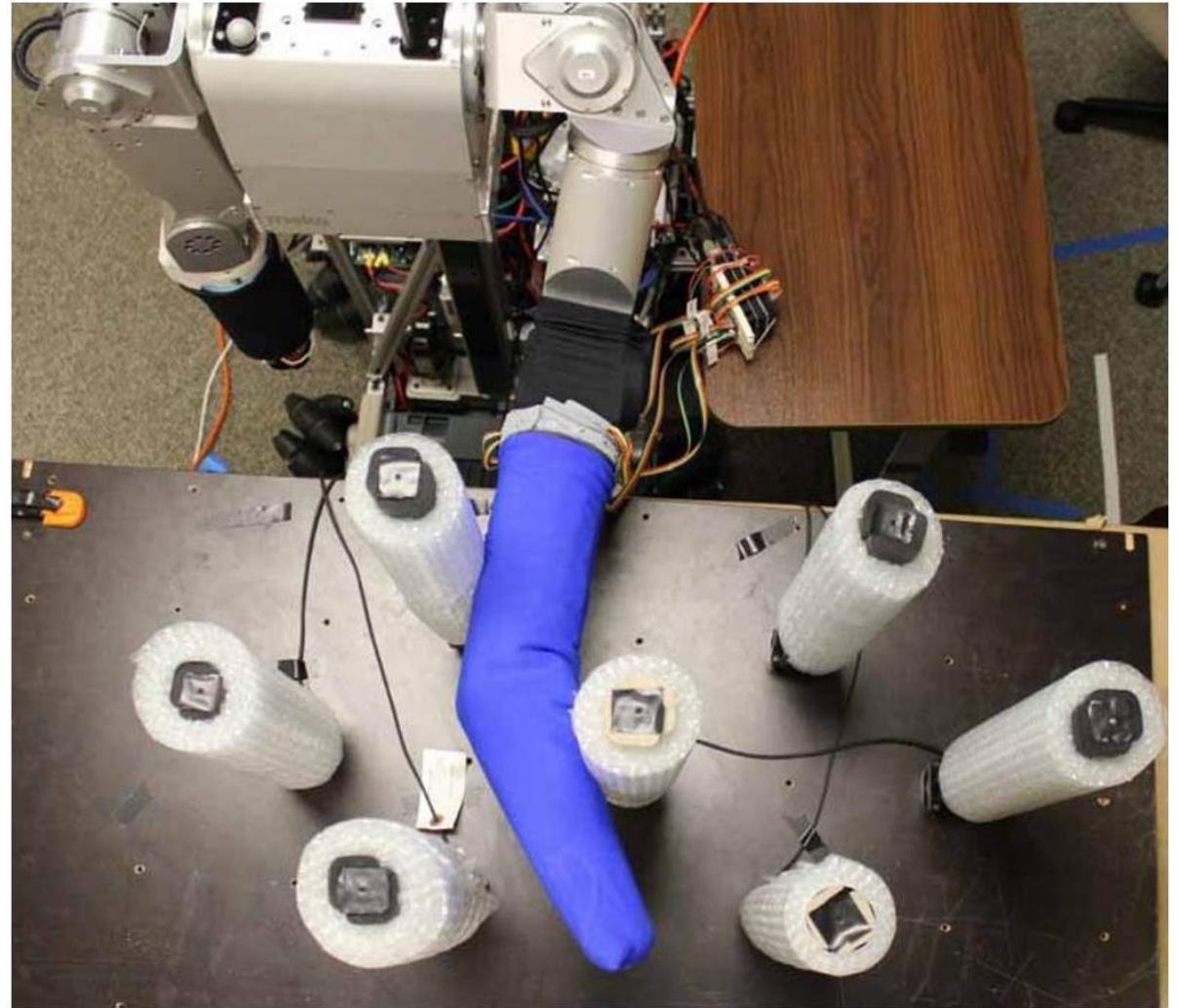


Advait Jain, Marc D. Killpack, Aaron Edsinger, and Charles C. Kemp, *Reaching in clutter with whole-arm tactile sensing*. The International Journal of Robotics Research, 32.4 (2013): 458-482.



Marc D. Killpack, Ariel Kapusta, and Charles C. Kemp, *Model predictive control for fast reaching in clutter*, *Autonomous Robots*, 2015.

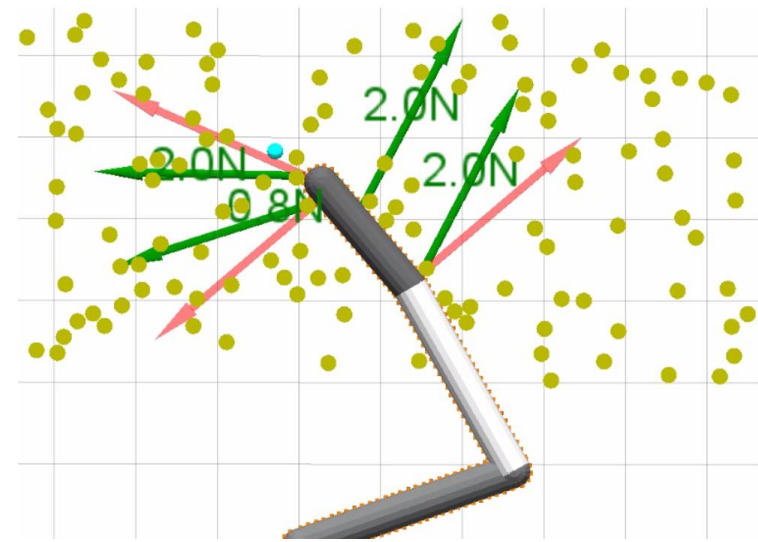
# Contact at the Joints



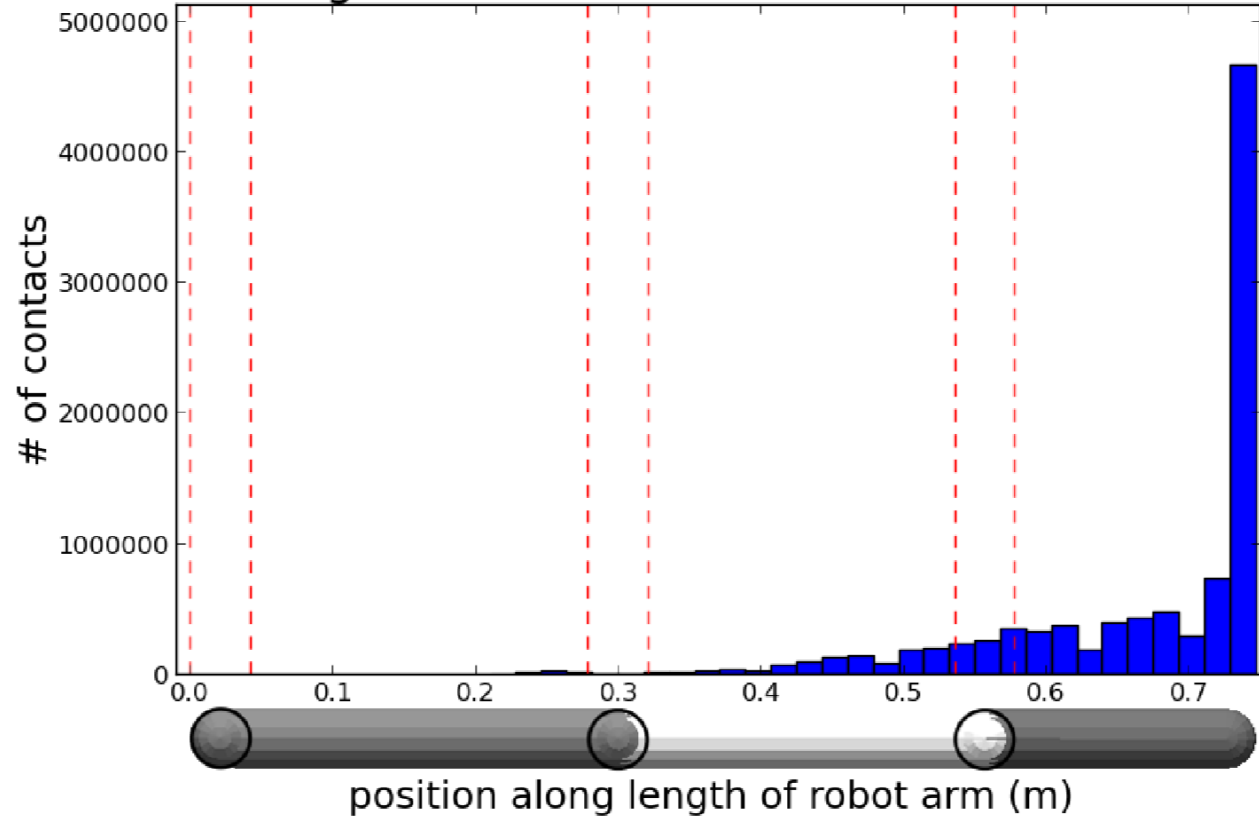
Tapomayukh Bhattacharjee, Advait Jain, Sarvagya Vaish, Marc D. Killpack, and Charles C. Kemp, *Tactile Sensing over Articulated Joints with Stretchable Sensors*, IEEE World Haptics Conference (WHC 2013), 2013.

# Contact at the Joints

Reaching Task

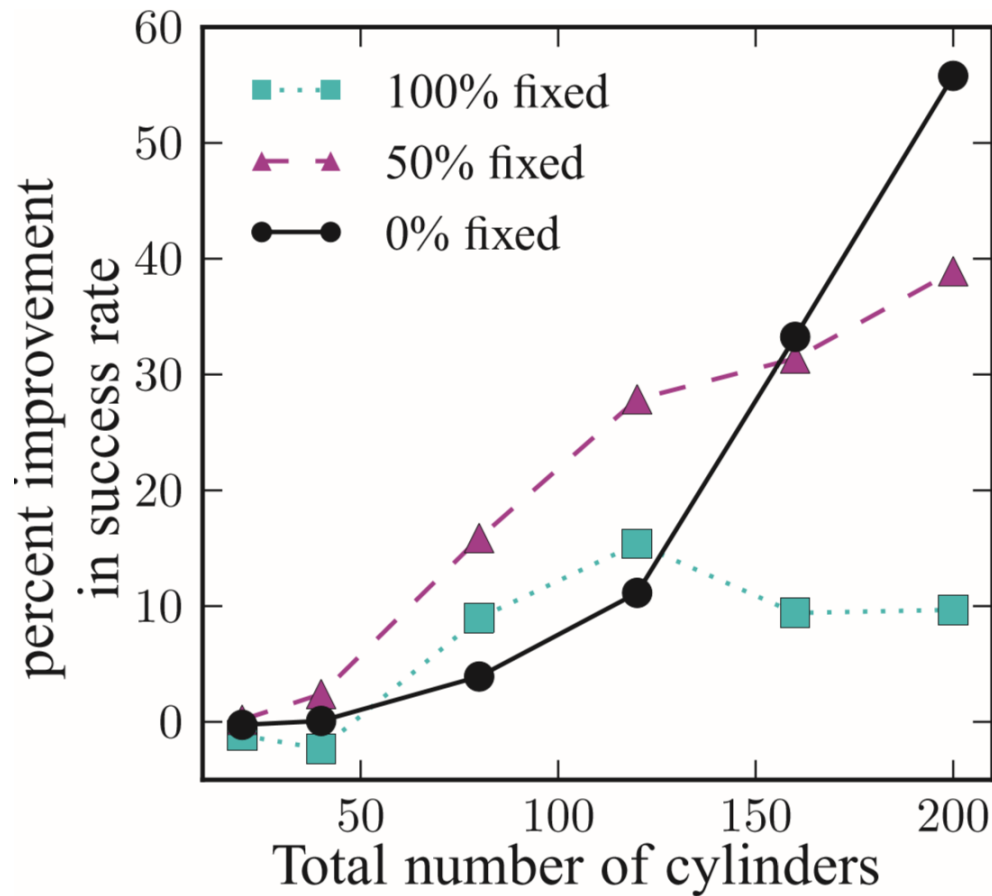


Histogram of contact locations on robot arm

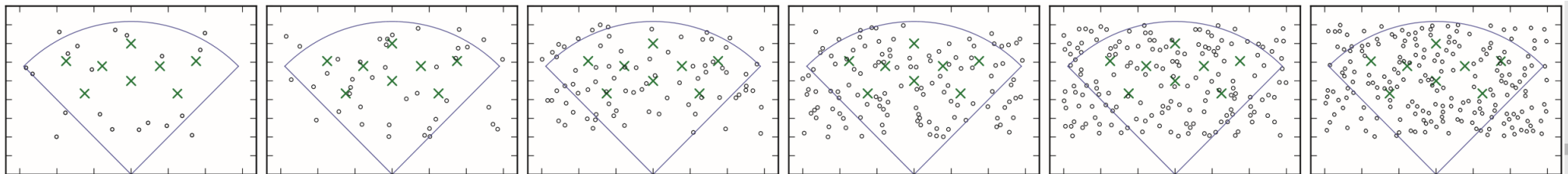


Tapomayukh Bhattacharjee, Advait Jain, Sarvagya Vaish, Marc D. Killpack, and Charles C. Kemp, *Tactile Sensing over Articulated Joints with Stretchable Sensors*, IEEE World Haptics Conference (WHC 2013), 2013.

