Manipulation systems

Gionata Salvietti

ROBOTIC MANIPULATION

Robotic Manipulation



1997- IBM supercomputer, Deep Blue became Chess World Champion

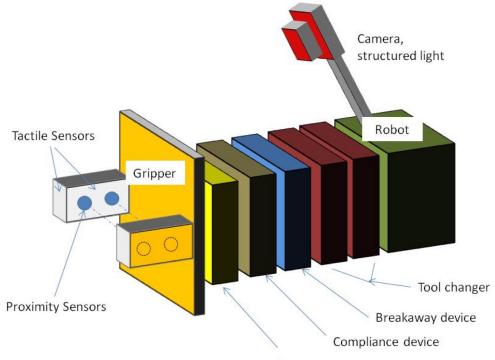
Rolling Justin @ DLR

https://youtu.be/SLqXZ3aFVQA



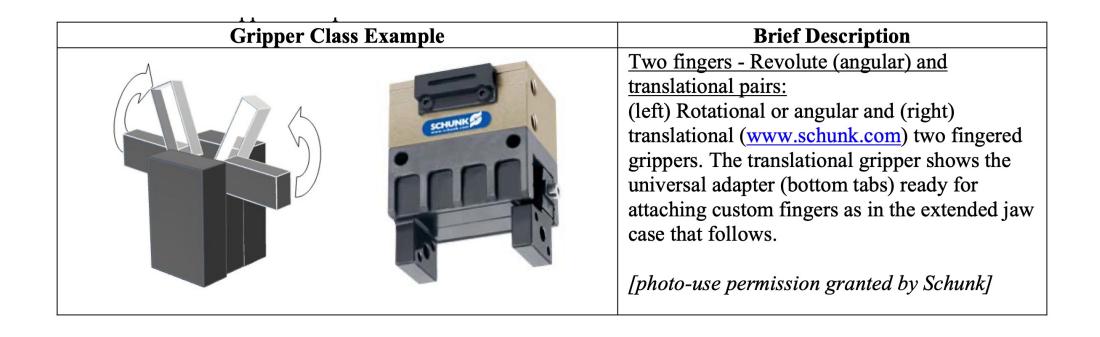
Manipulation technology

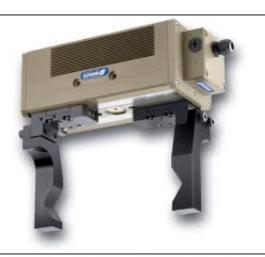
Robot end-effector



Force/Torque sensor

Bostelman, R. and Falco, J. (2012), Survey of Industrial Manipulation Technologies for Autonomous Assembly Applications, NIST Interagency/Internal Report (NISTIR), National Institute of Standards and Technology, Gaithersburg, MD, [online], https://doi.org/10.6028/NIST.IR.7844





Two Fingers - Extended jaws:

Translational two fingered gripper with extended jaws for grasping square or round parts. (www.schunk.com)

[photo-use permission granted by Schunk]



<u>Three fingers - Grasp at three points:</u>

Three fingered concentric, long stroke, belt-drive gripper (www.schunk.com)

[photo-use permission granted by Schunk]

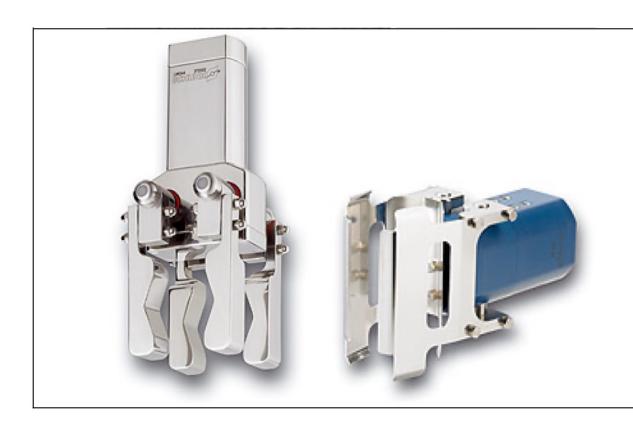




Spherical objects of differing size maintaining center:

Three fingered gripper (<u>www.robotiq.com</u>)

[photo-use permission granted by Robotiq]



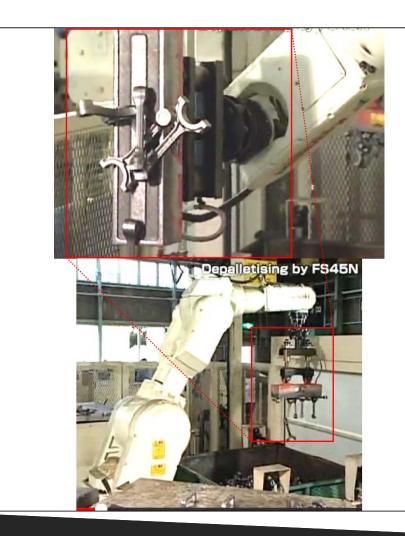
Special purpose grippers:

(top left) angular gripper with hygienic design and (top right) stacking gripper, both for the food processing industry (www.schunk.com); (bottom left) bagged product gripper and (bottom right) universal warehousing gripper that can handle bags, bundles, cases, etc. (www.fanucrobotics.com)

[photo-use permission granted by Schunk and Fanuc Robotics]



Special purpose end-effectors



Electromagnetic gripper:

Magnetic gripper handling piston rods (Kawasaki Robotics -

http://www.youtube.com/watch?v=eGPne8_sR4c)

Multipurpose end-effectors



DLR-HIT Hand II



Universal Gripper

Gripper vs Hands

- Structured environments
- Reliable
- Simple
- Low cost







- Unstructured environments
- Adaptable
- Complex
- Expensive







Gripper vs Hands

N. of DoFs











N. of apps in industries

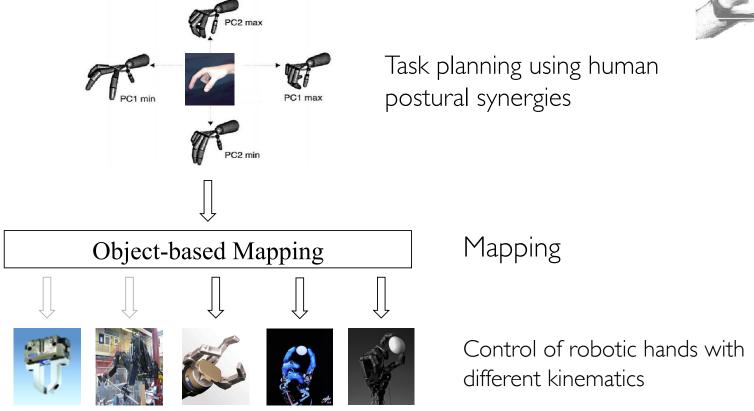
Why articulated hands are not so frequently used in industrial environment?

- Too complex from a control point of view: reaching, grasping, manipulation
- They suffer from a lack of standard <u>control approaches</u> that uses few parameters in spite of the mechanical complexity
- Simplify a complex control problem and analogies with industrial PID

SIMPLIFICATION BY CONTROL

A simplified control framework



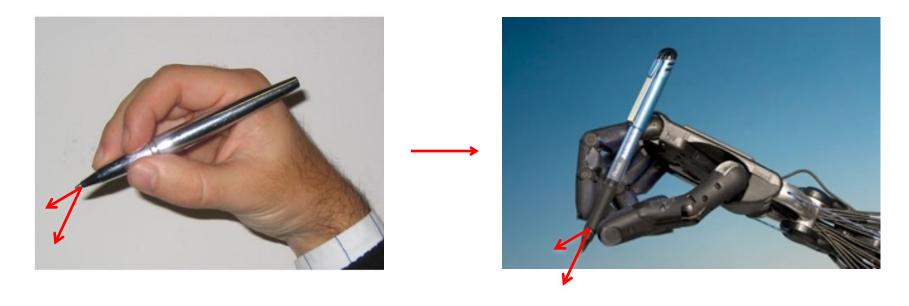


Use synergies to decrease the complexity of control algorithm for robotic hands

The basic idea is to plan the task with a paradigmatic hand and map the obtained movements onto the robotic hands

