Haptic rendering

Claudio Pacchierotti

CNRS



"Haptic rendering is the process of computing and generating forces in response to user interactions with virtual objects."

–J. Kenneth Salisbury

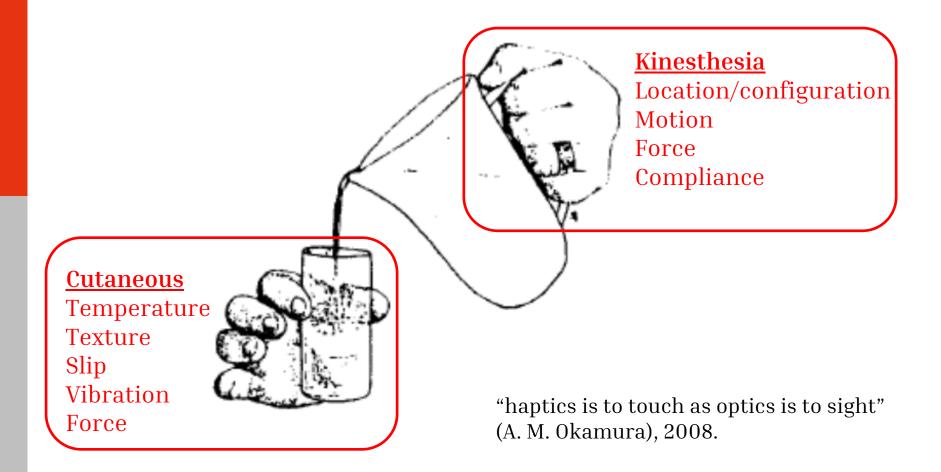


Today's Outline

- · The haptic loop
- God-object/Proxy model
- · Examples of rendering of a wall and a sphere
- · Data-driven rendering
- . Personalization of haptic rendering techniques

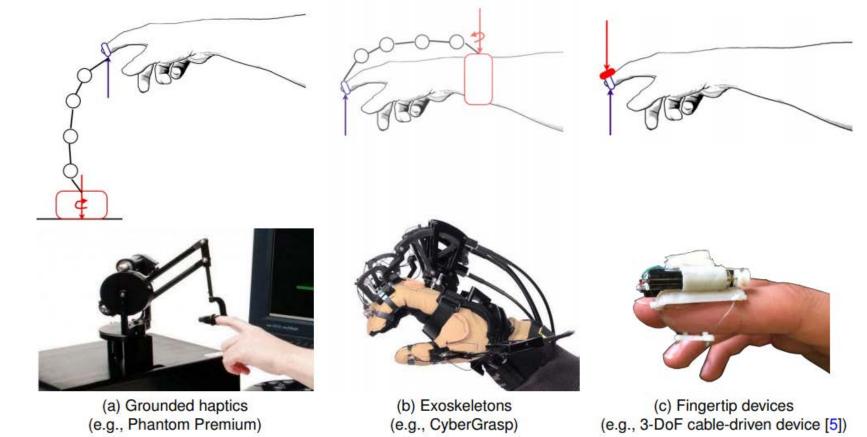


Haptics = Cutaneous and kinesthetic stimuli





From grounded to wearable haptics



C. Pacchierotti, S. Sinclair, M. Solazzi, A. Frisoli, V. Hayward, D. Prattichizzo. "Wearable haptic systems for the fingertip and the hand: taxonomy, review, and perspectives" IEEE Transactions on Haptics, 10(4):580-600, 2017.

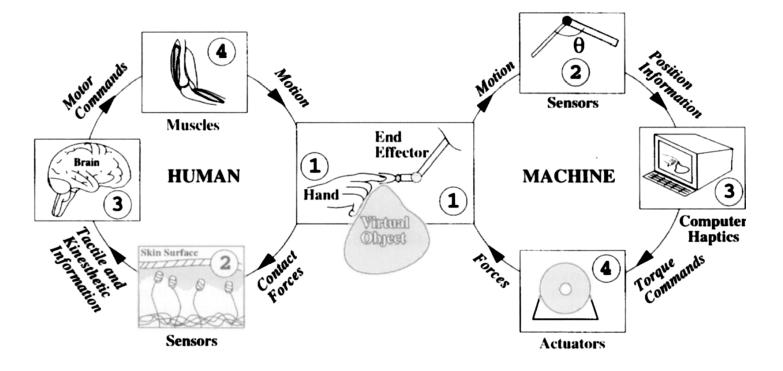


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Information & Power Flows

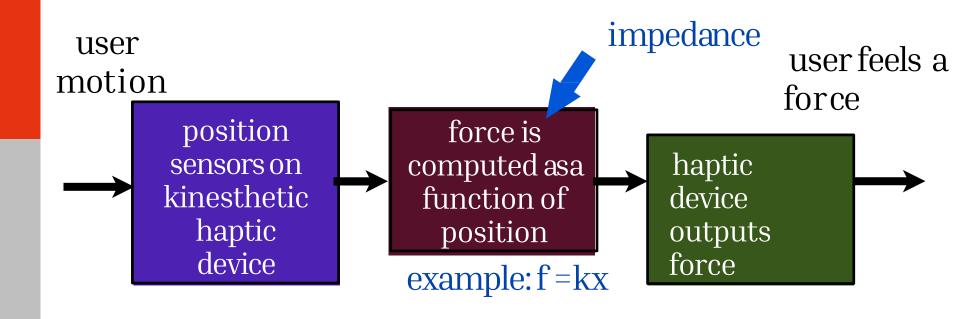


[From M. Srinivasan and C. Basdogan, Computers & Graphics 21(4), 1997.]



Impedance-type kinesthetic devices

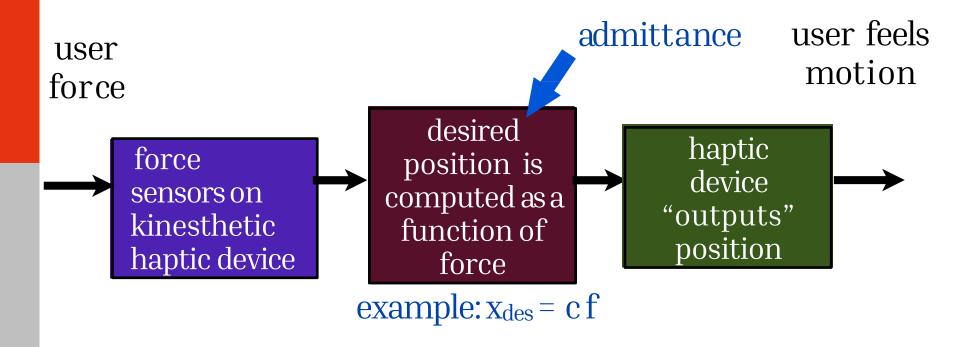
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most force feedback devices are of the "impedance" type

Admittance-type kinesthetic devices

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"admittance"-type devices are not as common

Impedance vs. Admittance Control

 Impedance devices sensed position commanded force



 Admittance devices sensed force commanded position





Impedance vs. Admittance Devices

• Impedance haptic devices cheaper to build, back-drivable

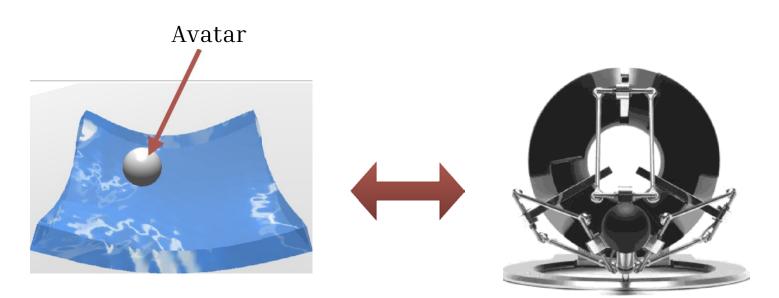


Admittance haptic devices
higher range of forces, requires force
sensor, generally less common



The Basics

How does a basic visual-haptic simulation work?



Virtual Environment

Haptic Interface



The Basics

How does a basic visual-haptic simulation work?