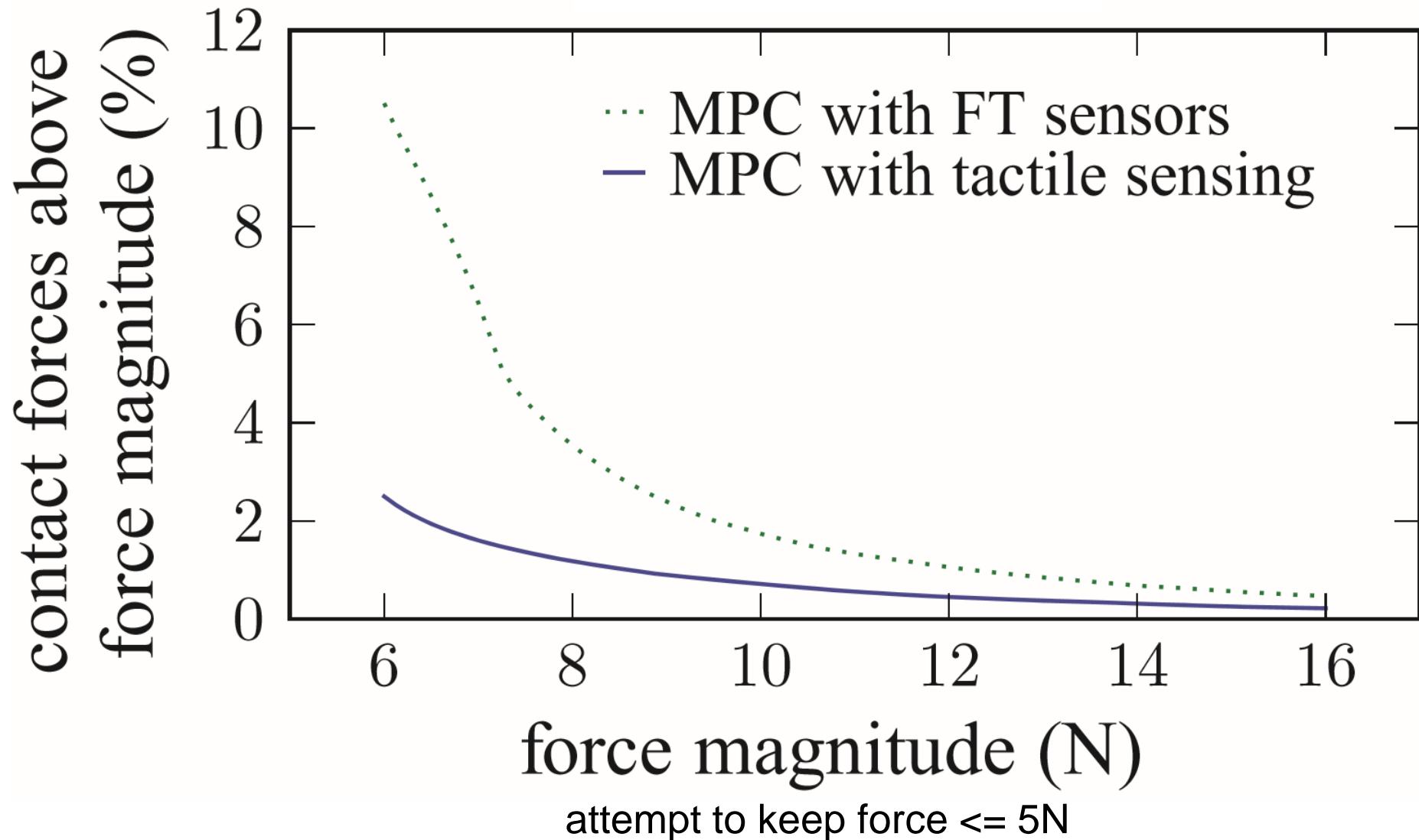
# Higher success rate with tactile sensing compared to per-link force-torque sensing



64,800 trials  
in simulation

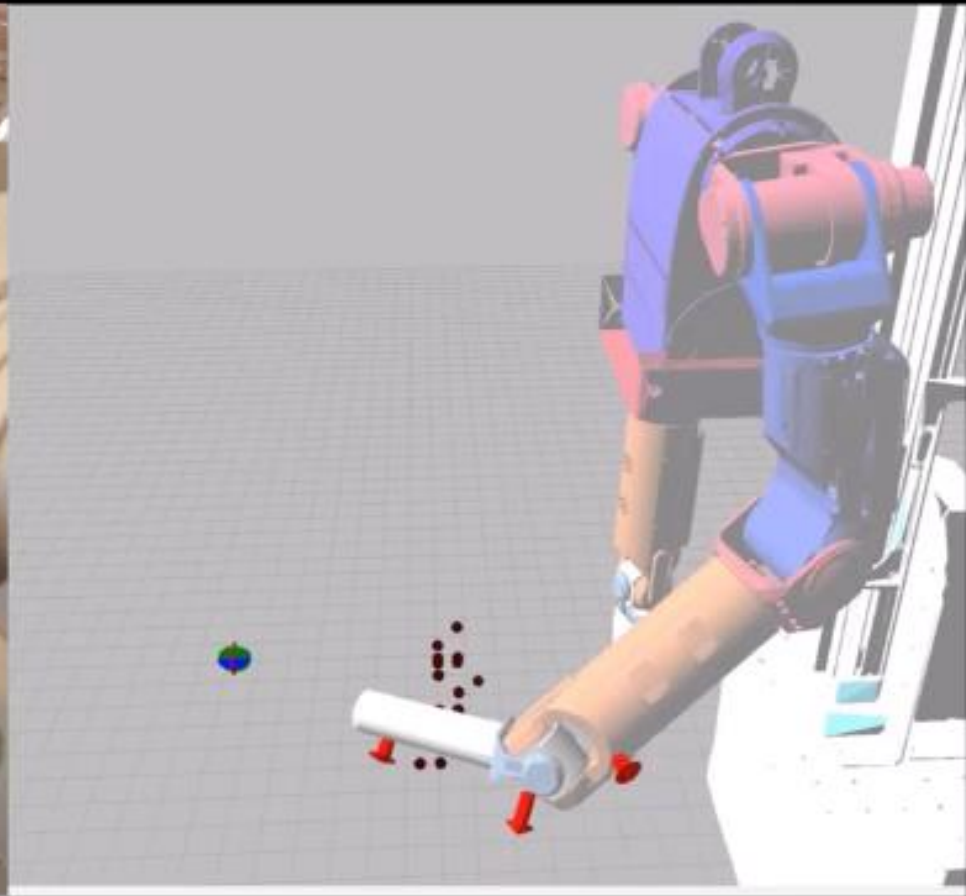


# Higher contact forces with per-link force-torque sensing





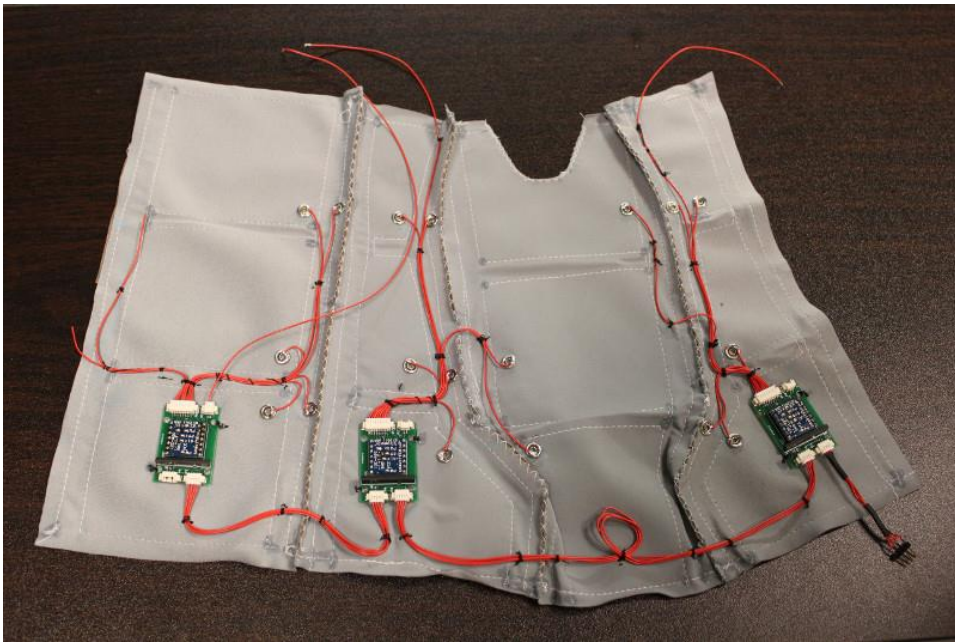
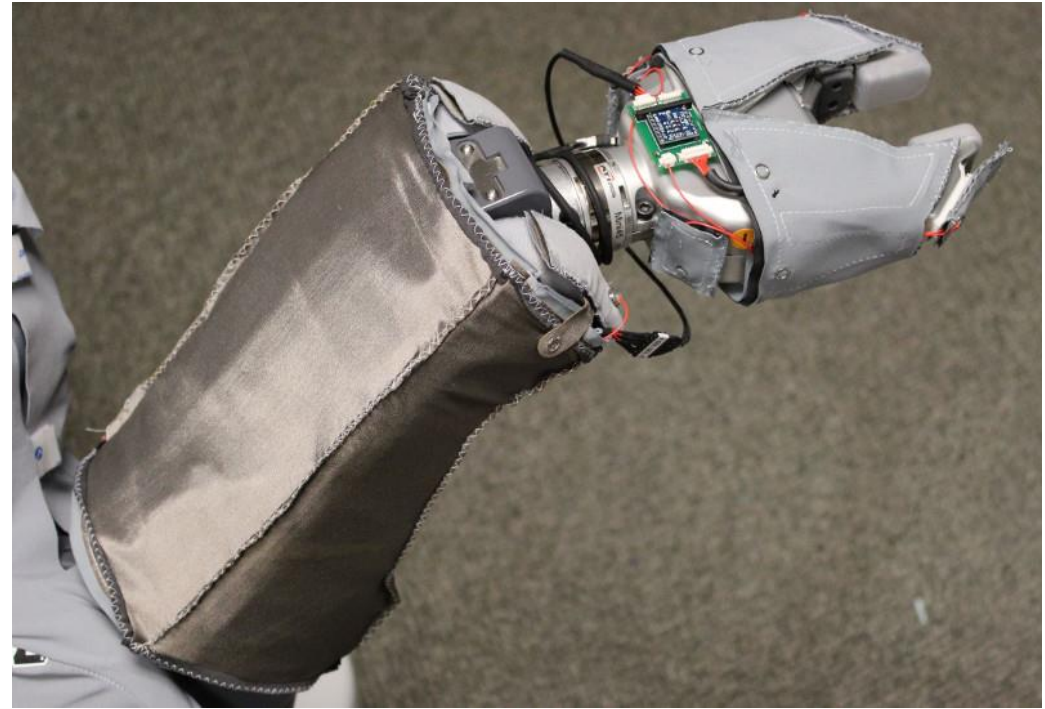
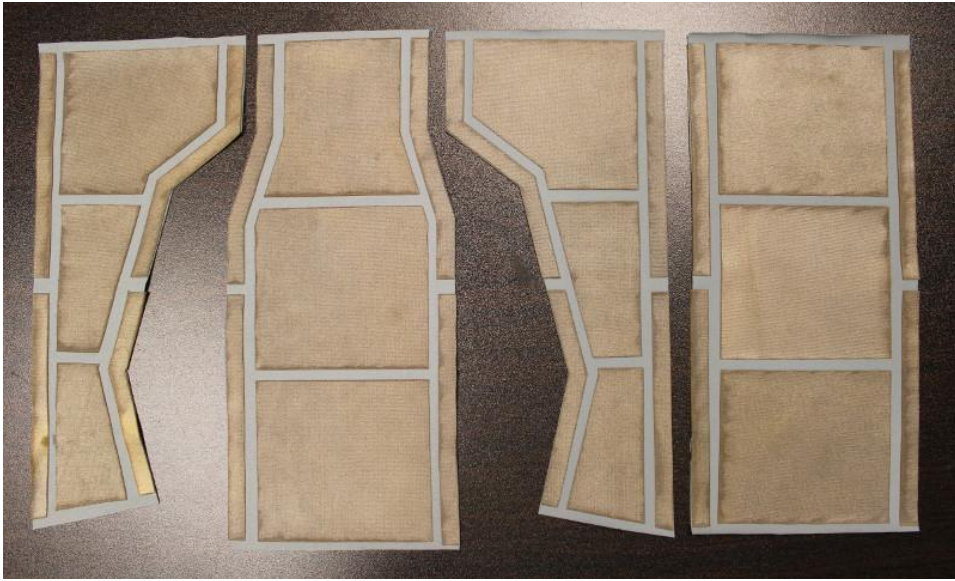
# Real-time Haptic Mapping