Object-based mapping for forces and motions

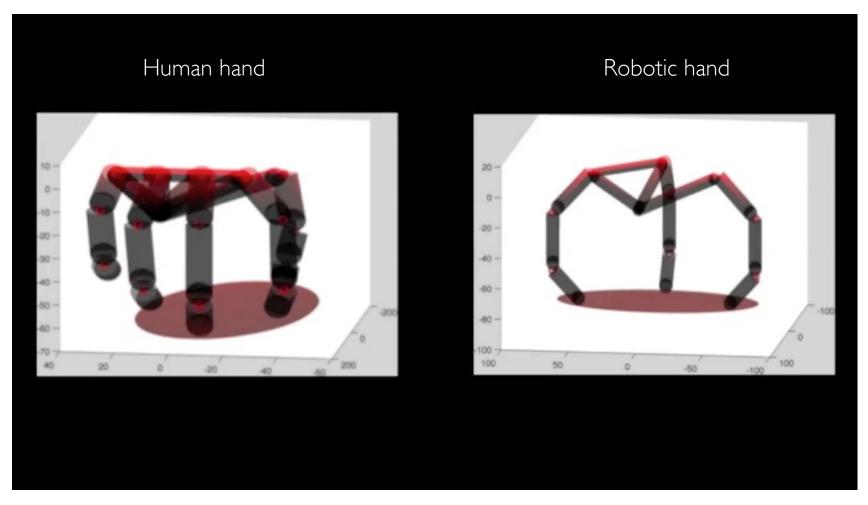
Human hand Robotic hand



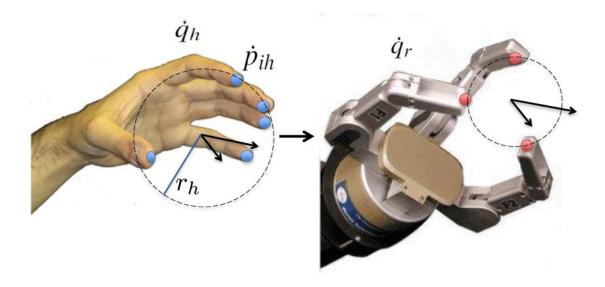
Object-based

- compute the object motion associated to a motion on the human hand
- assign that motion to the same object but with the robotic hand
- solving an inversion problem to compute the slave motion

Object-based abstraction for simplification



A virtual sphere



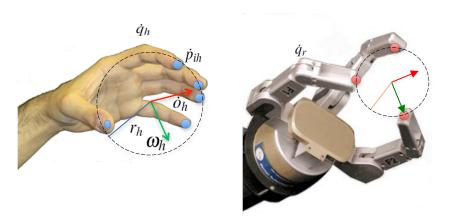
A set of reference points are chosen on the hand model in a given starting configuration

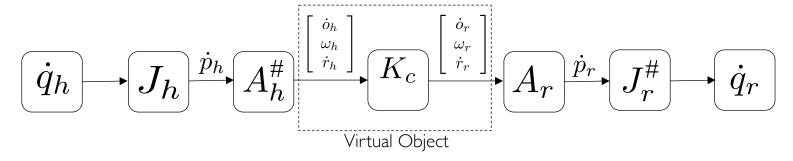
The virtual sphere is the minimum volume sphere containing the reference points

Rigid-body motion and radial deformation for the virtual sphere

Virtual sphere mapping algorithm

Reference points on the fingertips





 J_h, J_r Jacobian matrix

 K_c scaling matrix

Reference point velocity

$$\dot{p}_{ih} = \dot{o}_h + \omega_h \times (p_{ih} - o_h) + \dot{r}_h (p_{ih} - o_h)$$

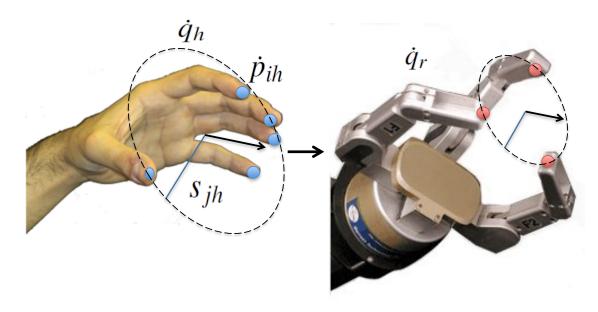
Extended Grasp matrix

$$A_h = \begin{bmatrix} I & -[p_{1h} - o_h]_{\times} & (p_{1h} - o_h) \\ \dots & \dots & \dots \\ I & -[p_{ih} - o_h]_{\times} & (p_{ih} - o_h) \end{bmatrix}$$

More information: a virtual ellipsoid

Virtual ellipsoid mapping

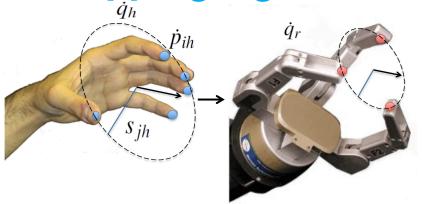
G. Gioioso, G. Salvietti, M. Malvezzi, D. Prattichizzo. An Object-Based Approach to Map Human Hand Synergies onto Robotic Hands with Dissimilar Kinematics. In Robotics: Science and Systems VIII, 2013

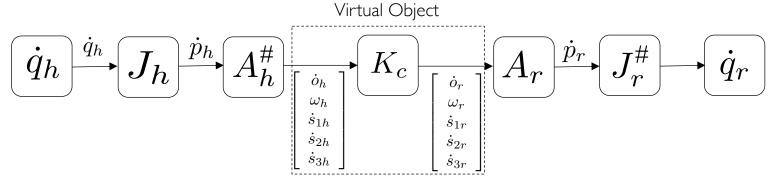


The virtual ellipsoid is the minimum volume ellipsoid Rigid-body motion and semi-axes deformations for the virtual ellipsoids

More accuracy but more parameters

Virtual ellipsoid mapping algorithm





 $J_h, J_r\,$ Jacobian matrix

scaling matrix

$$\dot{p}_{ih} = \dot{o}_h + \omega_h \times (p_{ih} - o_h) + \sum_{j=1}^{3} \dot{s}_{jh} \left[(p_{ih} - o_h)^{\mathsf{T}} \hat{s}_{jh} \right] \hat{s}_{jh}$$

Extended Grasp matrix

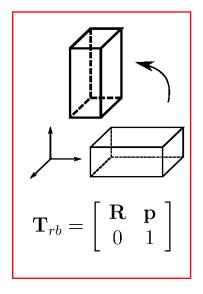
Reference point velocity
$$\dot{p}_{ih} = \dot{o}_h + \omega_h \times (p_{ih} - o_h) + \sum_{j=1}^{3} \dot{s}_{jh} \left[(p_{ih} - o_h)^{\top} \hat{s}_{jh} \right] \hat{s}_{jh}$$

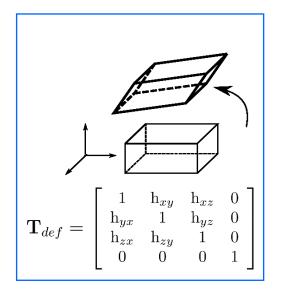
$$A_h = \begin{bmatrix} I & -S(p_{1h} - o_h) & \left[(p_{1h} - o_h)^{\top} \hat{s}_{1h} \right] \hat{s}_{1h} & \cdots \\ \cdots & \cdots & \cdots & \cdots \\ I & -S(p_{ih} - o_h) & \left[(p_{ih} - o_h)^{\top} \hat{s}_{1h} \right] \hat{s}_{1h} & \cdots \\ \cdots & \cdots & \cdots & \cdots \end{bmatrix}$$

Homogeneous transformation

G. Salvietti, M. Malvezzi, G. Gioioso, D. Prattichizzo. On the Use of Homogeneous Transformations to Map Human Hand Movements onto Robotic Hands., ICRA, 2014.

$$\mathbf{T} = \prod_{i=1}^{n_t} \mathbf{T}_{p,i}$$





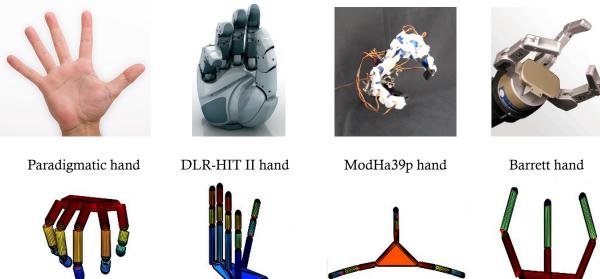
Rigid body motion: the distance between points and angles between vectors is preserved (rotation + translation).