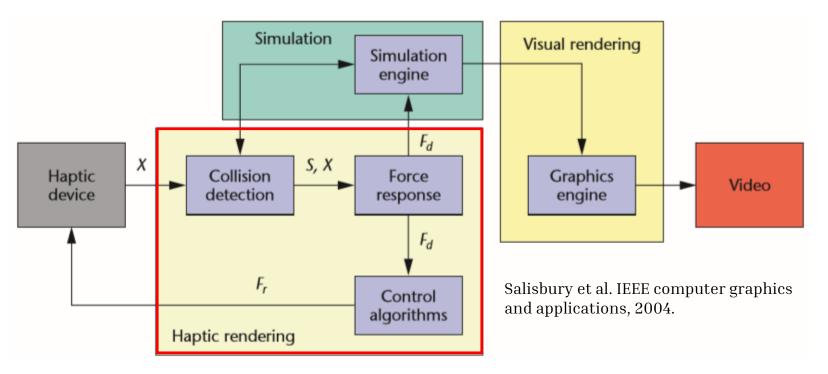


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Components of the haptic simulation



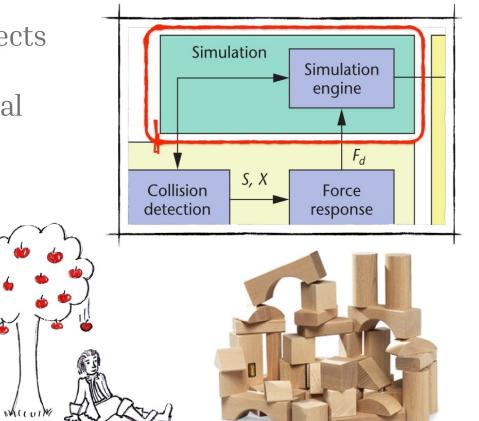
Collision-detection algorithms provide information about contacts S occurring between an avatar at position X and objects in the virtual environment. Forceresponse algorithms return the *ideal* interaction force F_d between avatar and virtual objects. Control algorithms return a force F_r to the user, approximating the ideal interaction force to the best of the device's capabilities.



Components of the haptic simulation

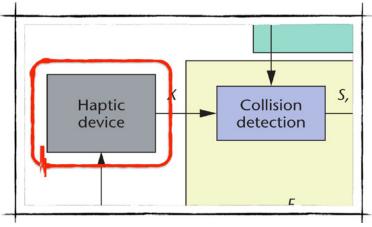
The Virtual Environment

- \cdot representations of virtual objects
- real-time simulation of physical behaviour



Components of the haptic simulation Haptic Device

We discussed them yesterday

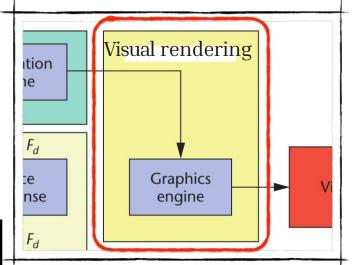






Components of the haptic simulation Visual Rendering

- Given a virtual environment, render its state on the screen (in real time)
- CHAI3D is framework to design haptic interactions in VR









Haptic vs. Visual Rendering







Bi-Directionality

Bi-directional information flow is a distinguishing feature of haptic interfaces.

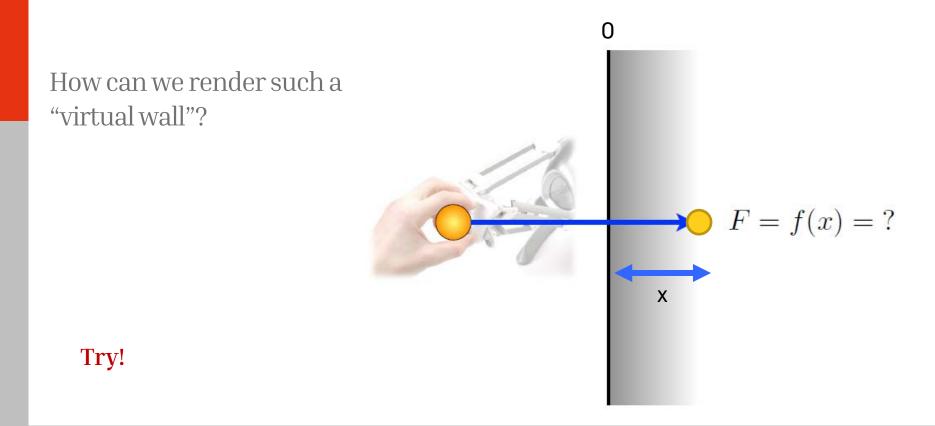


This feature has other consequences that we will not, however, cover in details in this class.

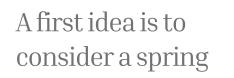


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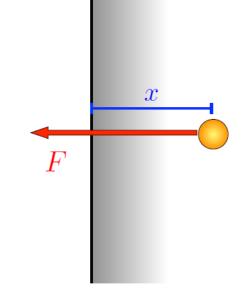
• A plane is one of the simplest virtual environments we can conceive and render



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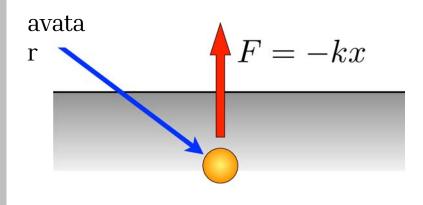


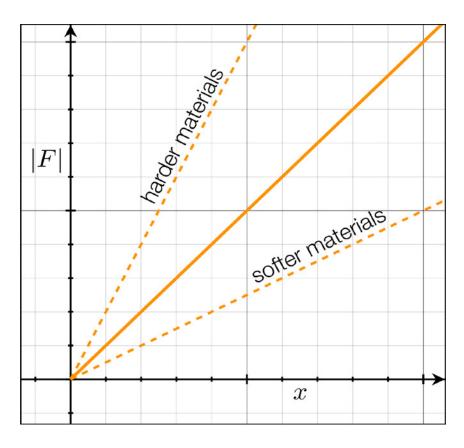
$$F(x) = \begin{cases} -kx & \text{if } x > 0 \\ 0 & \text{otherwise} \end{cases}$$

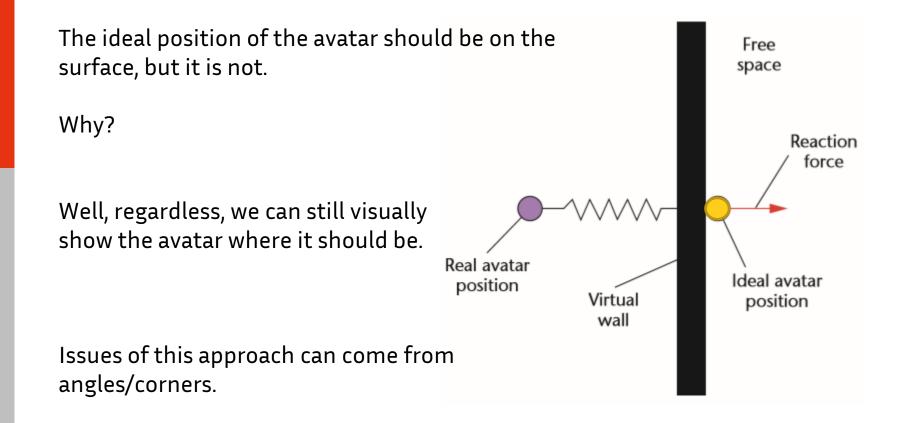


0

The elastic constant (the stiffness k) affects how the virtual wall feels.

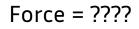


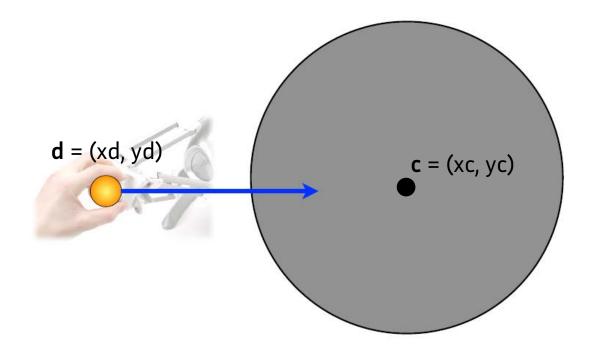




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What about another shape?