Realtime

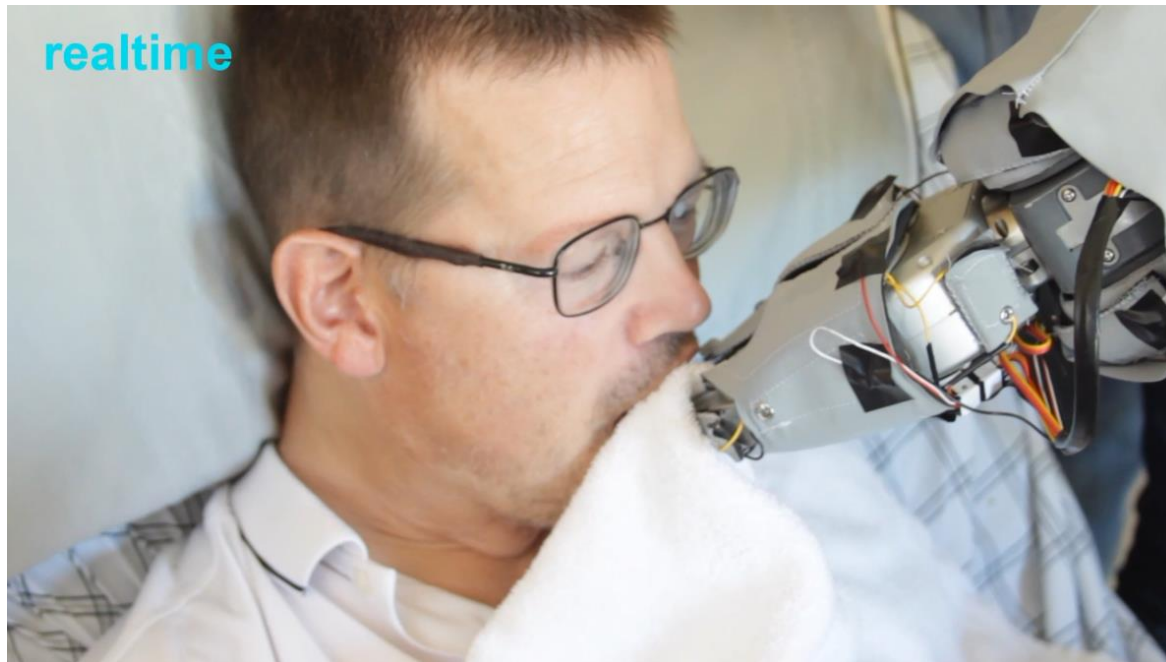
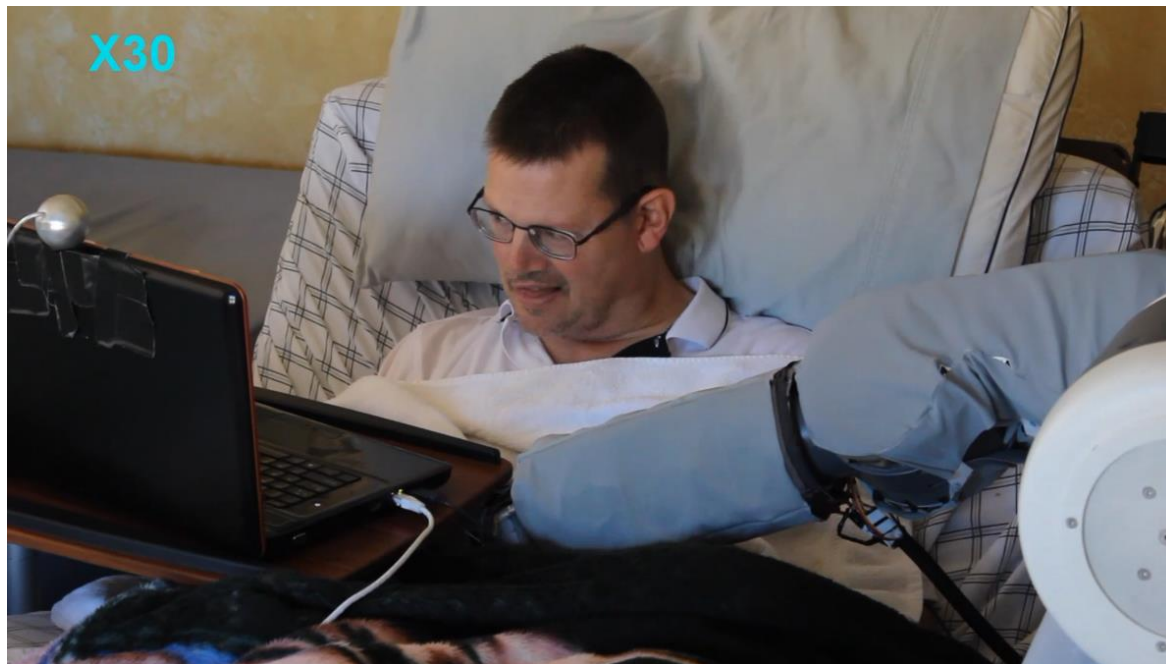
Tapomayukh Bhattacharjee, Phillip M. Grice, Ariel Kapusta, Marc D. Killpack, Daehyung Park, and Charles C. Kemp, *A Robotic System for Reaching in Dense Clutter that Integrates Model Predictive Control, Learning, Haptic Mapping, and Planning*, IROS 2014 workshop: 3rd Workshop on Robots in Clutter: Perception and Interaction in Clutter, 2014.

# Stretchable Fabric Tactile Sensors



- **41 discrete tactile sensor elements (taxels)**
  - 3 on upper arm
  - 22 on forearm
  - 16 on the gripper
- **Open hardware**

Phillip M. Grice, Marc D. Killpack, Advait Jain, Sarvagya Vaish, Jeffrey Hawke, and Charles C. Kemp, *Whole-arm Tactile Sensing for Beneficial and Acceptable Contact During Robotic Assistance*, 13th International Conference on Rehabilitation Robotics (ICORR), 2013.



# Picking Up a Cloth and Wiping Face in Bed

Phillip M. Grice, Marc D. Killpack, Advait Jain, Sarvagya Vaish, Jeffrey Hawke, and Charles C. Kemp, *Whole-arm Tactile Sensing for Beneficial and Acceptable Contact During Robotic Assistance*, 13th International Conference on Rehabilitation Robotics (ICORR), 2013.

# Grasping and Pulling up a Blanket in Bed



Phillip M. Grice, Marc D. Killpack, Advait Jain, Sarvagya Vaish, Jeffrey Hawke, and Charles C. Kemp, *Whole-arm Tactile Sensing for Beneficial and Acceptable Contact During Robotic Assistance*, 13th International Conference on Rehabilitation Robotics (ICORR), 2013.

# Henry Evans's Original Comments

During the tests:

“It is very compliant”

**“I like it.”**

“I think it's a good safety feature because **it hardly presses against me** even when I tell it to.”

**“It really feels safe to be close to the robot.”**

A week after the tests:

“Skin

**Overall awesome**

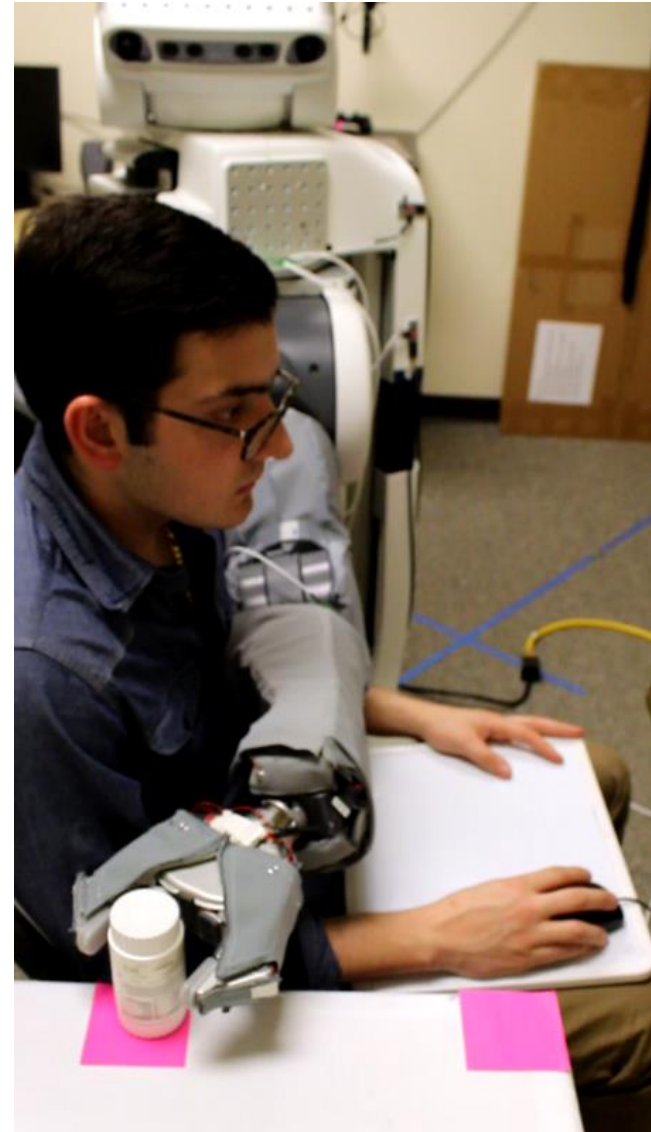
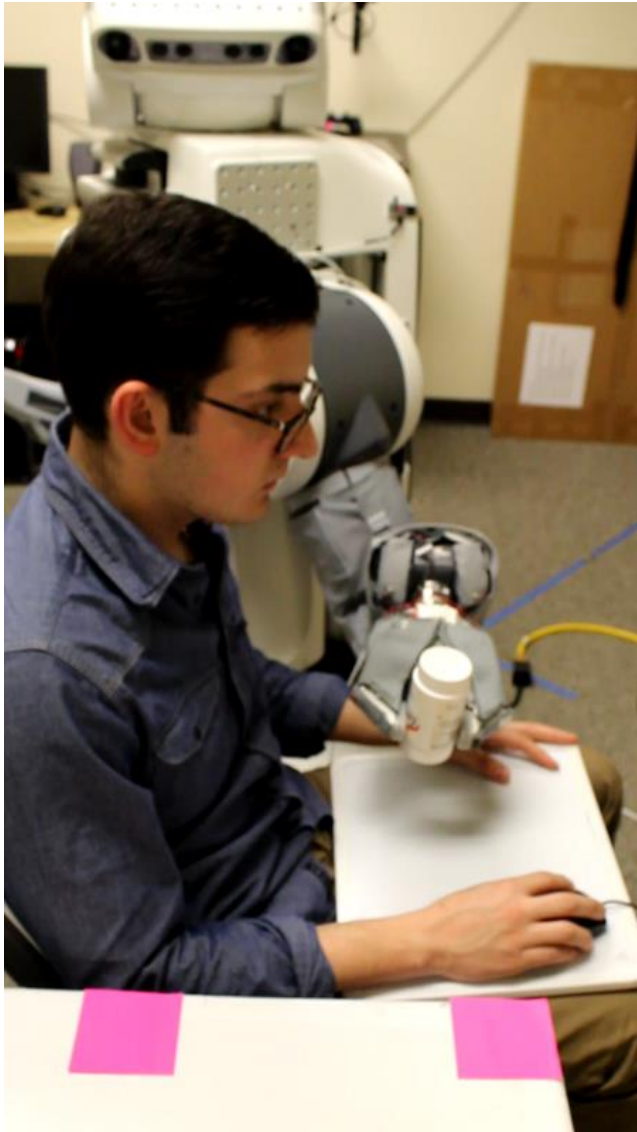
**Feels VERY safe**