Isotropic and non-isotropic transformations modify the object size by scaling factors in x, y, and z.

Shear transformations displace each point in fixed direction by an amount proportional to its signed distance from a line that is parallel to that direction.

The SynGrasp Toolbox





Syngrasp is a Matlab Toolbox for the analysis of grasping, suitable both for robotic and human hands. Available at http://syngrasp.dii.unisi.it - More than 7500 downloads

The hand modeling allows to define compliance at the contacts and joints level

Controllable forces and object movement, manipulability analysis, grasp quality measures, easy graphical representation of the hands

SIMPLIFICATION BY DESIGN

Robotic soft hands and grippers



Intrinsically compliant – Safe interaction

Underactuation

Tendon driven/pneumatic actuation

Robust

Interaction with the environment

Soft Hands EU project SOMA

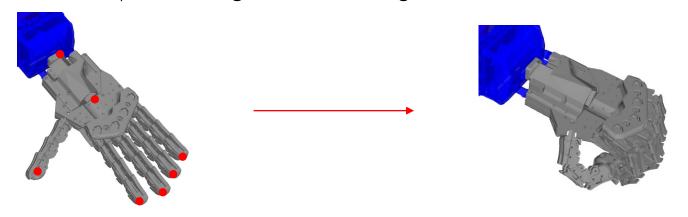
https://www.youtube.com/watch?v=b9pFjNNPtnA



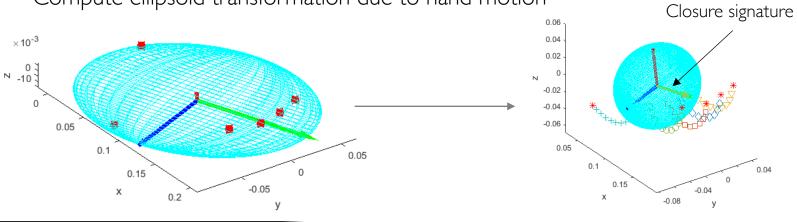
Modeling: Closure signature



Track the reference points during free-hand closing motion

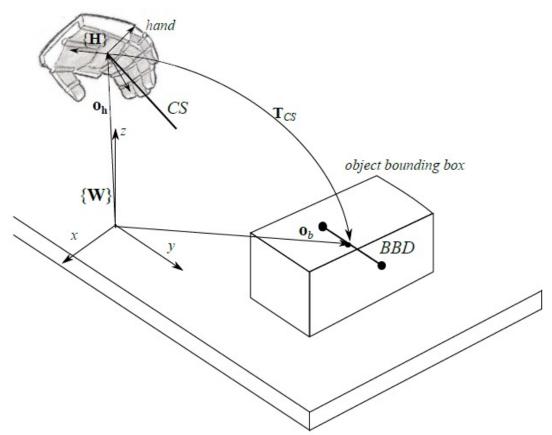


Compute ellipsoid transformation due to hand motion



Planning

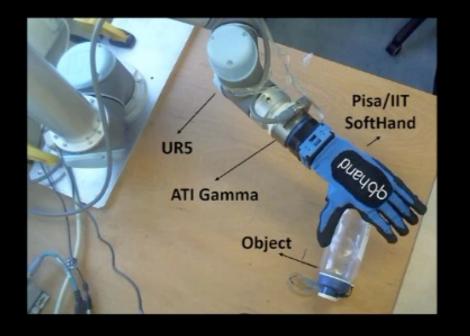
Top grasps exploiting the CS



Closure signature - Experimental validation

Experimental Setup

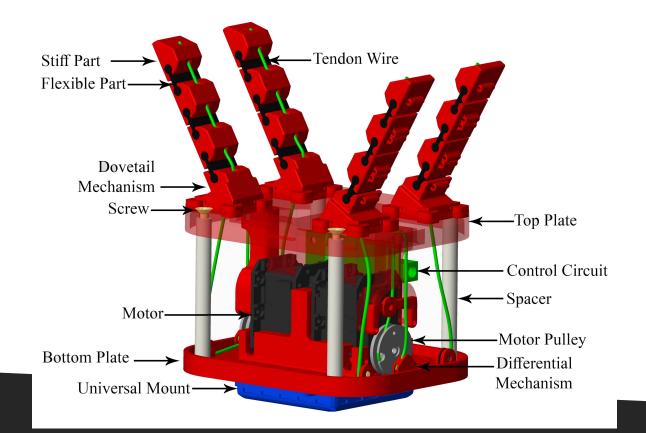
- UR5 robot arm
- ATI force/torque sensor at the wrist
- Pisa/IIT SoftHand
- Xtion PRO depth camera

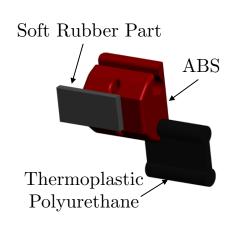


SOFT-RIGID GRIPPER DESIGN

Intrinsically safe and adaptability

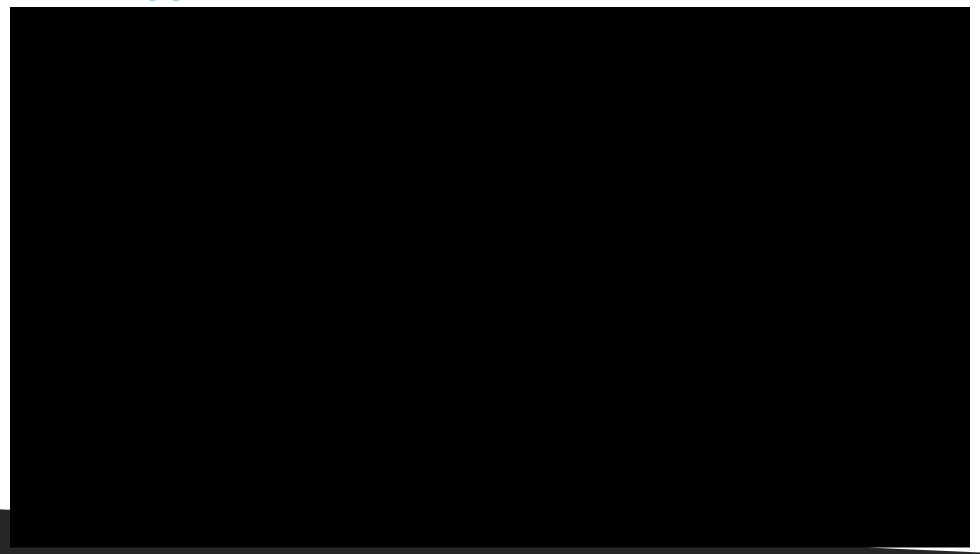
Softness for safety and adaptability



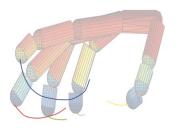


Modular fingers

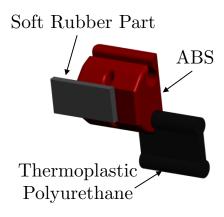
The Co-Gripper



Fingertip trajectory design



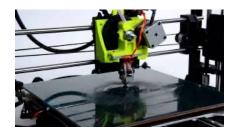
Design/select fingertip trajectories



Modular fingers

$$\mathbf{k}_{q_k} = \mathbf{Q}_k^{-1} \mathbf{T}_k^{\mathrm{T}} \delta \mathbf{f}_k$$

Compute relative joint stiffness ratio



Built the soft module with the K computed

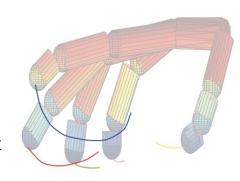


assemble the fingers

From trajectory to stiffness ratio

Relation between tendon applied force f and the torques at joints $oldsymbol{ au}=\mathbf{T}^T\mathbf{f}$

where $\mathbf{T} \in R^{n_t \times n_q}$ is a transformation matrix whose elements depend on finger pulleys' sizes and tendon routing topology and we assume independent from hand posture



We assume that the torque actuated by the motor τ_a balances the torque generated by the deformation of elastic elements in the soft joints

$$\delta \boldsymbol{\tau}_a = \mathbf{K}_q(\mathbf{q}) \delta \mathbf{q}$$

At time instant k, the joint stiffness necessary to move di hand joints of applying a force is computed as

$$\mathbf{k}_{q_k} = \mathbf{Q}_k^{-1} \mathbf{T}_k^{\mathrm{T}} \delta \mathbf{f}_k$$
 where $\mathbf{Q}_k \in R^{n_q imes n_q}, \mathbf{Q}_k = diag(\delta q_k)$