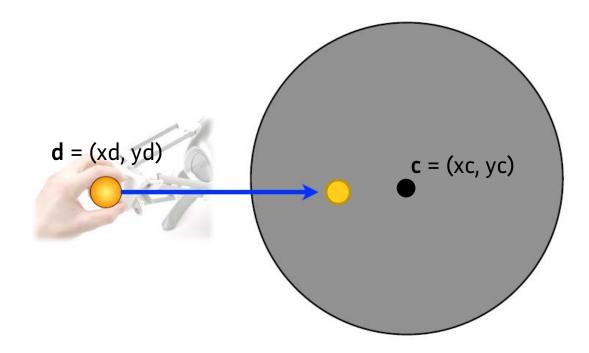
Try!





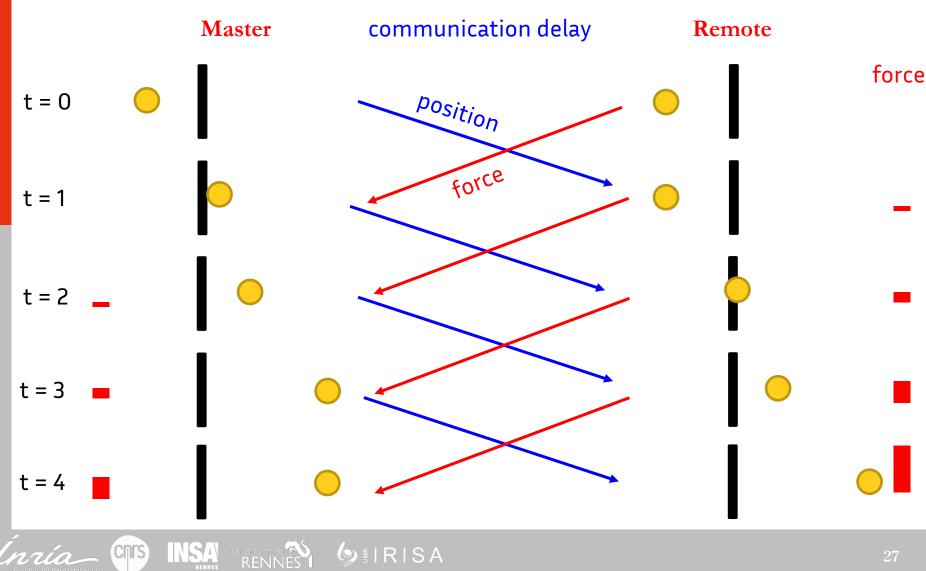
What about another shape?

Force = -k (**d** - **c**)

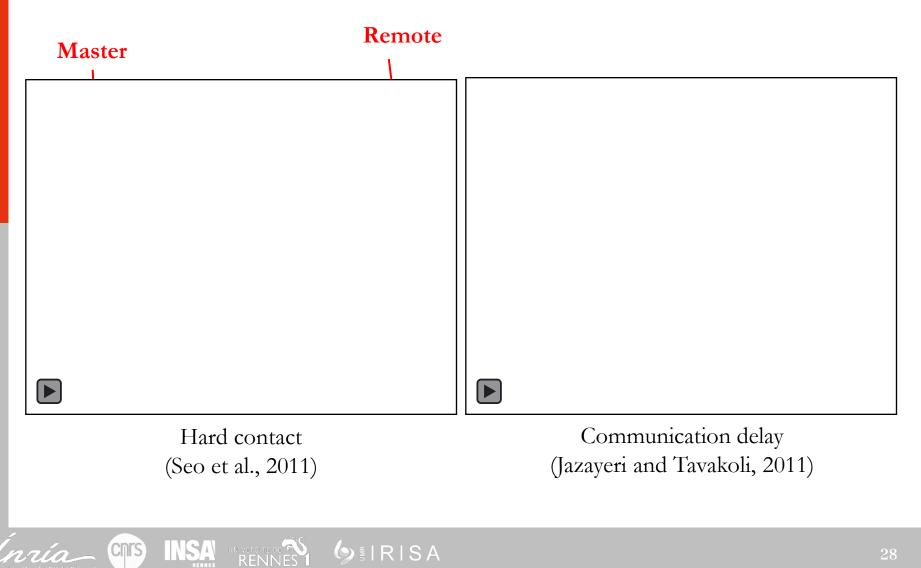




Issues with spring-only interactions



Issues with spring-only interactions

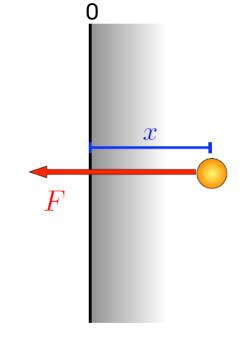


Simple interaction: adding a damper

Considering only a spring might feel too reactive and even introduce severe safety issues in the presence of communication delays.

We can add a damper to the system.

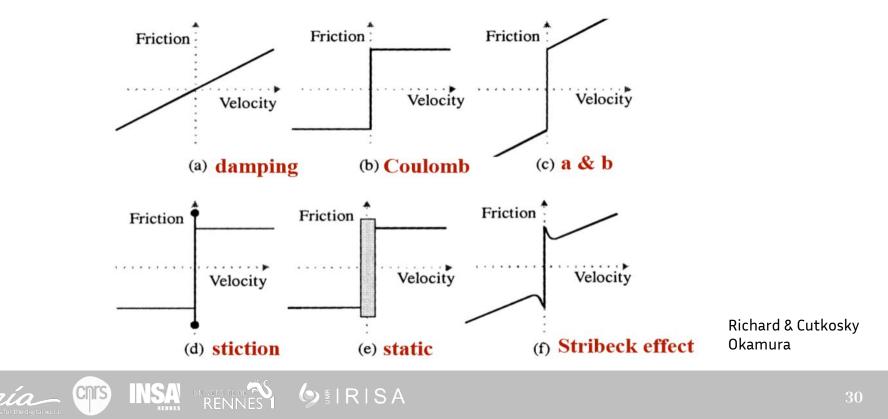
$$F(x) = \begin{cases} -kx - \mathbf{b} \, \dot{\mathbf{x}} & \text{if } x > 0\\ 0 & \text{otherwise} \end{cases}$$



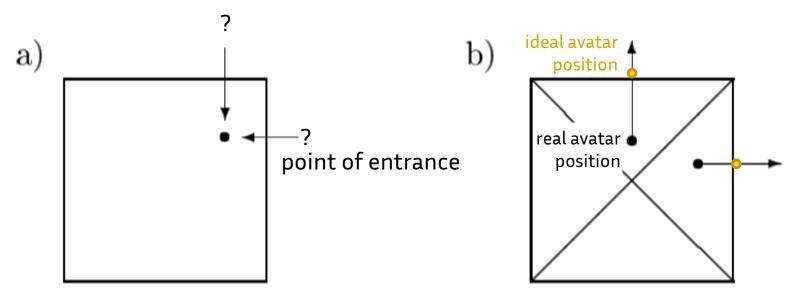
Simple interaction: "frictional" damping

However, doing all this, surfaces can feel unnaturally slippery. Of course, • friction would help, but it can be difficult to implement.

A good idea is to add a damping to motions *parallel* to the surface.

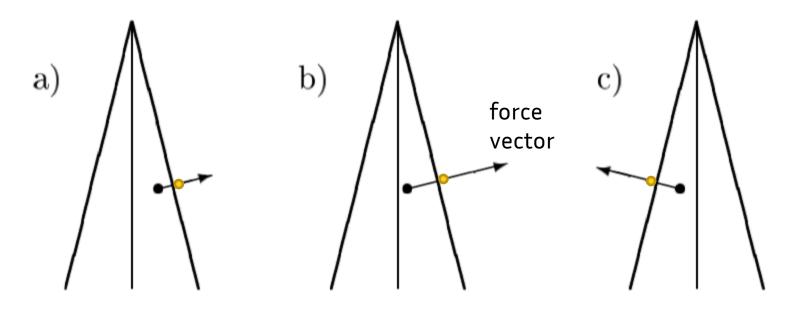


Issues with more complex objects



Two possible paths to reach the same location in a square; without a history we do not know which path was taken by the user. Vector field method: subdivide the square's area and assume that the user entered from the closest edge. The force vectors are normal to the edge and proportional to distance penetrated.

Issues with more complex objects



Push through of thin objects. a) user touches surface and feels a small force, b) as he/she pushes harder he/she penetrates deeper into the object, until c) he/she passes more than halfway through the object where the force vector changes direction and pushes him/her out the other side.