Faster than motion planning

**It just wriggles around obstacles”**

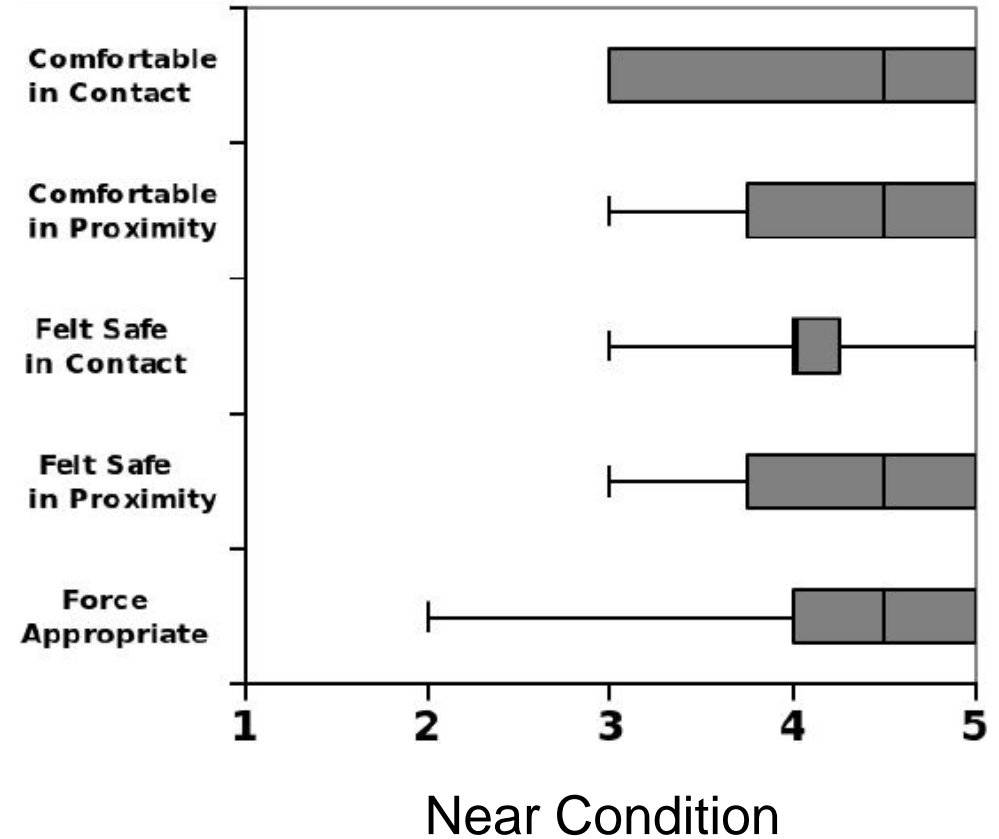
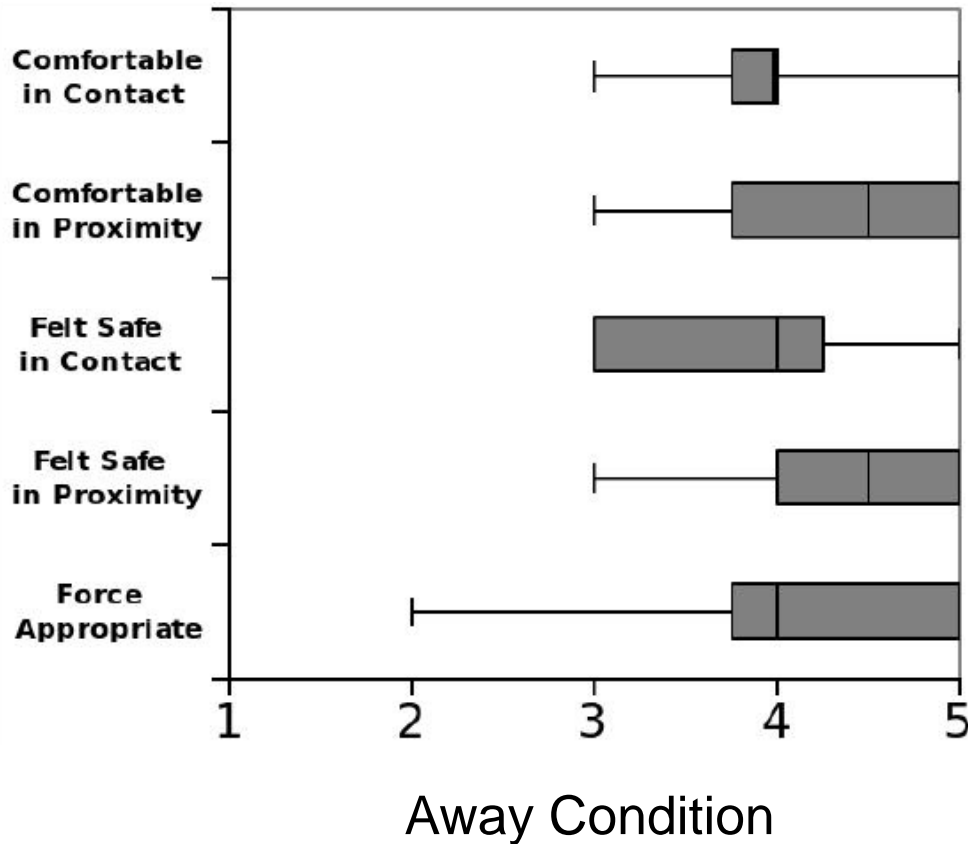
**“DEFINITELY keep developing this !”**

# Will contact be acceptable to others?



Phillip M. Grice, Marc D. Killpack, Advait Jain, Sarvagya Vaish, Jeffrey Hawke, and Charles C. Kemp, *Whole-arm Tactile Sensing for Beneficial and Acceptable Contact During Robotic Assistance*, 13th International Conference on Rehabilitation Robotics (ICORR), 2013.

# 8 Able-bodied Participants

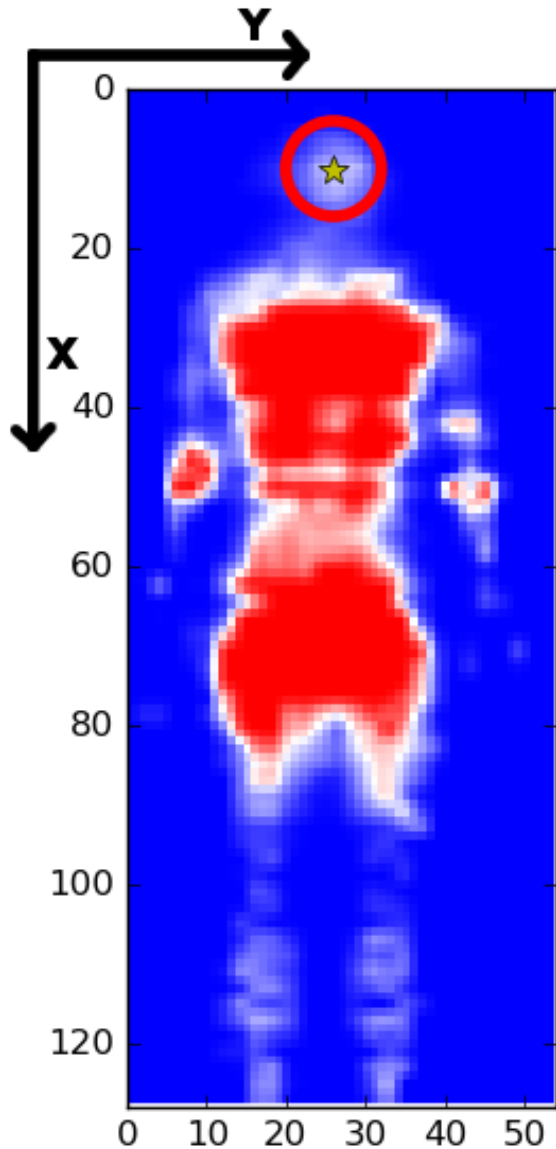


1: Strongly Disagree, 2: Disagree, 3: Neither Agree nor Disagree, 4: Agree, 5: Strongly Agree

# **Whole-Body Tactile Sensing for Unconventional Robots**



# Perceptual Collaboration Between Robots



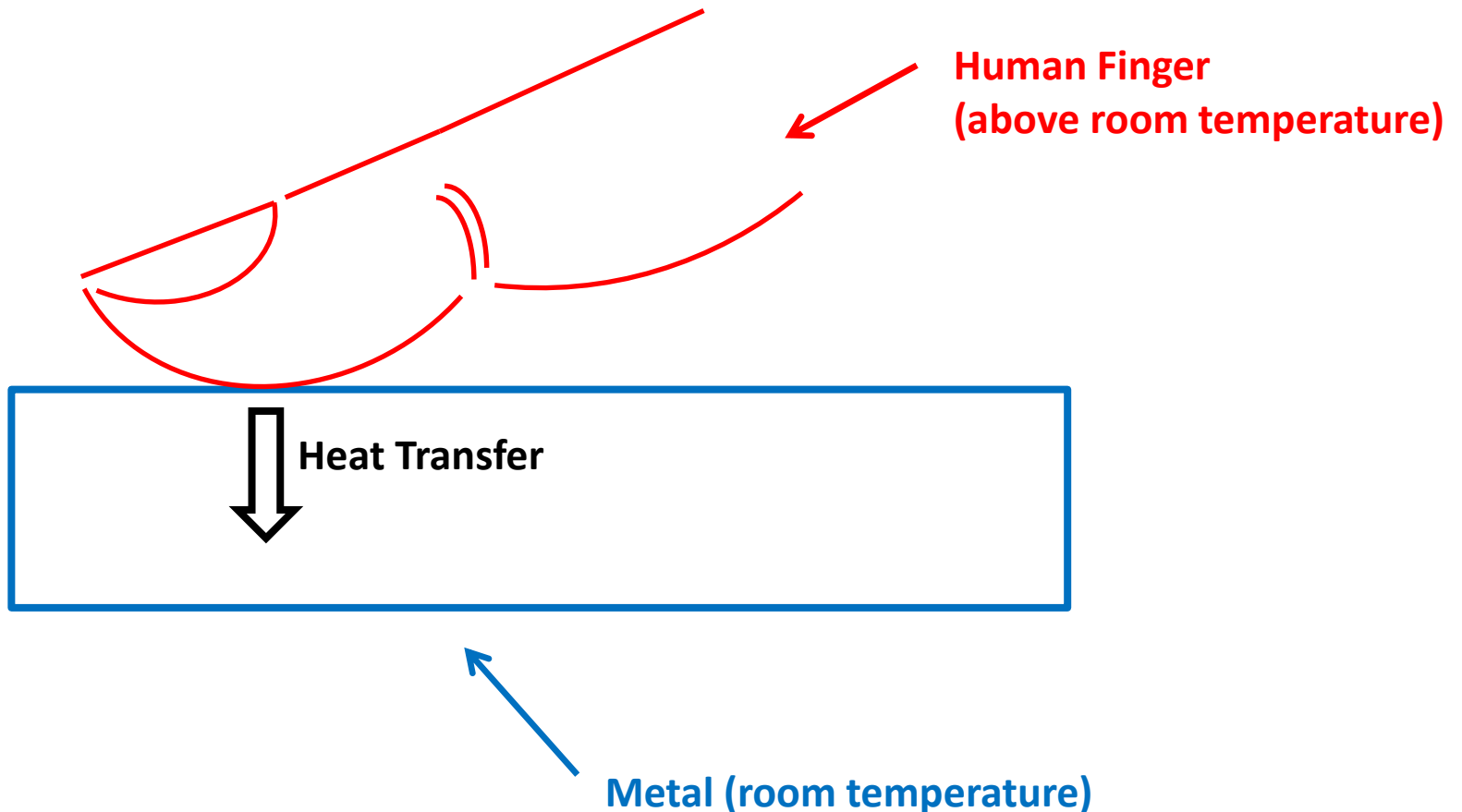
Ariel Kapusta, Yash Chitalia, Daehyung Park, and Charles C. Kemp, *Collaboration Between a Robotic Bed and a Mobile Manipulator May Improve Physical Assistance for People with Disabilities*, RO-MAN 2016 Workshop on behavior adaptation, interaction and learning for assistive robots (BAILAR 2016), 2016.