From stiffness to module design



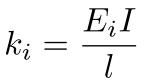
From beam Theory

$$\delta_i = \frac{-f_{rj}hl^2}{2E_iI}$$

$$\theta_i = \frac{-f_{rj}hl_i}{E_iI_i} \qquad I_i = \frac{w_it_i^3}{12}$$

Second moment of area

$$I_i = \frac{w_i t_i^3}{12}$$



E Young's modulus

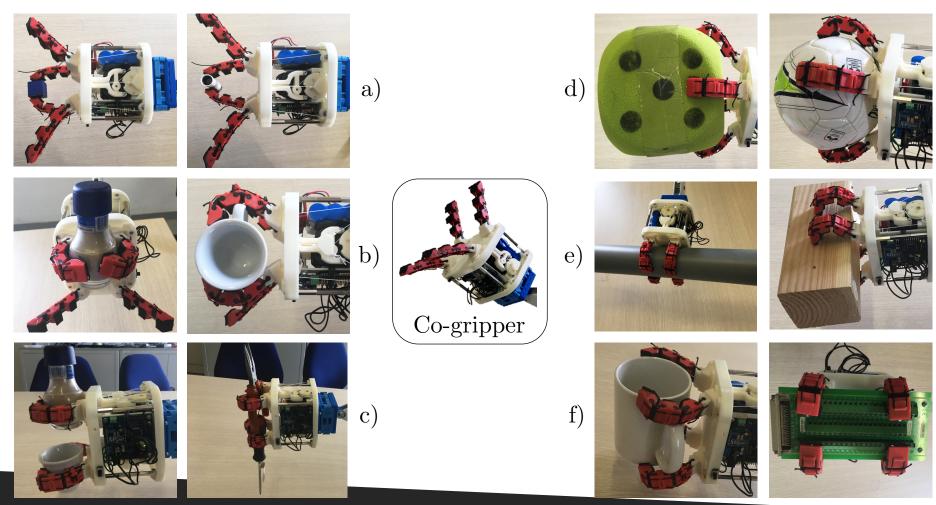
$$k_i = f(\rho_i)$$

infill percentage density p

ř	
Infill density $\rho\%$	E (MPa)
10	1.07
30	1.38
50	2.07
70	6.53
90	9.45
100	10.50

Thermoelastic Polyurethane

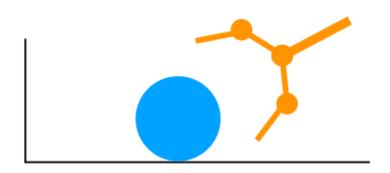
Possible grasps



Two Fingers Four Fingers

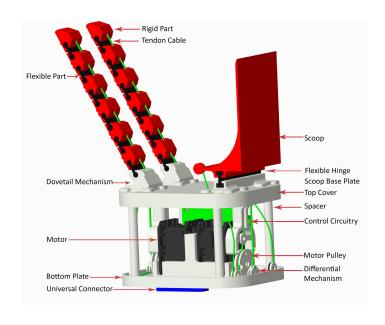
How soft hand can exploit rigid constraints?

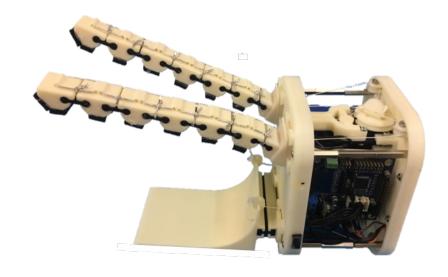
The idea



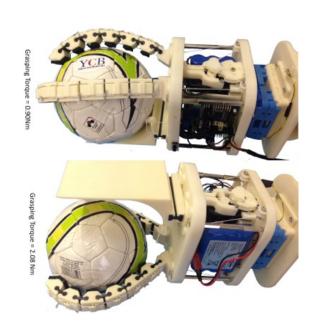


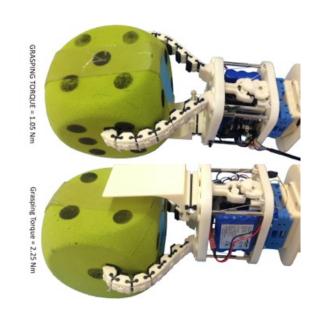
The design



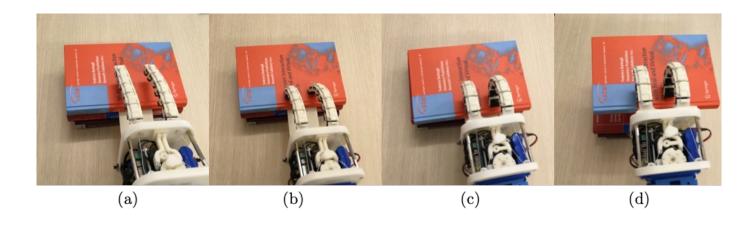


Reduction of grasp force





Grasping strategies



ScoopGripper

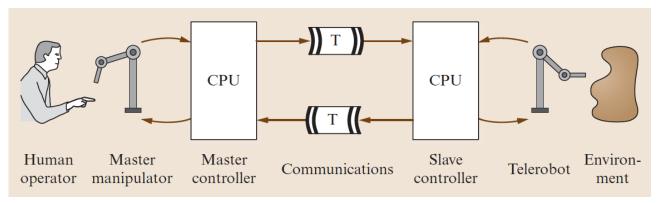


Salvietti et al., ICRA 2019

ROBOTIC TELEOPERATION

Telemanipolazione bilaterale con multicontatto

Bilateral teleoperation with similar master/slave robots



Niemeyer et al., Chapter on Telerobotics, Handbook of Robotics



Robot Simili



Multimodal telepresent control of DLR Rollin' JUSTIN

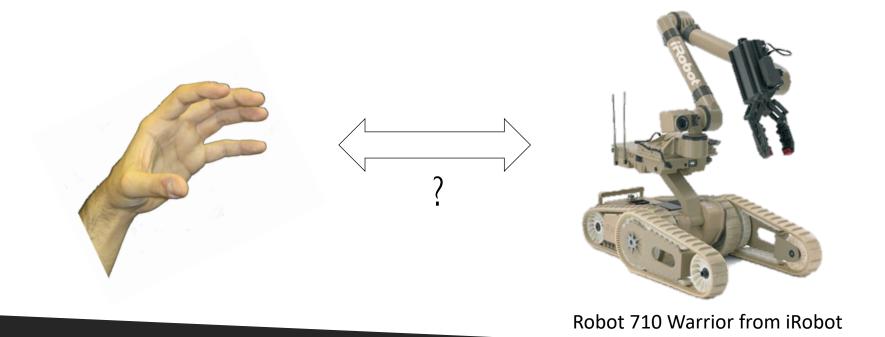
Philipp Kremer, Thomas Wimböck, Jordi Artigas, Simon Schätzle, Klaus Jöhl, Florian Schmidt, Carsten Preusche, Gerd Hirzinger



Dissimilar Kinematics

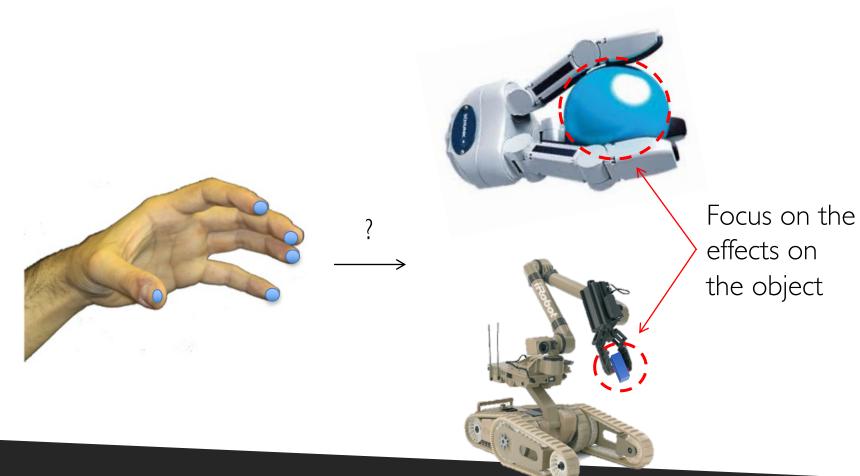
Very dissimilar kinematic structure are typically put in correspondence