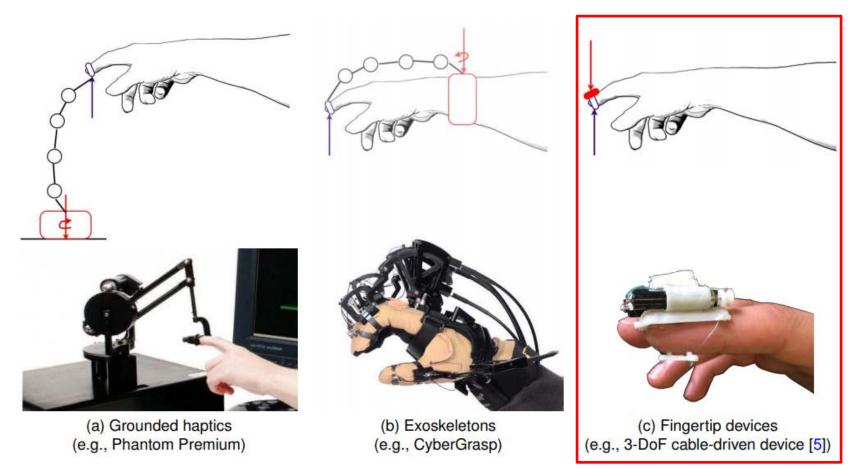


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Cutaneous feedback device





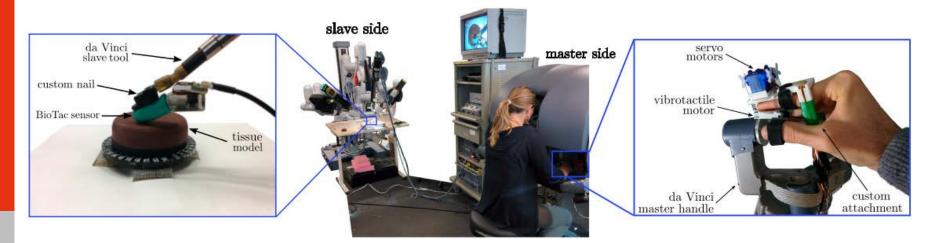


Let's try to imagine an interesting application!

- Interaction in Virtual Reality
- Gaming
- E-commerce
- Medical diagnosis
- Surgical robotics



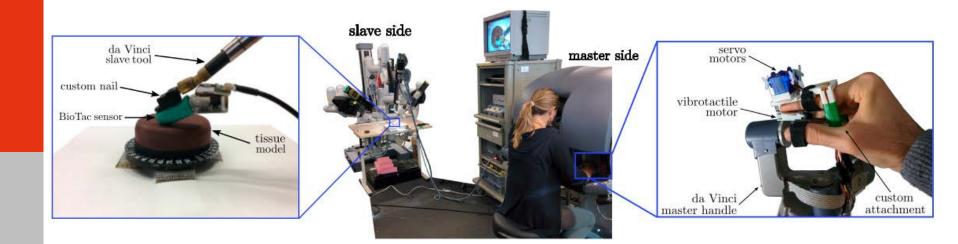
Palpation using the da Vinci Surgical System



The haptic system is composed of a

- (1) BioTac tactile sensor, in charge of registering contact forces and vibrations at the operating table, and
- (2) a cutaneous feedback device, in charge of applying contact forces and vibrations to the surgeon.

Palpation using the da Vinci Surgical System

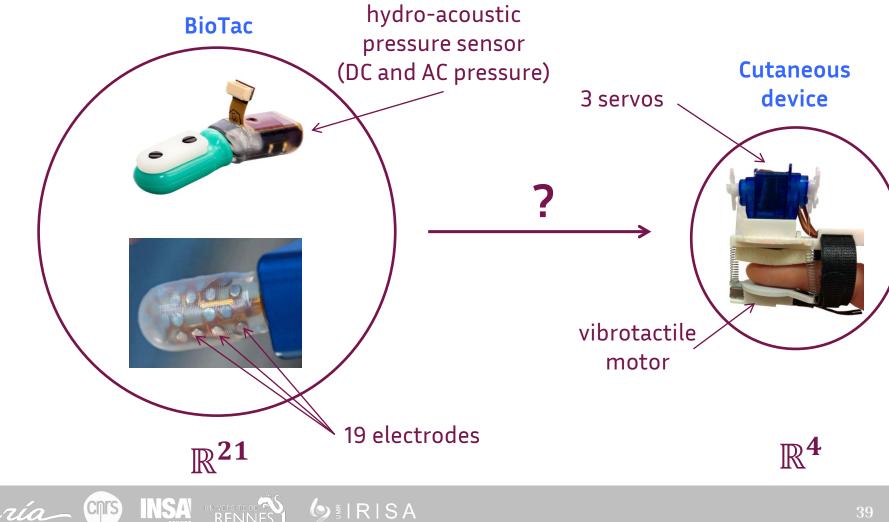


Contact deformations and vibrations sensed by the BioTac are <u>directly mapped to input commands</u> for the cutaneous device's motors using a model-free data-driven algorithm.

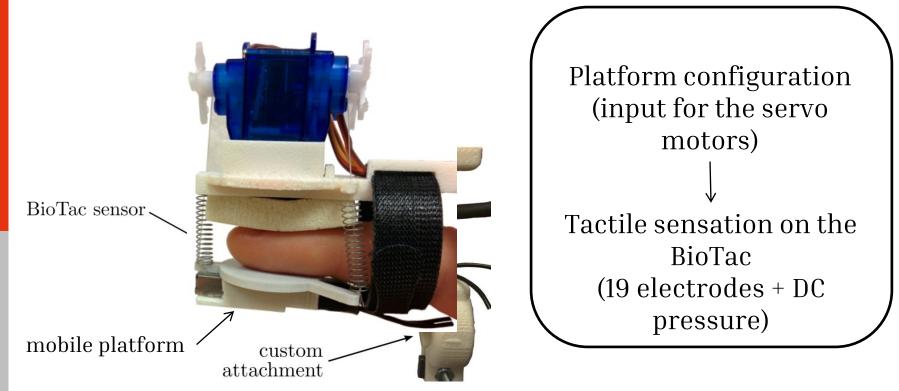
How?



Mapping tactile sensations from the BioTac to the cutaneous device



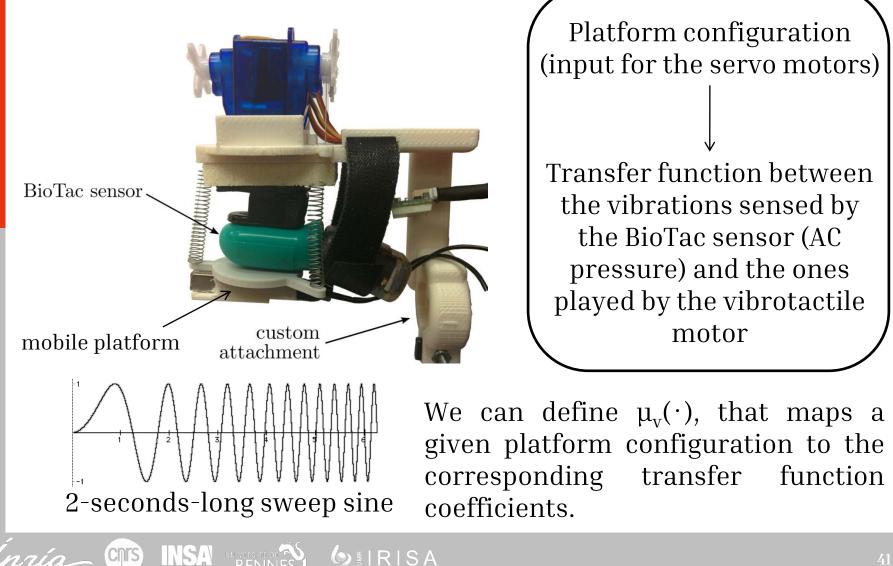
Data collection (servo motors)



We can define $\mu_d(\cdot)$, that maps a given BioTac sensation to the corresponding platform configuration.