# Whole-Arm Tactile Sensing

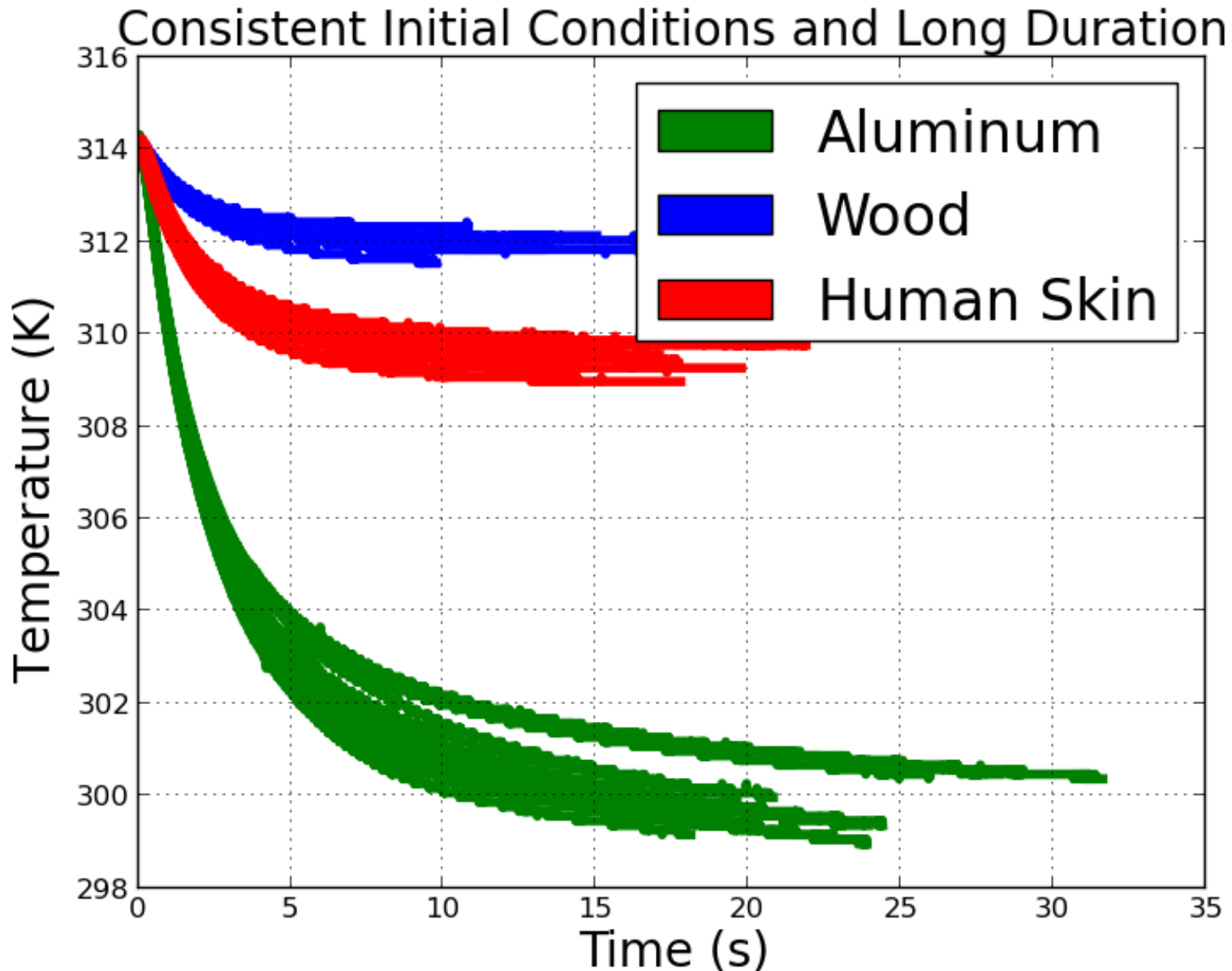
- Reach locations in clutter while keeping forces low
  - Reaching around the human body
  - Outperformed per-link force-torque sensing
- Challenge
  - Immature technology for large area tactile sensing
- Permitting contact
  - Makes more poses reachable
  - Reduces line of sight sensing requirements
  - Creates opportunities to sense through touch (e.g., incidental contact)

# **Thermal Tactile Sensing**

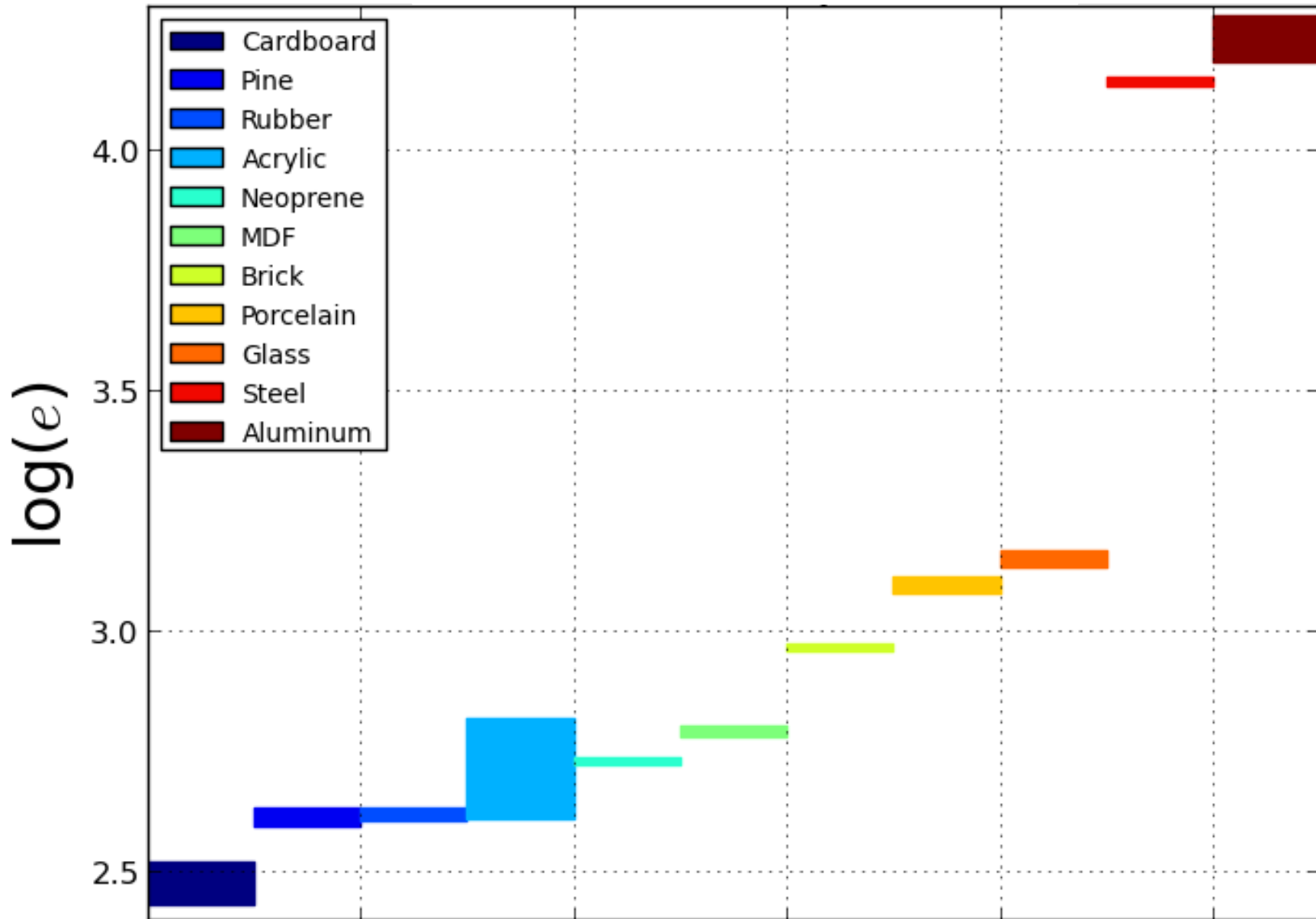
# Active Thermal Sensing



# Active Thermal Sensing



# Thermal Effusivity



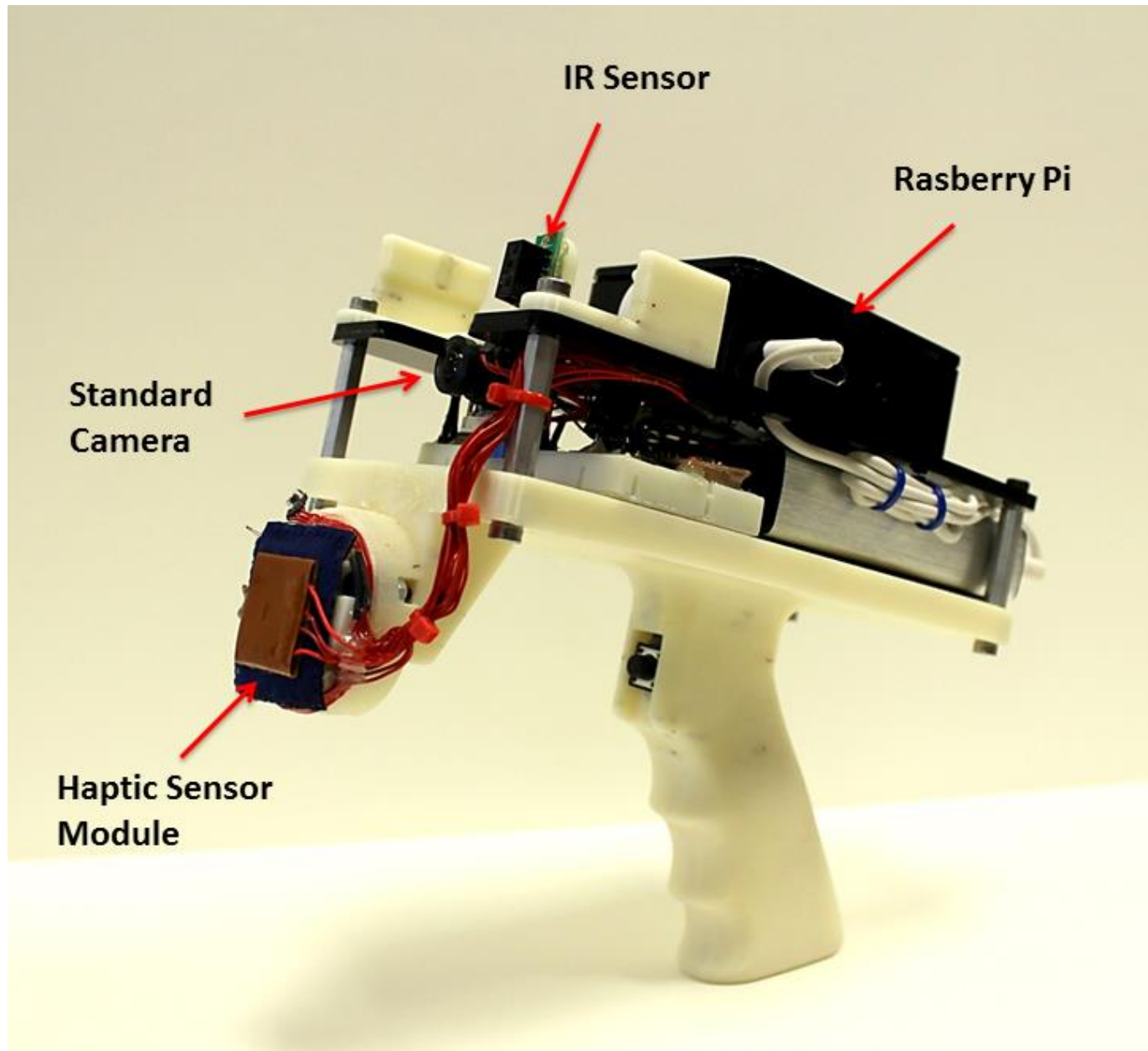
# Passive Thermal Sensing



[image] found on the internet and used without permission

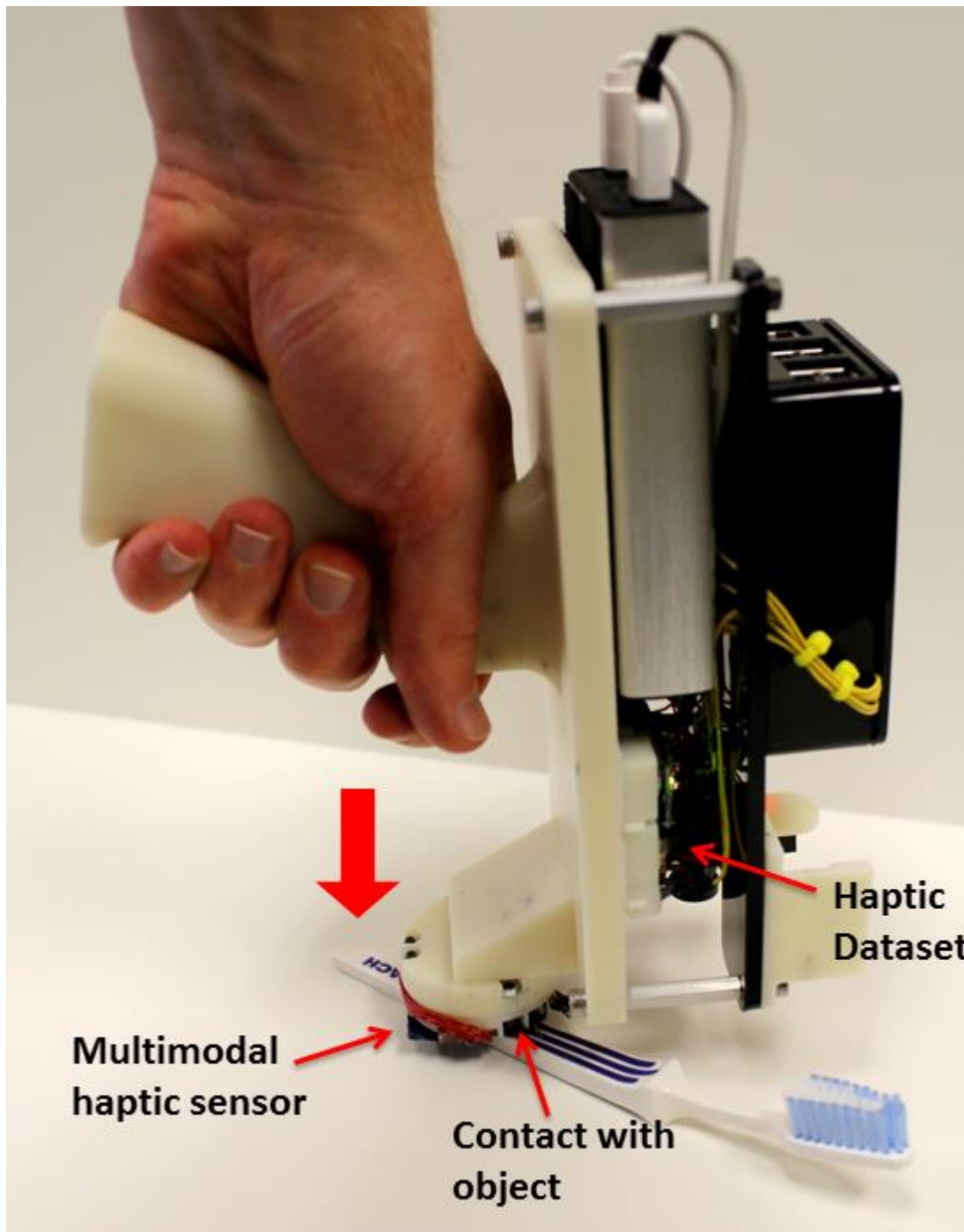
Tapomayukh Bhattacharjee, Joshua Wade, Yash Chitalia, and Charles C. Kemp, *Data-Driven Thermal Recognition of Contact with People and Objects*, IEEE Haptics Symposium, 2016.