Different number of contact points between master and slave

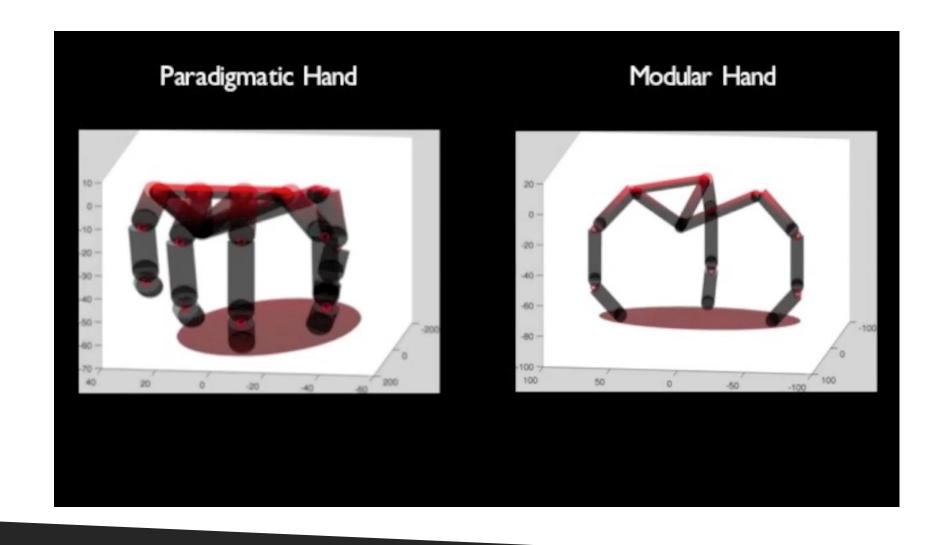


Focus on the object motion

Human hand motions have to be replicated by robotic devices with a dissimilar structure

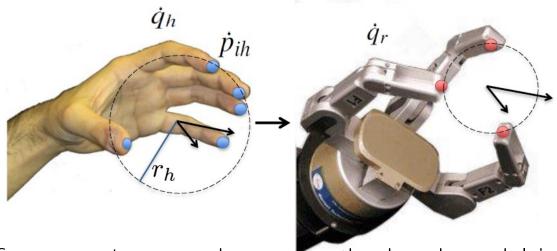


From human hand to robotic hands



A virtual sphere

Virtual sphere mapping

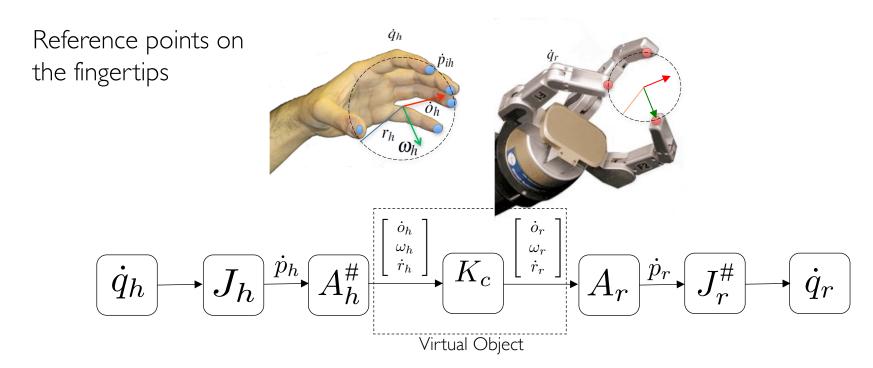


A set of reference points are chosen on the hand model in a given starting configuration

The virtual sphere is the minimum volume sphere containing the reference points

Rigid-body motion and radial deformation for the virtual sphere

Virtual sphere mapping algorithm



 J_h, J_r Jacobian matrix

scaling matrix

$$\dot{p}_{ih} = \dot{o}_h + \omega_h \times (p_{ih} - o_h) + \dot{r}_h (p_{ih} - o_h)$$

Extended Grasp matrix

Reference point velocity
$$\dot{p}_{ih} = \dot{o}_h + \omega_h \times (p_{ih} - o_h) + \dot{r}_h (p_{ih} - o_h) \qquad A_h = \begin{bmatrix} I & -[p_{1h} - o_h]_{\times} & (p_{1h} - o_h) \\ \cdots & \cdots & \cdots \\ I & -[p_{ih} - o_h]_{\times} & (p_{ih} - o_h) \\ \cdots & \cdots & \cdots \end{bmatrix}$$

From robotic hands to human hand

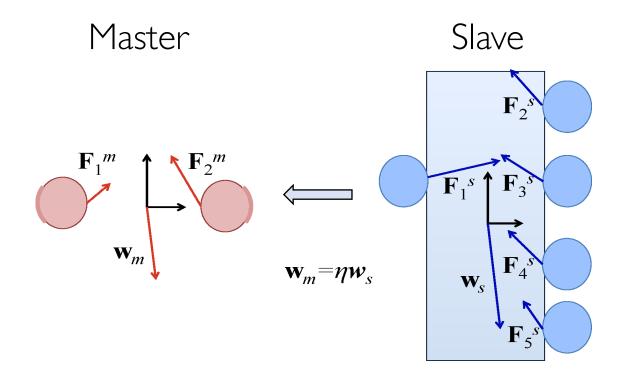
Master Slave

2 contact points

more contact points

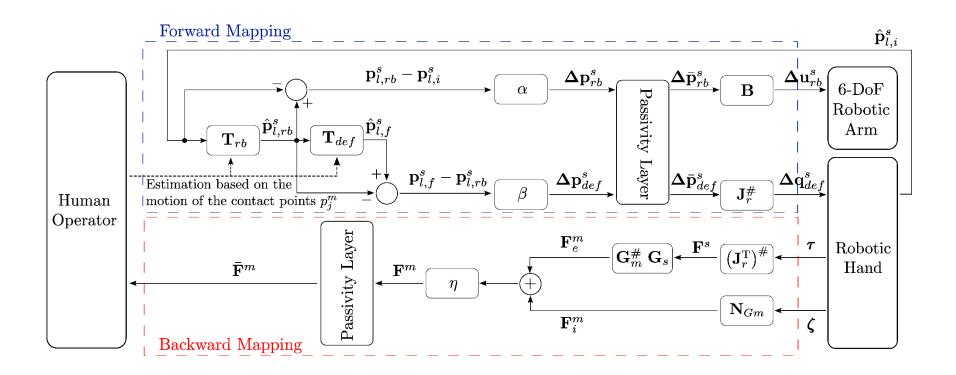
Backward mapping

How to evaluate forces to be rendered on the master side?



We impose that the same wrench is applied on the master virtual object

Schematic overview of the bilateral telemanipulation framework



Experiments

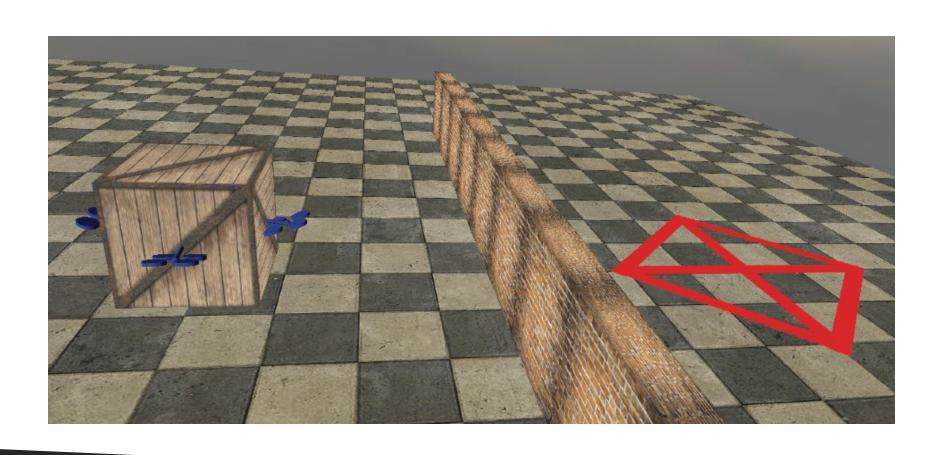


G. Salvietti, L. Meli, G. Gioioso, M. Malvezzi, D. Prattichizzo. **Multi-Contact Bilateral Telemanipulation with Kinematic Asymmetries**. *IEEE/ASME Transaction on Mechatronics*. 2017.