S I R I S A

Data collection (vibrotactile motor)



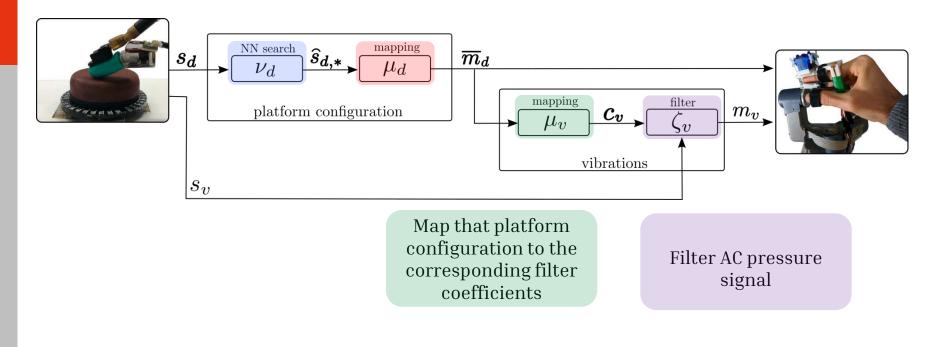
Mapping (simplified)

Search for the closest sensations experienced by the BioTac during the data collection

SA

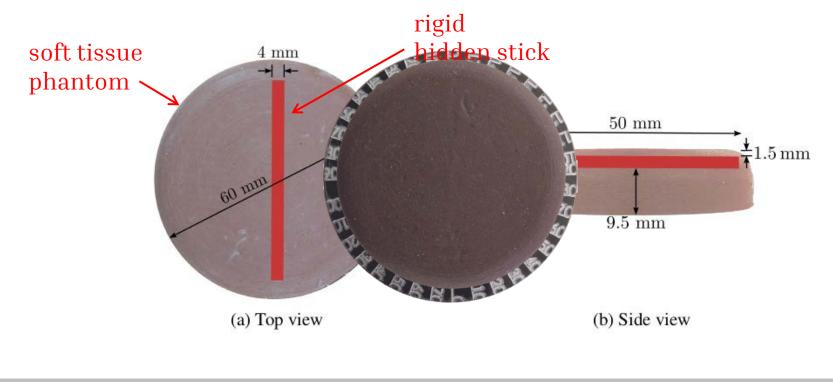
CINICS

Map those sensations to the corresponding platform configuration



Experimental evaluation

The remote environment is composed of a tissue phantom heart model. A plastic stick is embedded into the tissue model at 1.5 mm from the surface, and it is not visible from the outside. The plastic stick simulates the presence of a calcified artery.





Experiment



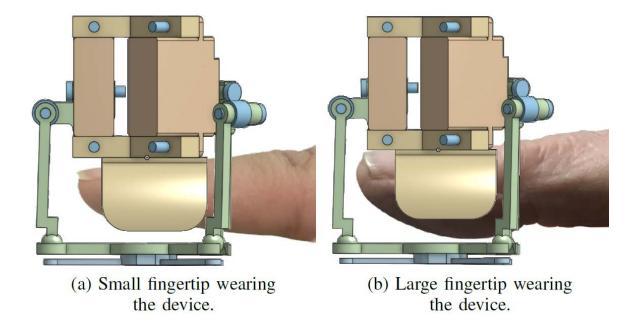




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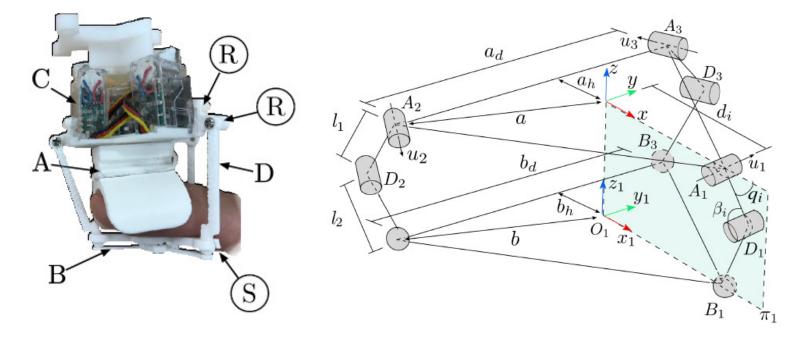


Representative problem: the same fingertip haptic device will elicit different sensations on fingertips having different size and shape.

Personalized haptics optimizes the device design for a target fingertip, so as to to always elicit the desired haptic sensation.

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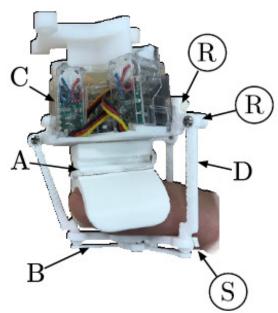
Let's consider the example of this fingertip haptic device.

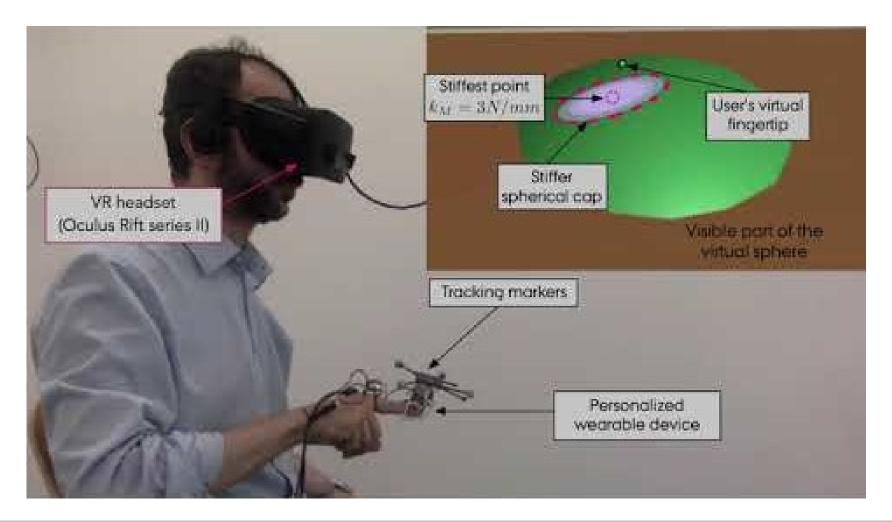


It is composed of a static upper body (A) and a mobile platform (B): the body is located above the nail, supporting three servo motors (C), while the mobile platform contacts the finger pulp. Three legs (D) connect the mobile platform with the static body. **S**IRISA

Malvezzi et al. (2020) came up with a system to personalize the design of this wearable tactile device for a given fingertip, considering three sub-problems in sequence:

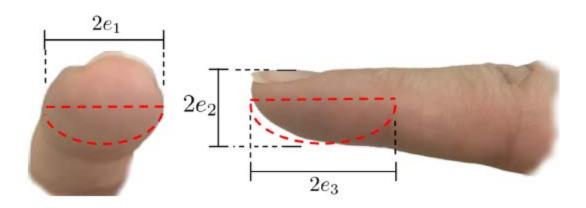
- mobile platform (end-effector) dimensions are defined on the basis of the user's finger dimensions and the device target workspace, i.e., the surface of the finger that will be involved in the cutaneous stimulation.
- static platform dimensions are consequently defined so that, during the cutaneous stimuli application, only the mobile platform interacts with the fingertip and no undesired contacts with the legs occur.
- articulated legs lengths are defined to avoid kinematic singularities in any of the device operative configurations.







Let's try!



My right index finger has

e1 = 9.5 mm; e2 = 8.5 mm; e3 = 18 mm.



Demonstration



We can personalized the behavior of wearable haptic interfaces to fit ones specific characteristics. We can modify the design of the haptic interface or adjust the rendering technique.

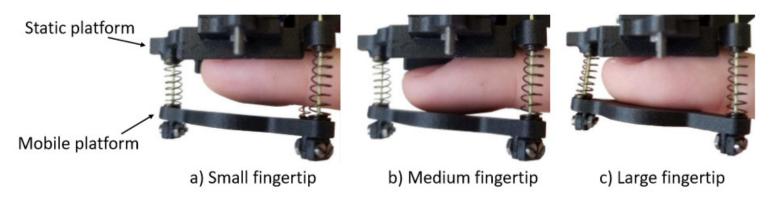
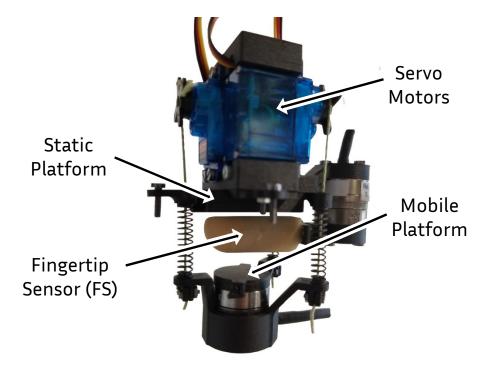


Fig. 1. Problem: the same end-effector configuration elicits different sensations for fingertips of different size, including users with (a) small, (b) medium, and (c) large fingertips. Our personalization methods aim to account for the user's fingertip shape so as to always elicit the desired haptic sensation.