# Data-Driven Thermal Recognition



Tapomayukh Bhattacharjee, Joshua Wade, Yash Chitalia, and Charles C. Kemp, *Data-Driven Thermal Recognition of Contact with People and Objects*, IEEE Haptics Symposium, 2016.

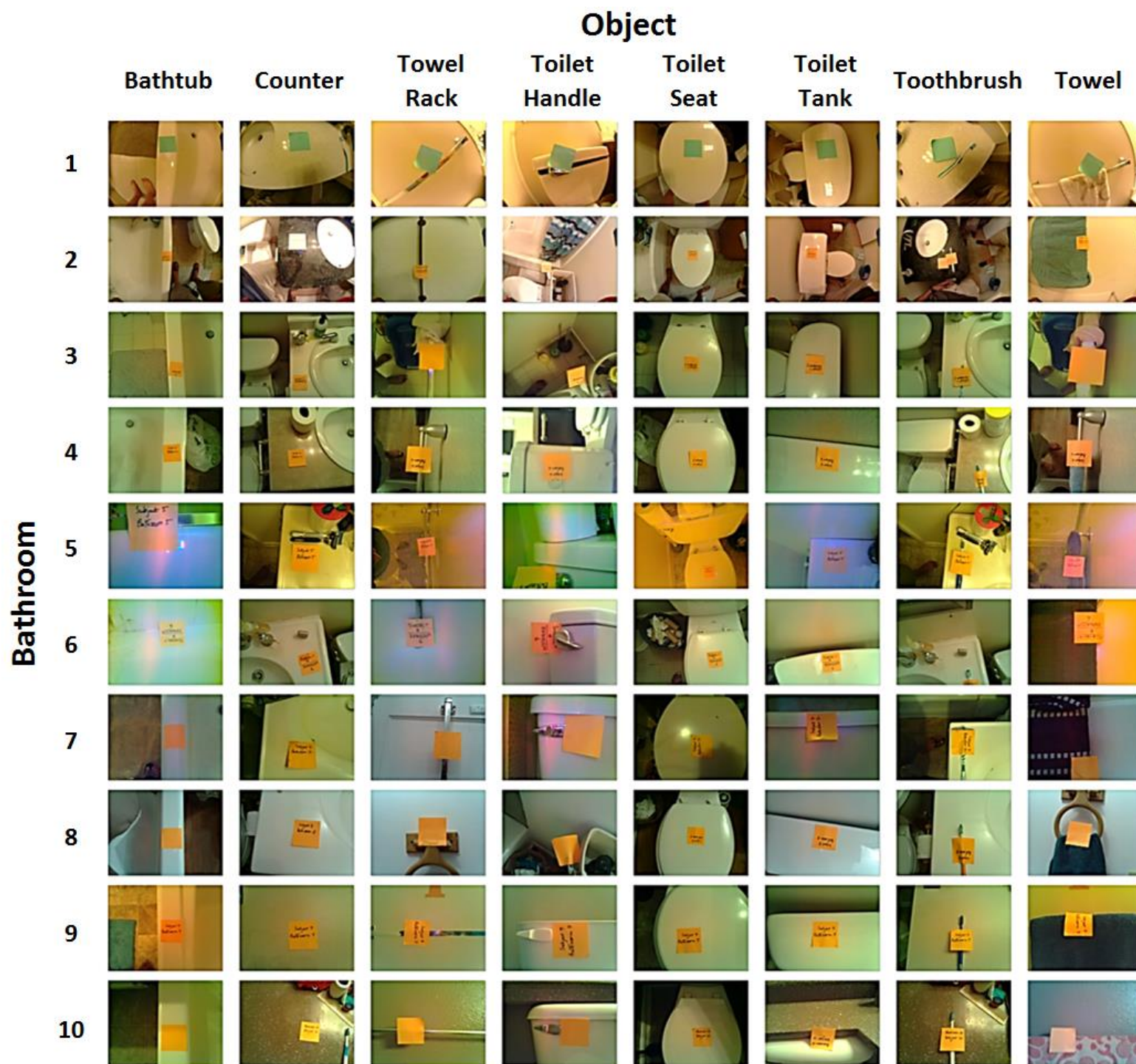




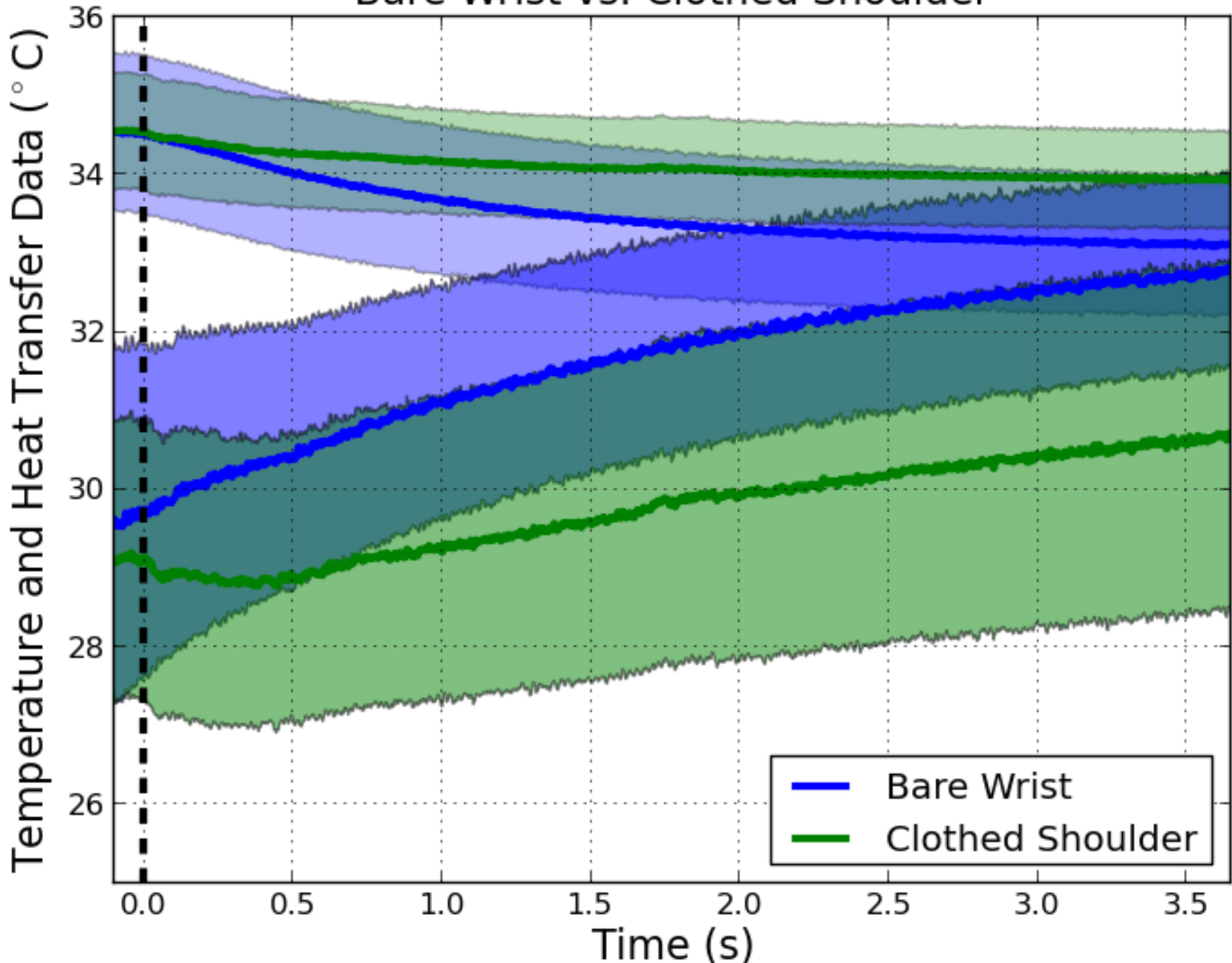
**1. Wrist**



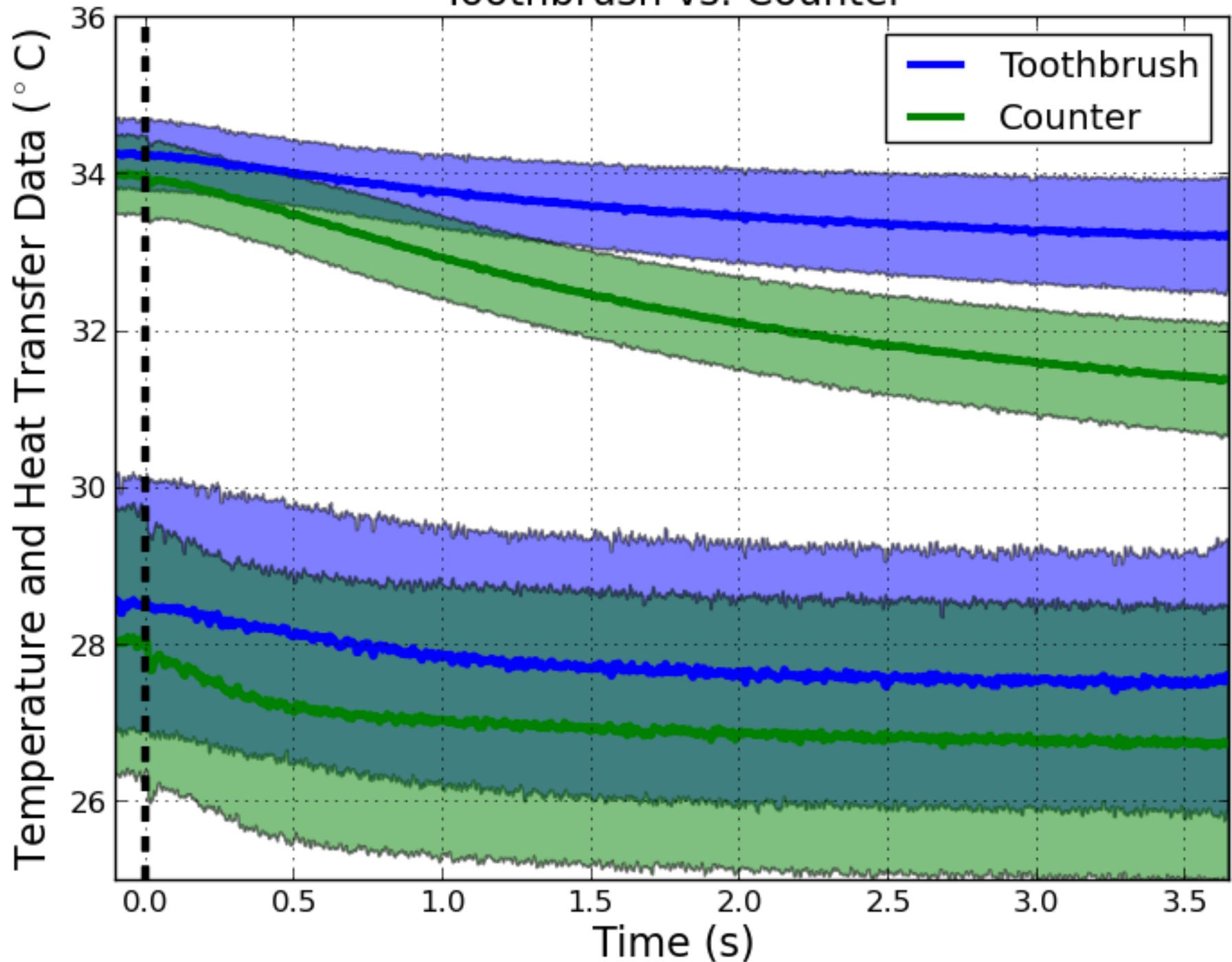
**3. Shoulder**



# Bare Wrist vs. Clothed Shoulder



# Toothbrush vs. Counter



# Passive thermal sensing performed best for humans vs. objects

98.75% with 3.65s of contact



**Vs.**



Counter



Toothbrush



Bathtub



Toilet Handle



Toilet Tank



Toilet Seat



Towel Rack



Towel

[images] found on the internet and used without permission

# Passive & active thermal sensing together performed best for objects vs. objects

92.14% : generalizing to new locations in same environment  
84% : generalizing to different environments  
with 3.65s of contact