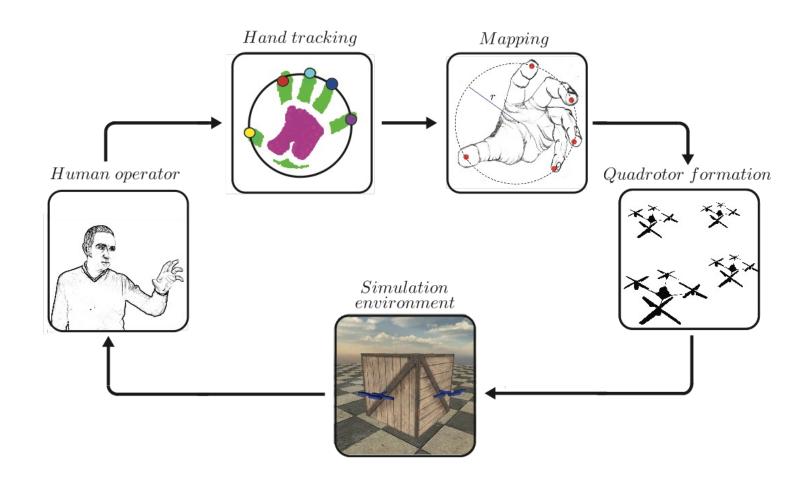
Going wearable...



The flying hand: an UAV formation to cooperatively grasp and carry objects

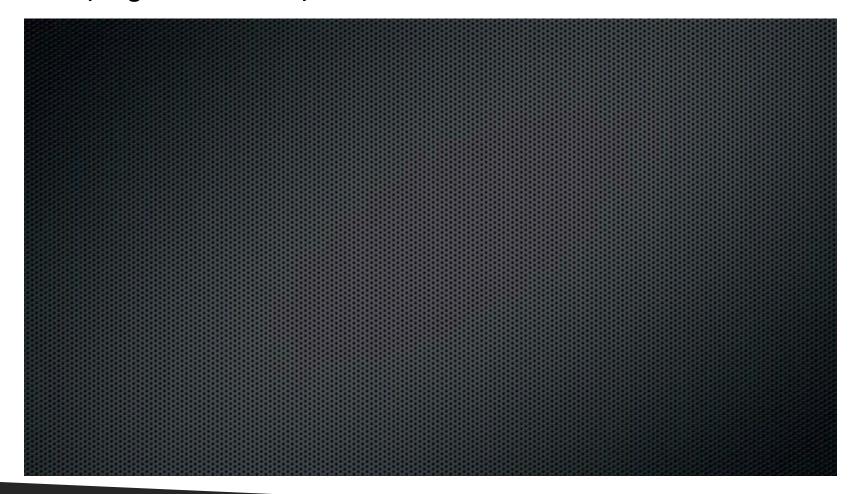


Mapping from hand onto formations



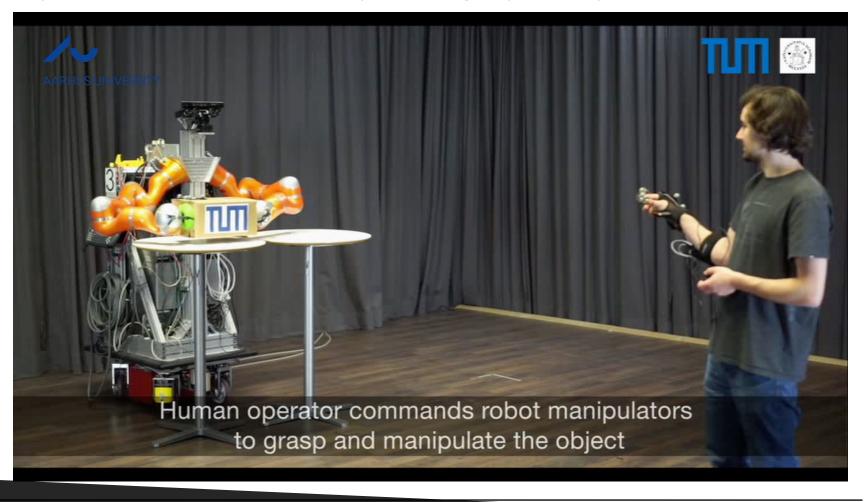
Cooperative Robot Grasping

The flying hand - cooperation with Antonio Franchi CNRS



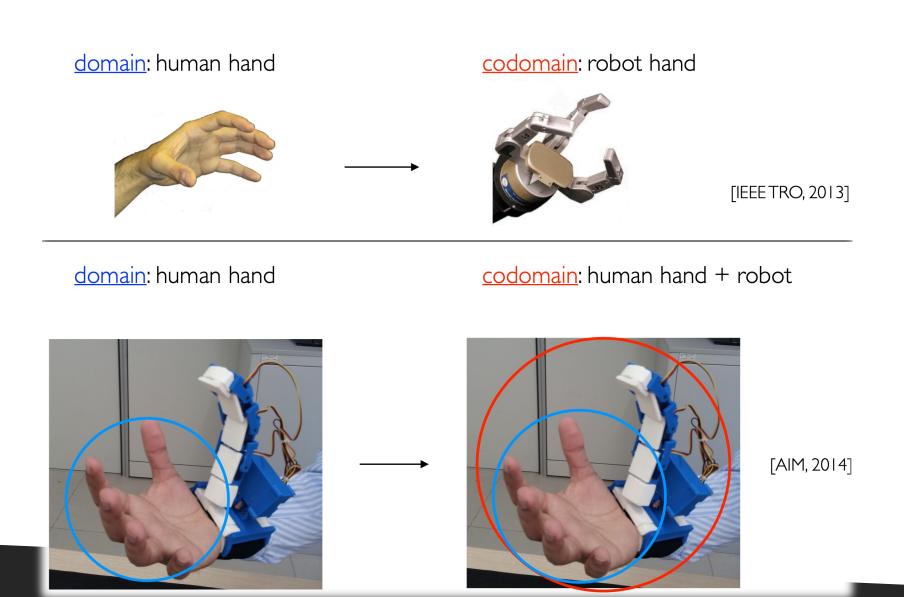
Cooperative Robot Grasping

Haptic feedback in case of cooperative grasp – cooperation with TUM

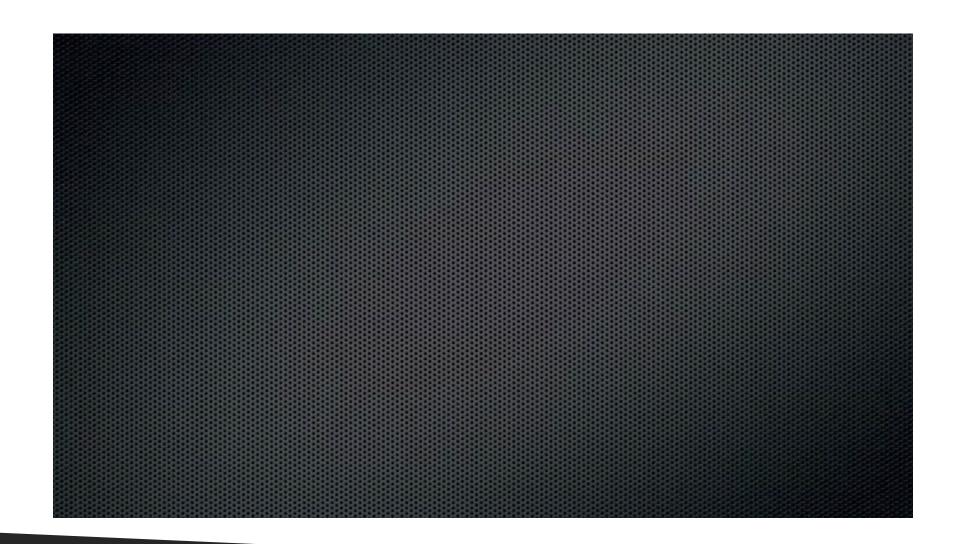


S. Music, G. Salvietti, et al. "Human-Multi-Robot Teleoperation for Cooperative Manipulation Tasks using Wearable Haptic Devices", IROS 17, in review ToH 18.