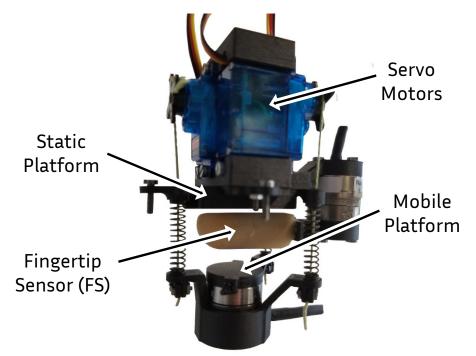
Let's recall the data-driven technique in this context.



First, we place the fingertip sensor (FS) in the device.



Let's recall the data-driven technique in this context.



We then move the end-effector of the device to a wide range of configurations, and the effect of each of these configurations is registered on the FS.



Let's recall the data-driven technique in this context.



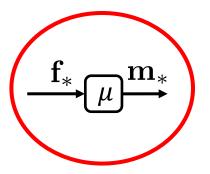


Let's recall the data-driven technique in this context.

With this data, we define the mapping

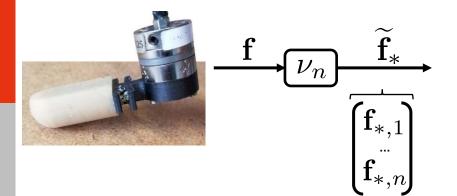
 $\mu(\mathbf{f}_*) = \mathbf{m}_{*},$

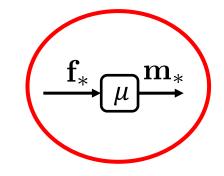
where \mathbf{m}_* is a vector of motor inputs and \mathbf{f}_* is the corresponding FS output.





Let's recall the data-driven technique in this context.

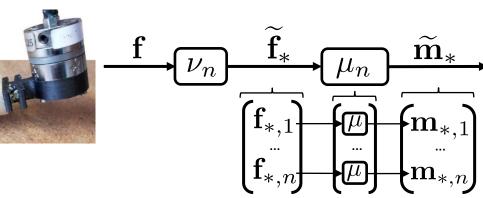


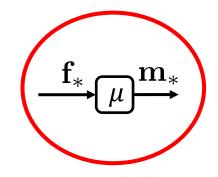


To render a new remote sensation, \mathbf{f} ... we first find the *n* most similar sensor outputs measured during data collection, \mathbf{f}_* .



Let's recall the data-driven technique in this context.

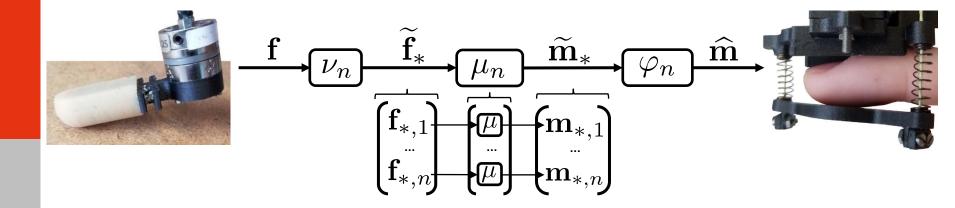




We then convert each element in $\tilde{\mathbf{f}}_*$ back to its corresponding motor inputs, using the lookup table.

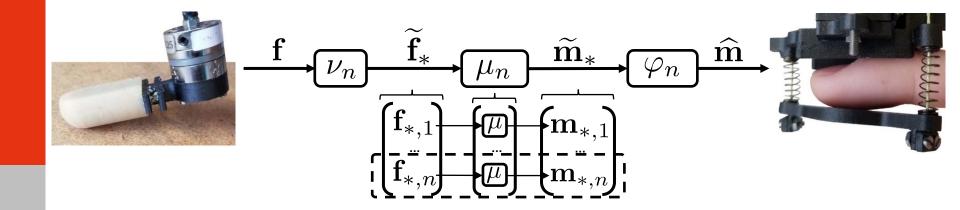


Let's recall the data-driven technique in this context.

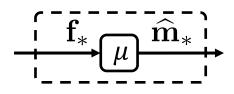


Finally, the *n* motor inputs, $\widetilde{\mathbf{m}}_*$, are averaged into one motor input, $\widehat{\mathbf{m}}_*$, weighting each input according to the distance between the original sensation, \mathbf{f} , and the one elicited by $\mathbf{m}_{*,i}$ during data collection, $\mathbf{f}_{*,i}$.

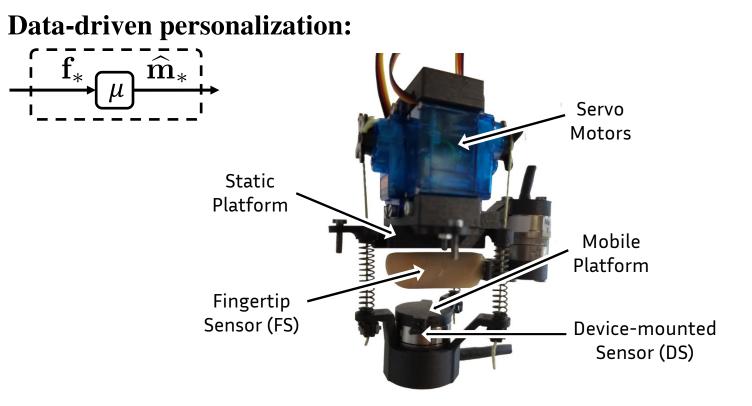




Both of our personalization methods adjust the vector of motor inputs, \mathbf{m}_* , associated with a sensation, \mathbf{f}_* , in this FS-centered calibration.







First, we adjust our FS-centered calibration procedure to include data from a device-mounted sensor (DS).

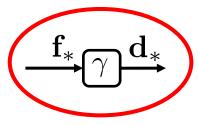


Data-driven personalization: $f_* \mu \hat{m}_*$

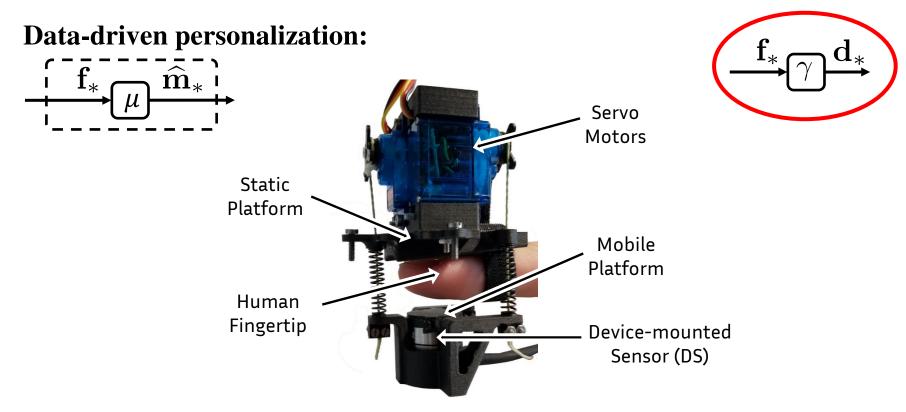
With this data, we define an additional mapping

 $\gamma(\mathbf{f}_*) = \mathbf{d}_{*},$

where \mathbf{f}_* is an FS output and \mathbf{d}_* is the corresponding DS output.

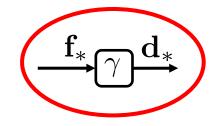






Then we have the human wear the device, and we create a user-centered lookup table.