[images] found on the internet and used without permission

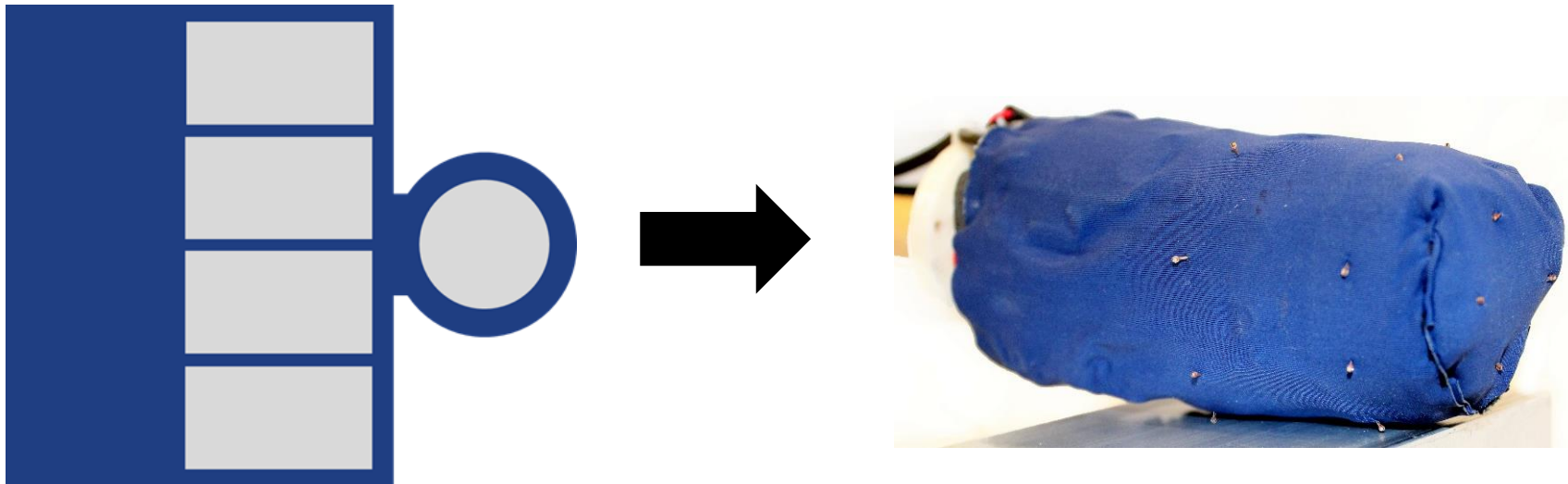
# Force and Thermal Sensing with a Fabric-Based Skin



## A Prototype with 5 multimodal taxels : Force, Active Thermal, and Passive Thermal

Joshua Wade, Tapomayukh Bhattacharjee, and Charles C. Kemp, *Force and Thermal Sensing with a Fabric-based Skin*, IROS Workshop on See, Touch, and Hear : 2nd Workshop on Multimodal Sensor-based Robot Control for HRI and Soft Manipulation, 2016.

# A Prototype with 5 Multimodal Taxels

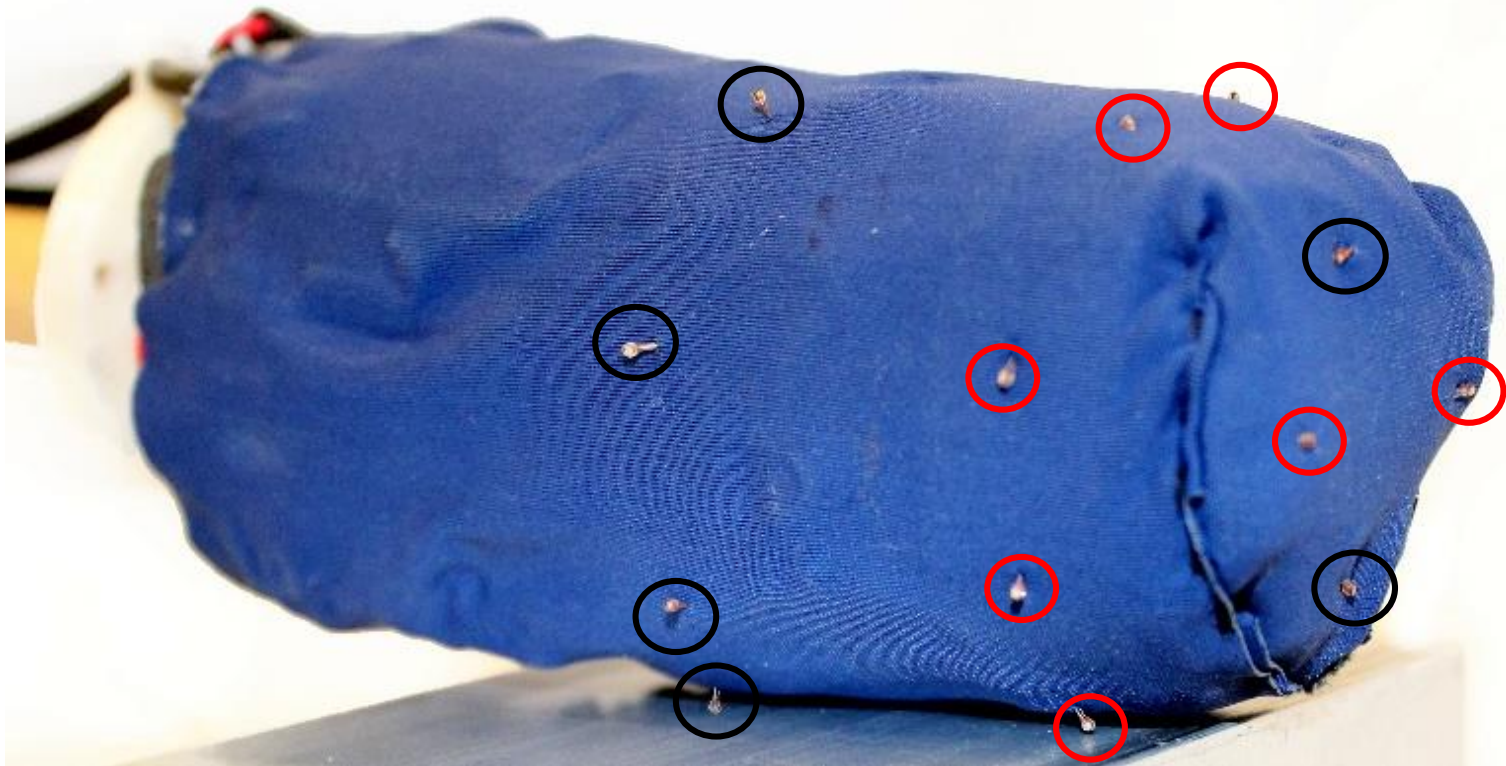


## 5 Force Sensing Taxels

Joshua Wade, Tapomayukh Bhattacharjee, and Charles C. Kemp, *Force and Thermal Sensing with a Fabric-based Skin*, IROS Workshop on See, Touch, and Hear : 2nd Workshop on Multimodal Sensor-based Robot Control for HRI and Soft Manipulation, 2016.



# A Prototype with 5 Multimodal Taxels

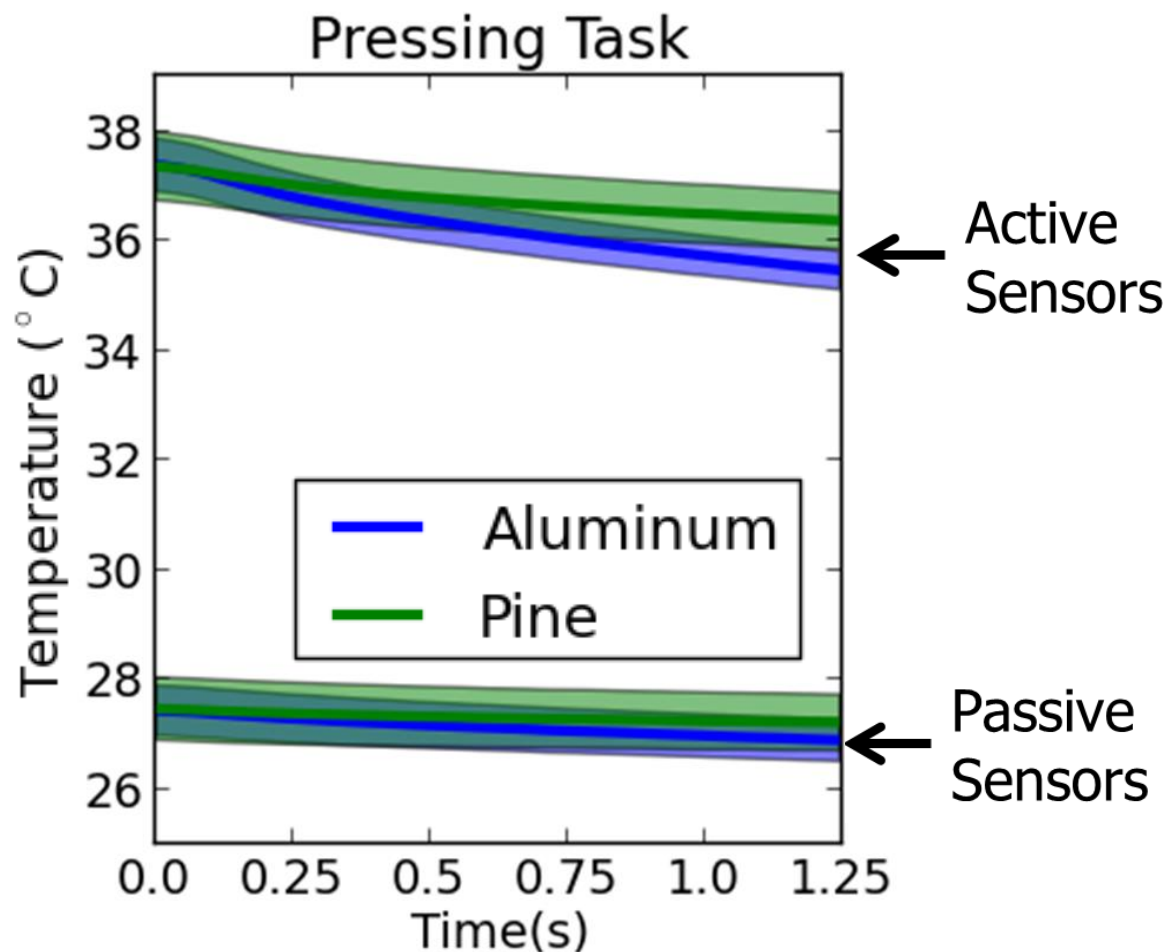
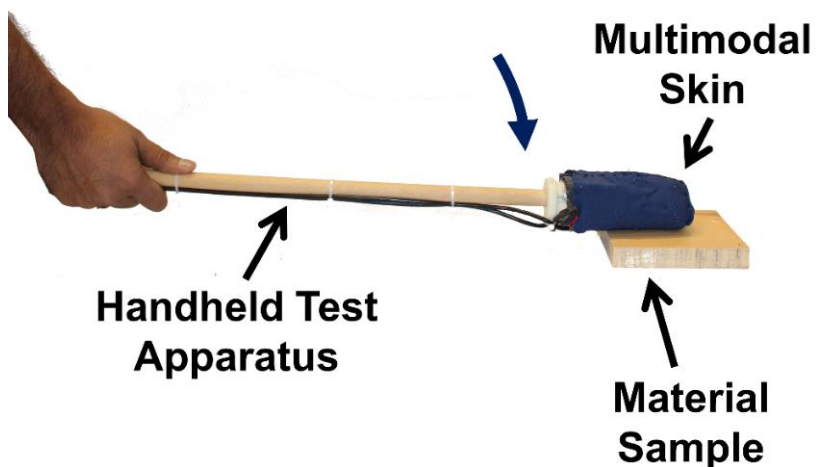


- **10 Actively Heated Thermistors**
- **10 Passive Thermistors**

Joshua Wade, Tapomayukh Bhattacharjee, and Charles C. Kemp, *Force and Thermal Sensing with a Fabric-based Skin*, IROS Workshop on See, Touch, and Hear : 2nd Workshop on Multimodal Sensor-based Robot Control for HRI and Soft Manipulation, 2016.