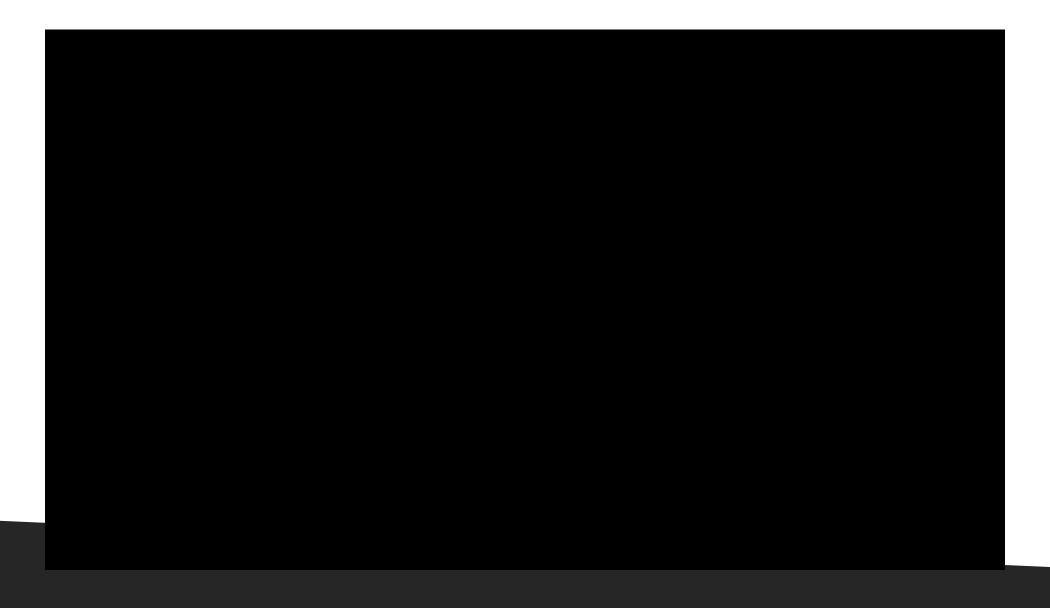
The Sixth Finger: Mapping from domain to codomain



The Robotic Sixth Finger

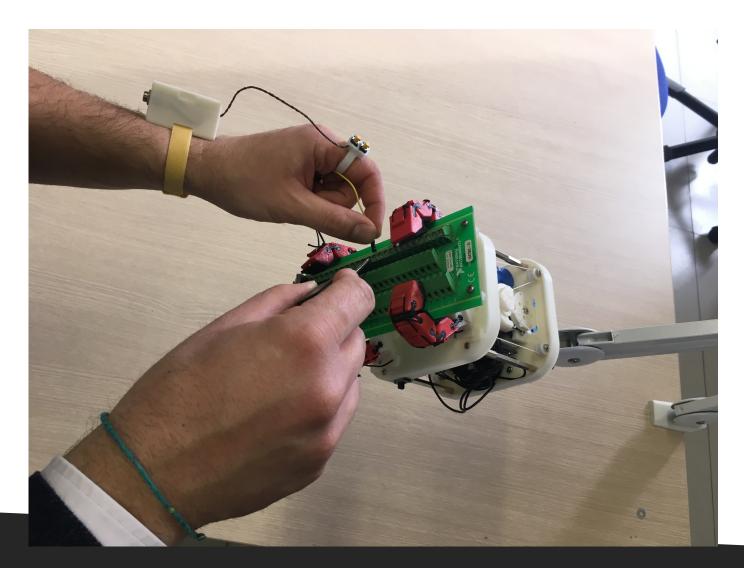


Il Sixth Finger per i pazienti di ICTUS



HUMAN CENTRED ROBOTICS

Cooperative manipulation



Cooperative manipulation

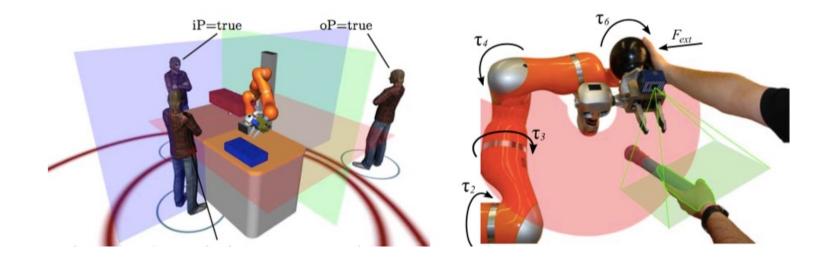


Human intention recognition

Perception

Cameras, e.g., Leapmotion, Kinect

Contact interaction, tactile sensors, F/T sensors



Human intention recognition

Artificial intelligence and machine learning techniques to let robots generalize across the actions captured from humans

In a workplace scenario the set of actions are reduced and could be classified

Implicit vs explicit control



Sensorimotor interfaces

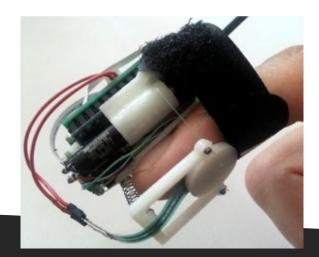
The robot can communicate with the worker through the sense of touch

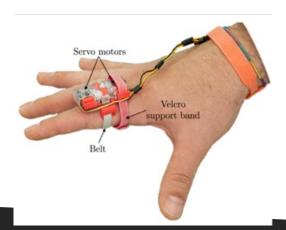
robust communication

wearability and portability

awareness





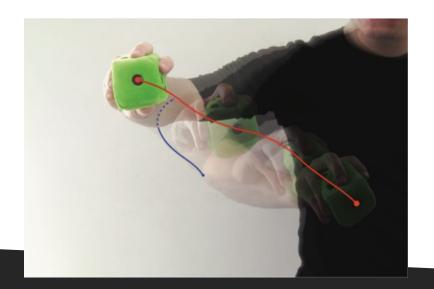


Use of interfaces

Guidance

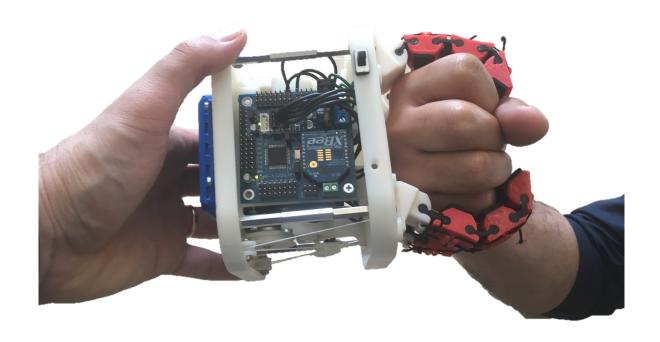
Alert

Force feedback in teleoperation





Design of a novel generation of cooperative grippers for smart factories



Cooperative robotics so far...

Main focus on safety and on robotic arms



Kuka LBR iiwa



Franka Emika Panda



Rethink Robotics Sawyer

Interaction with the operator

Many times at the end-effector level



Collaborative grippers



Co-act SCHUNK



RG6 ThinkBot

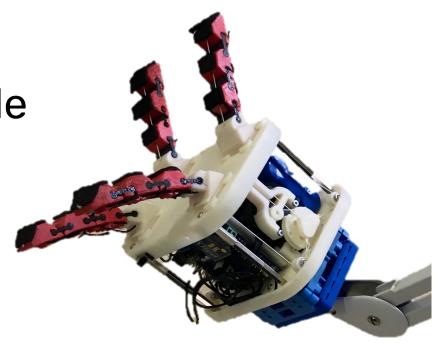
Four main guidelines for cooperative grippers