Data-driven personalization:



With this data, we define the mapping

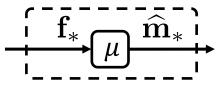
$$\alpha_U(\mathbf{d}_{*,U}) = \mathbf{m}_{*,U},$$

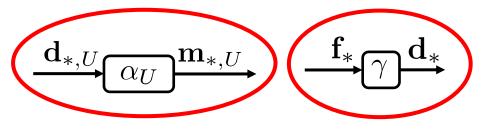
where $\mathbf{m}_{*,U}$ is a vector of motor inputs and $\mathbf{d}_{*,U}$ is the corresponding DS output, for a particular user, *U*.

$$\underbrace{\mathbf{d}_{*,U}}_{\alpha_U} \mathbf{m}_{*,U}$$



Data-driven personalization:

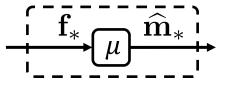


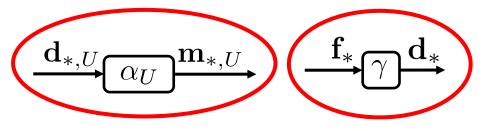


We can now use these functions to personalize the FS-centered calibration data.



Data-driven personalization:



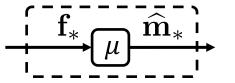


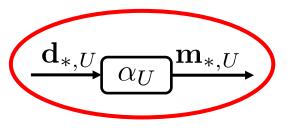


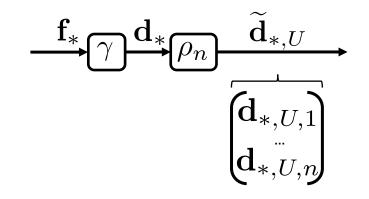
First, we take every FS sensation registered during the FS-centered calibration, \mathbf{f}_* , and find the corresponding DS sensation, \mathbf{d}_* .



Data-driven personalization:



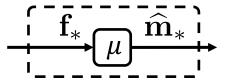


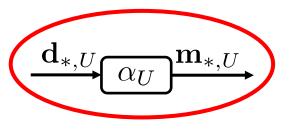


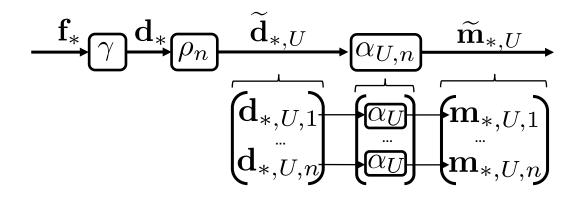
Then, we find the *n* nearest DS sensations observed during the user-centered calibration, $\tilde{d}_{*,U}$.



Data-driven personalization:



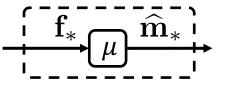


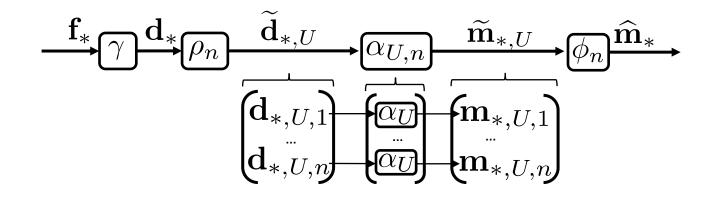


We then convert each element in $\widetilde{\mathbf{d}}_{*,U}$ back to its corresponding motor inputs, $\widetilde{\mathbf{m}}_{*,U}$, using the lookup table.



Data-driven personalization:

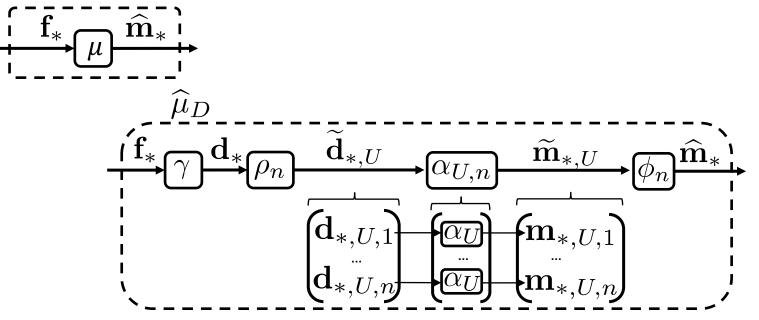




Finally, we find the weighted average of these motor inputs, $\widehat{\mathbf{m}}_{*}$.



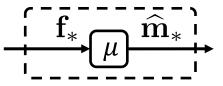
Data-driven personalization:



This procedure gives us the complete adjustment for our data-driven personalization approach.



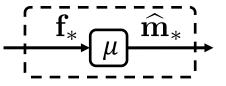
Data-driven personalization:



After calibrating our device for both the FS and the user, we can remove the DS and replace our end-effector with the original flat surface.



Geometric personalization:



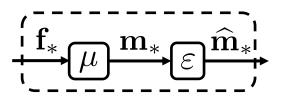
Now let's look at our geometric personalization approach.



Geometric personalization:

 $\xrightarrow{\mathbf{f}_{*}} \mu \xrightarrow{\widehat{\mathbf{m}}_{*}}$

We define our geometric adjustment by



where ε adjusts each motor input registered during the FS-centered data collection based on the geometry of the target finger.



Geometric personalization:

 \mathbf{m}_{*}

 $\begin{array}{c} S_0 \quad \mathbf{z}_0 \\ \mathbf{y}_0 \\ \mathbf{p}_F \quad S_1 \quad \mathbf{z}_1 \\ \mathbf{p}_c \quad \mathbf{y}_1 \\ \mathbf{$

For each motor command \mathbf{m}_* in the FS-centered calibration, we first find the corresponding mobile platform position and orientation, given by \mathbf{p}_c and S_1 , respectively.



Geometric personalization:

 \mathbf{m}_{*}

We then compute the point \mathbf{p}_o at which the undeformed surface of the FS is parallel to our mobile platform, given by

$$\mathbf{p}_o = \mathbf{p}_F - r_F \mathbf{z}_1$$



Geometric personalization:

 \mathbf{m}_{*}

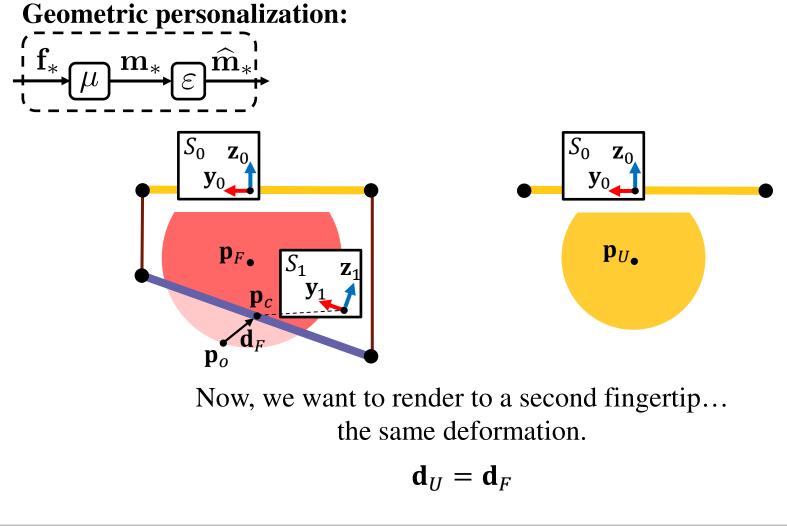
 \mathbf{m}_{*}

 $\begin{array}{c} S_0 \quad \mathbf{z}_0 \\ \mathbf{y}_0 \\ \mathbf{p}_F \quad S_1 \quad \mathbf{z}_1 \\ \mathbf{p}_c \quad \mathbf{y}_1 \\ \mathbf{p}_o \\ \mathbf{p}_o \end{array}$