# Experiments : Representative Manipulation Task 1

## Pressing Task

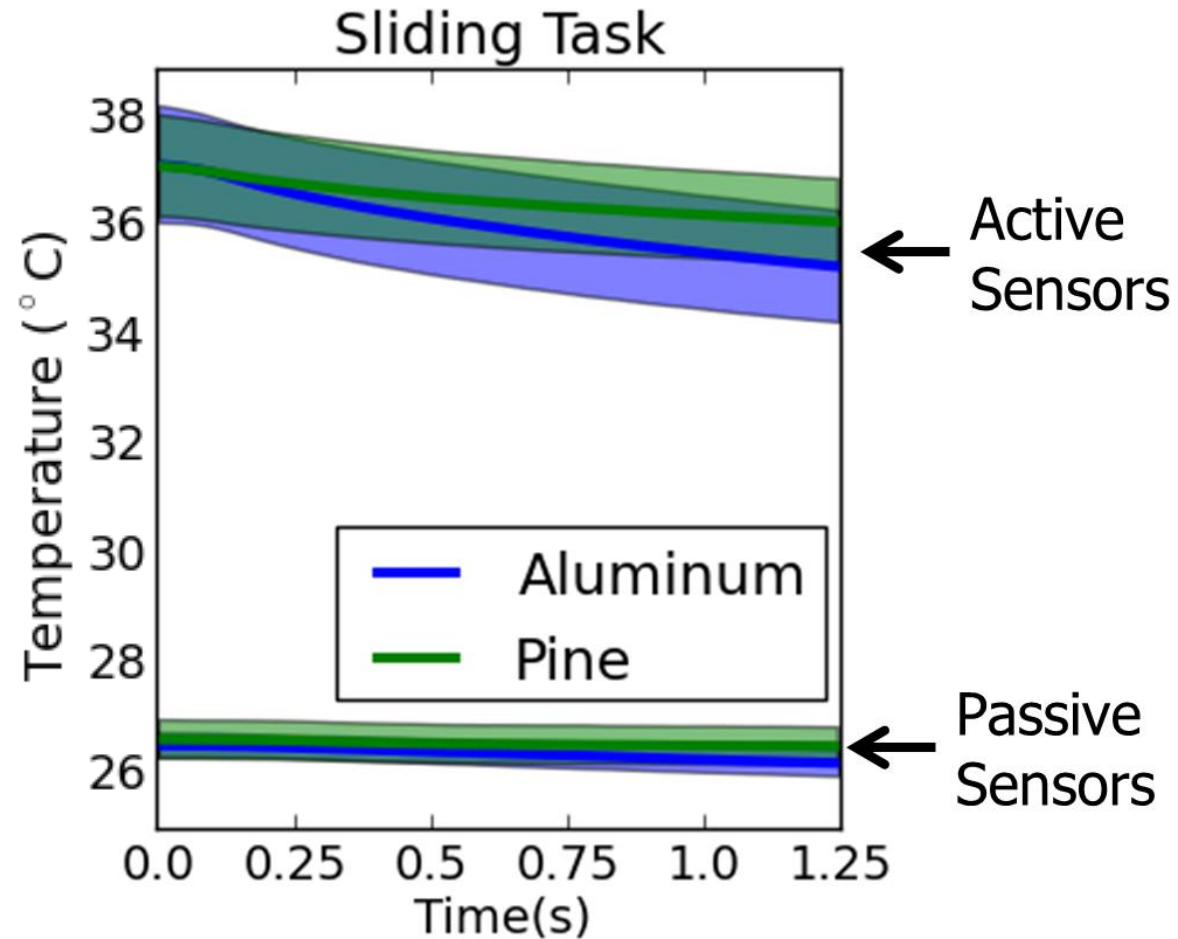
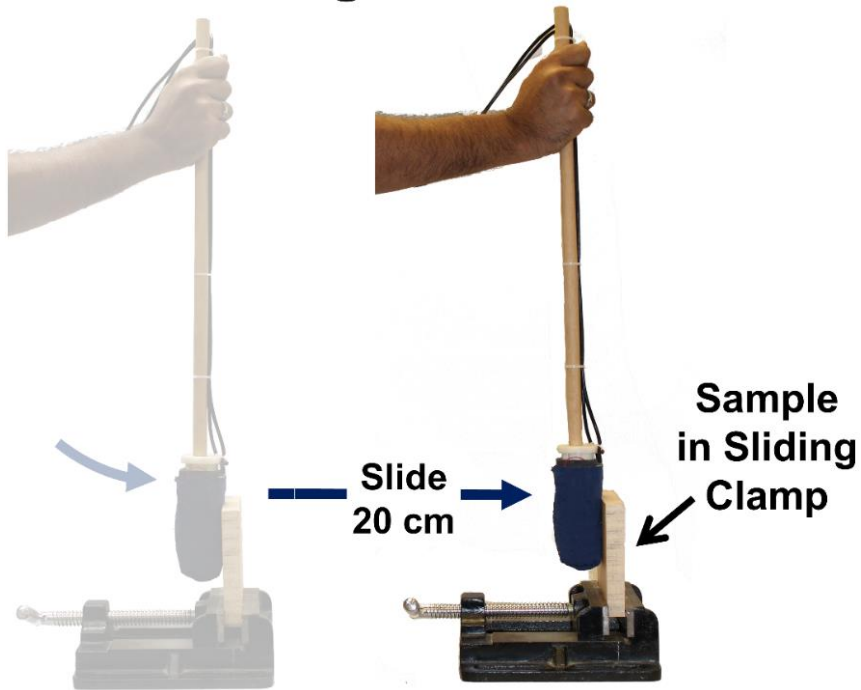


## Distinguish between Aluminum and Pine

Joshua Wade, Tapomayukh Bhattacharjee, and Charles C. Kemp, *Force and Thermal Sensing with a Fabric-based Skin*, IROS Workshop on See, Touch, and Hear : 2nd Workshop on Multimodal Sensor-based Robot Control for HRI and Soft Manipulation, 2016.

# Experiments : Representative Manipulation Task 2

## Sliding Task

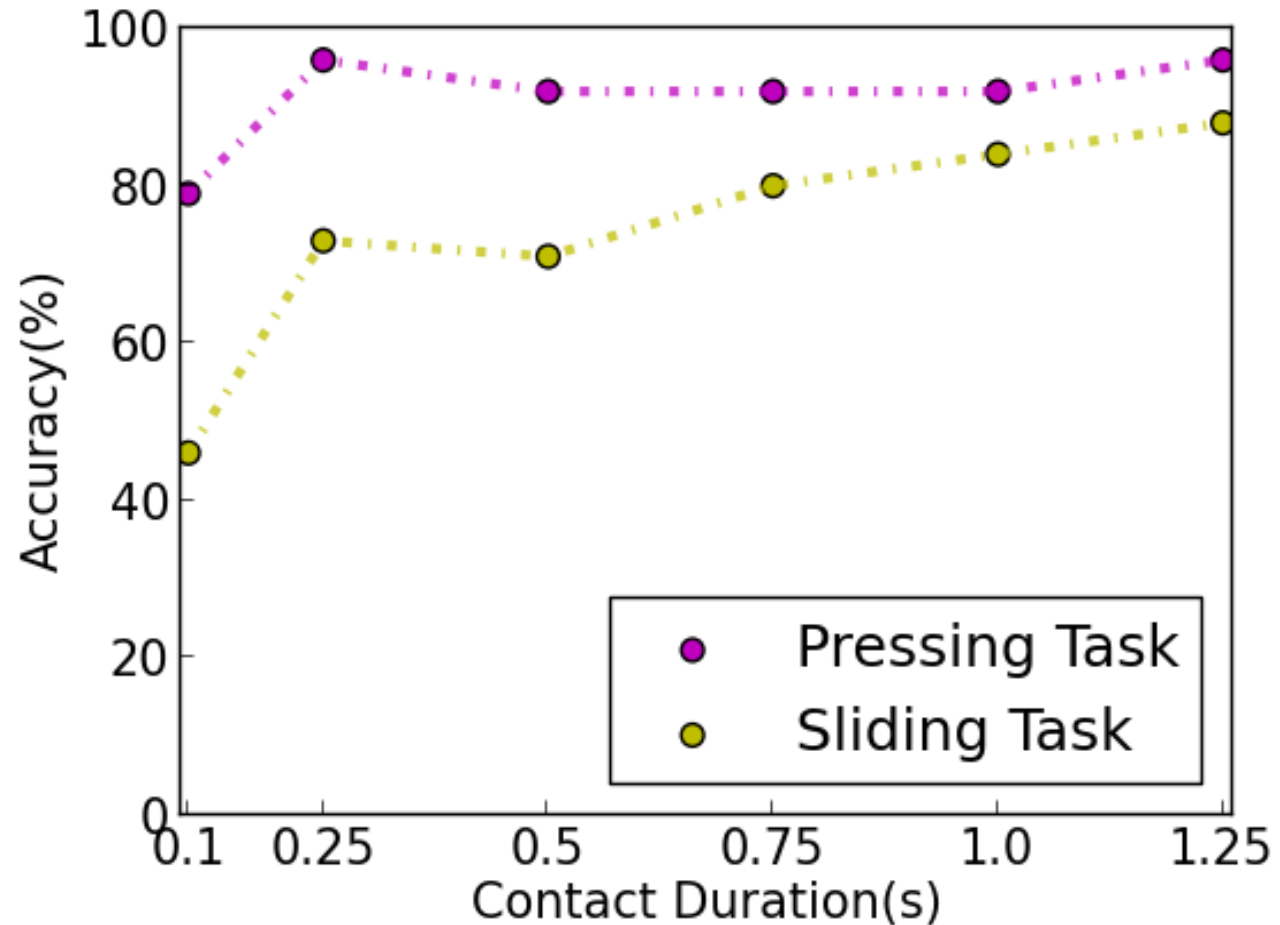


## Distinguish between Aluminum and Pine

Joshua Wade, Tapomayukh Bhattacharjee, and Charles C. Kemp, *Force and Thermal Sensing with a Fabric-based Skin*, IROS Workshop on See, Touch, and Hear : 2nd Workshop on Multimodal Sensor-based Robot Control for HRI and Soft Manipulation, 2016.

# Results : Aluminum vs. Pine

Contact Duration(s)	Recognition Accuracy	
	Pressing Task	Sliding Task
0.25	<b>96%</b>	73%
1.25	<b>96%</b>	<b>88%</b>



Joshua Wade, Tapomayukh Bhattacharjee, and Charles C. Kemp, *Force and Thermal Sensing with a Fabric-based Skin*, IROS Workshop on See, Touch, and Hear : 2nd Workshop on Multimodal Sensor-based Robot Control for HRI and Soft Manipulation, 2016.

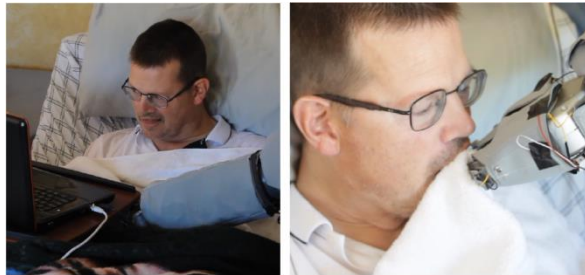
# Thermal Tactile Sensing

- Less sensitive to contact mechanics than force sensing
- Recognize contact with
  - Materials with distinct effusivities
  - Human body
  - Task-relevant objects
- Challenges
  - Time to heat up
  - Time for heat to transfer

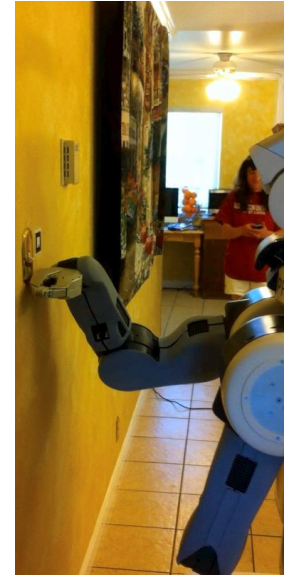
# Assistive mobile manipulation at home is feasible for people with severe motor impairments using conventional interfaces.



General purpose robot from Willow Garage used in this research.



Henry pulls up a blanket and wipes his face for himself while in bed at home using a robot with intelligent tactile sensing.



Henry operates devices in his house for himself with autonomous robot actions.



Henry shaves himself at home using a web-based application for shaving.

Henry Evans is severely impaired due to a brainstem stroke. He operates the robot using a mouse pointer that he controls using motion of his head and his fingers via an off-the-shelf head tracker and mouse buttons.

*(Research was performed as part of the collaborative Robots for Humanity project.)*

# Haptic Sensing for Assistive Robots

Haptic Sensing	Capability	Assistive Tasks
Data-driven models of forces	Common sense about forces	shaving, door opening, dressing, feeding
Whole-arm tactile sensing	Reach locations in clutter	Reach locations around the human body
Thermal tactile sensing	Recognize contact with task-relevant categories	human vs. environment, toothbrush vs. counter, tactile foreground vs. tactile background





 U.S. Department of Health and Human Services



Administration for Community Living



[images] found on the internet and used without permission



# Credit

Many thanks go to the students, postdocs, collaborators, participants, and colleagues who made this work possible.

To learn more, please visit:

<http://healthcare-robotics.com>

**GRADUATING SOON!**



**Tapo Bhattacharjee**



**Phil Grice**