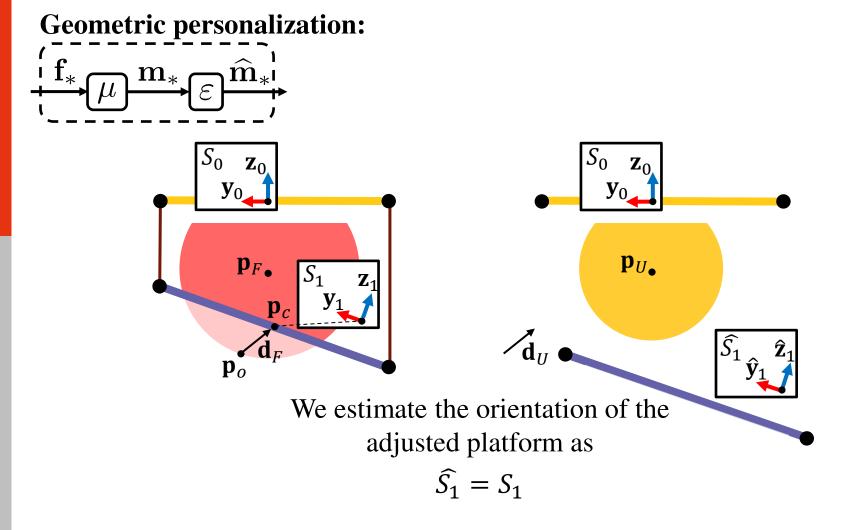
We are then able to estimate how much and in which direction the mobile platform deforms the sensor

$$\mathbf{d}_F = \mathbf{p}_c - \mathbf{p}_o$$

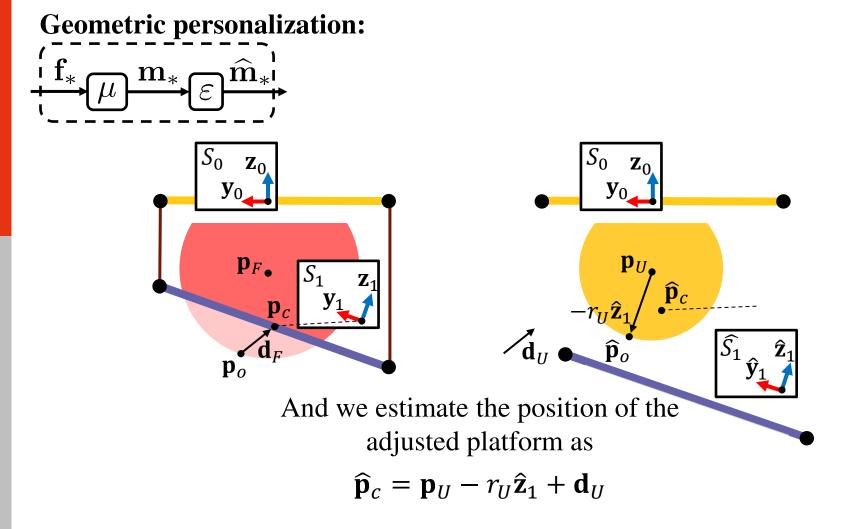




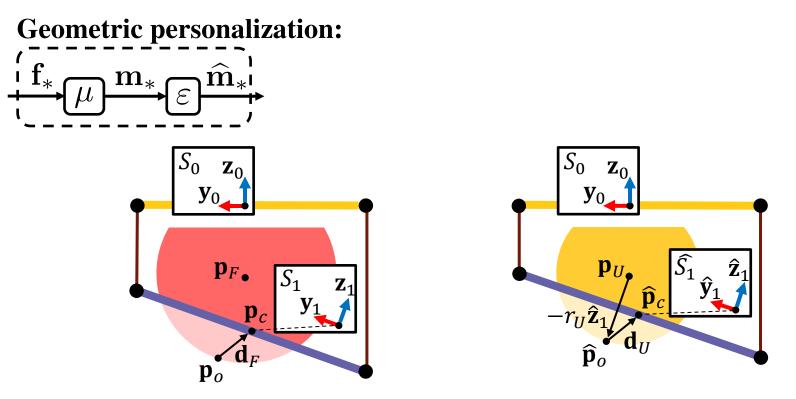












Finally, we convert our adjusted mobile platform orientation and position into adjusted motor inputs \widehat{m}_* .

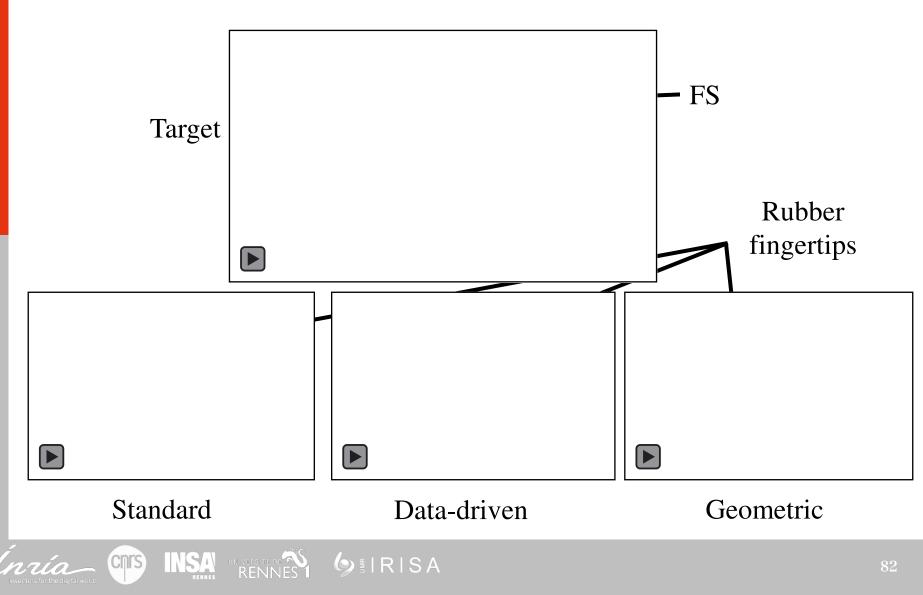


Evaluation example

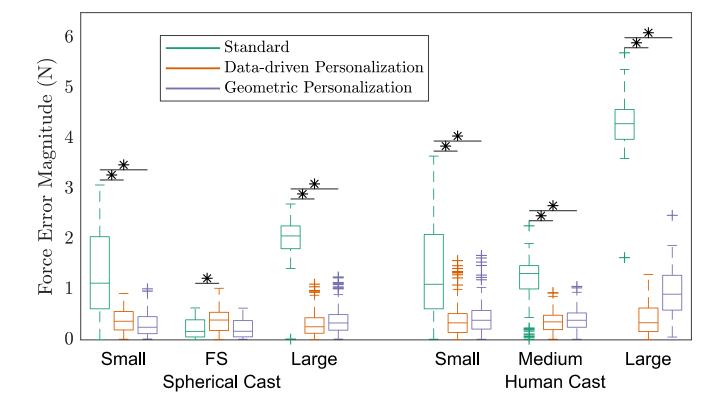
Render of a set of predefined tactile interactions to rubber or human fingertip using either

the **standard**, **non-personalized** method, the **data-driven personalization** method, or the **geometric personalization** method.



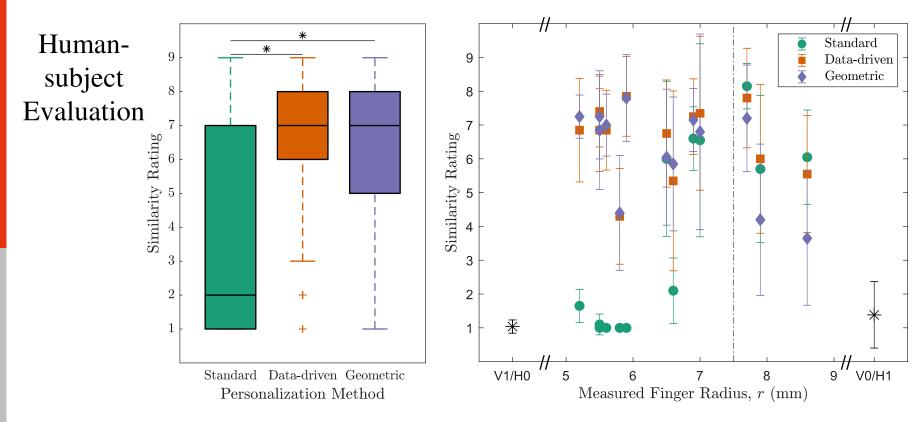


Objective Evaluation



The data-driven and geometric personalizations significantly reduced the force rendering error compared to the standard approach for all of the rubber fingertips other than the FS.

SIRISA



We found that similarity ratings were significantly higher for both the data-driven and geometric personalizations than for the standard approach.



Thank you!



Questions?