Intrinsically safe and adaptable

Ease of interface

Portability

Ease of reconfiguration



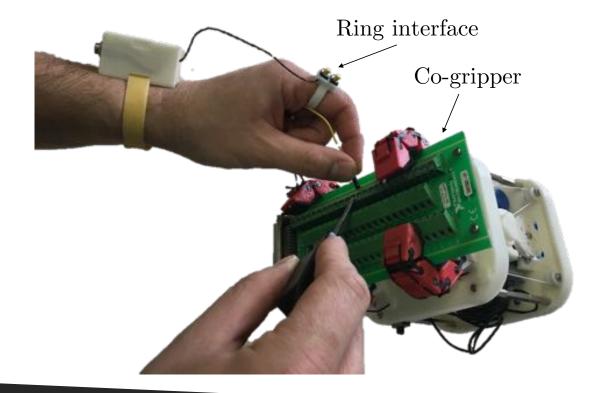
The CoGripper

Intrinsically safe and adaptability

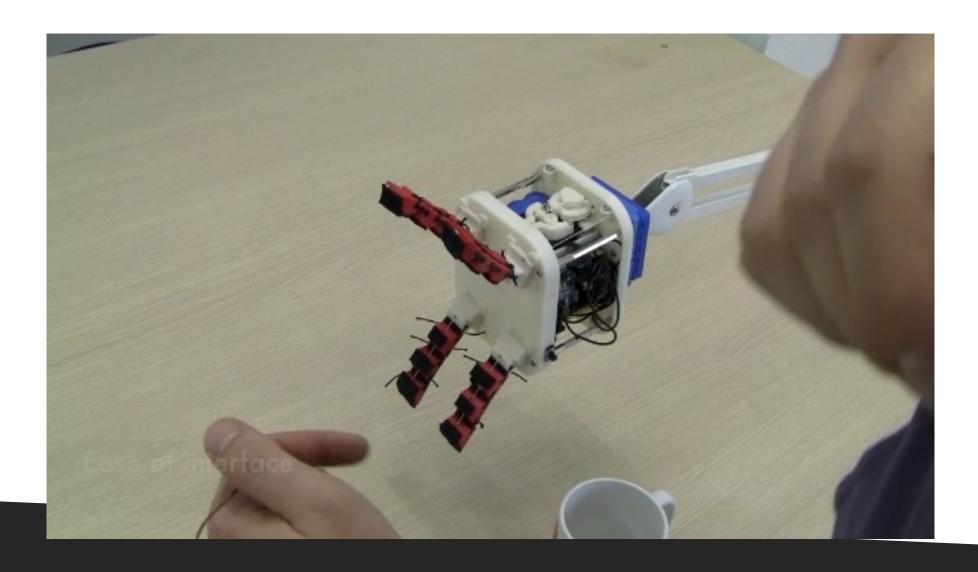


Ease of interface

Intuitive and simple

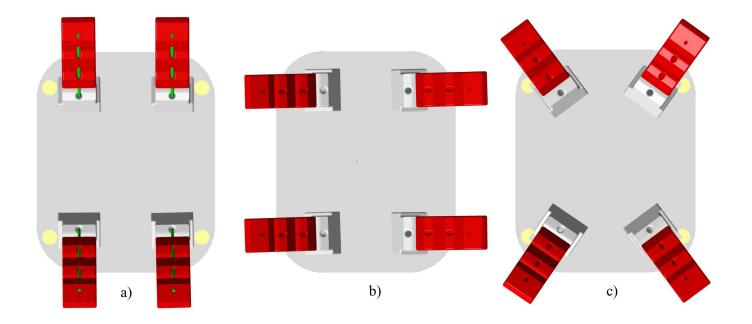


Ease of interface



Ease of reconfiguration

Reorientation of the fingers



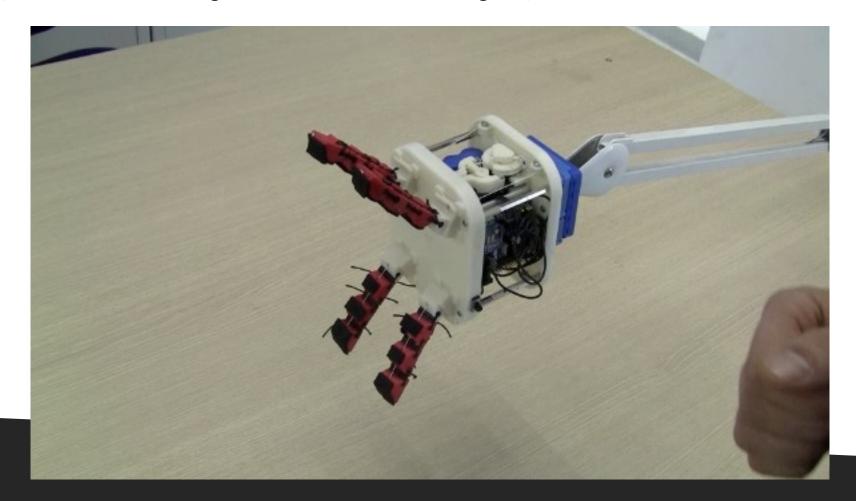
Fingers' orientation on the palm can be manually reconfigured by the operator. The combination of the modularity possible finger trajectory design through stiffness and finger configurations meet the requirements for

Ease of reconfiguration



Ease of reconfiguration

Flexion trajectories of the fingers – Power and Pinch grasps



Portability

The CoGripper is wireless and can be used also with passive supports

Possible to be used also without a robotic arm as support for particular processes

No re-grasp needed



Portability

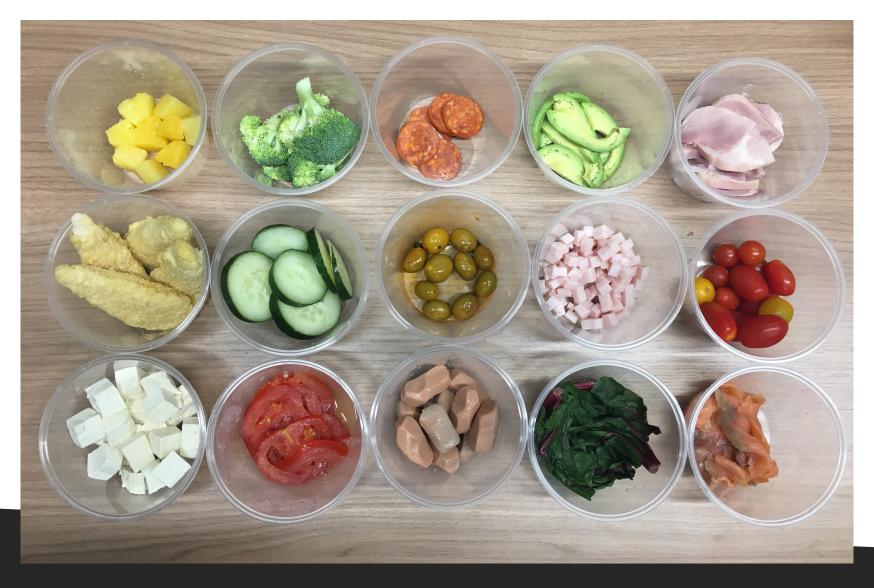


Bilateral haptic collaboration for cooperative grippers



AN EXAMPLE

Gripper for food handling @REMYRobotics

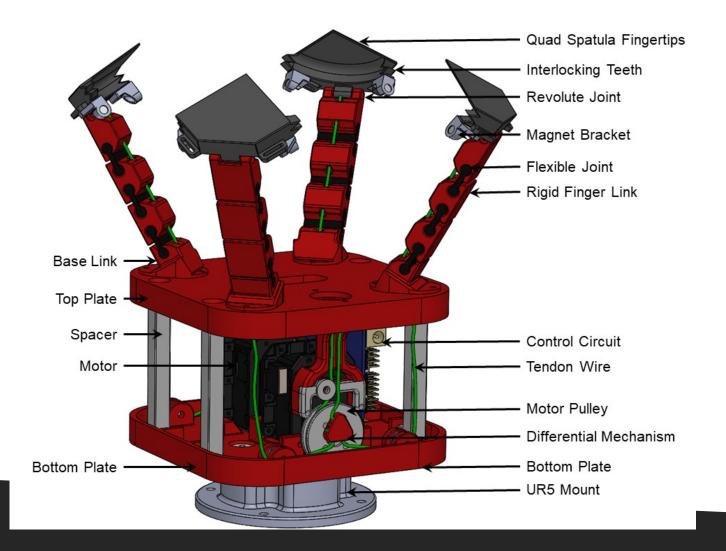


Classic two-fingers grippers

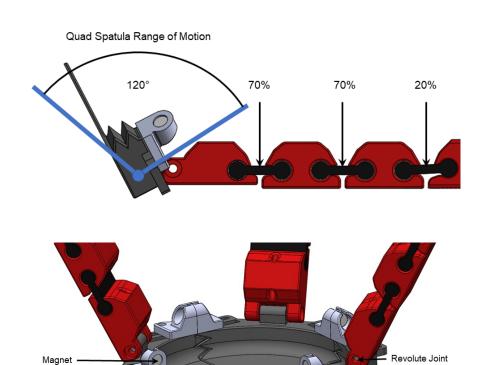


THE IDEA: SPECIALISE THE FINGERTIPS

The Quad-Spatula Gripper



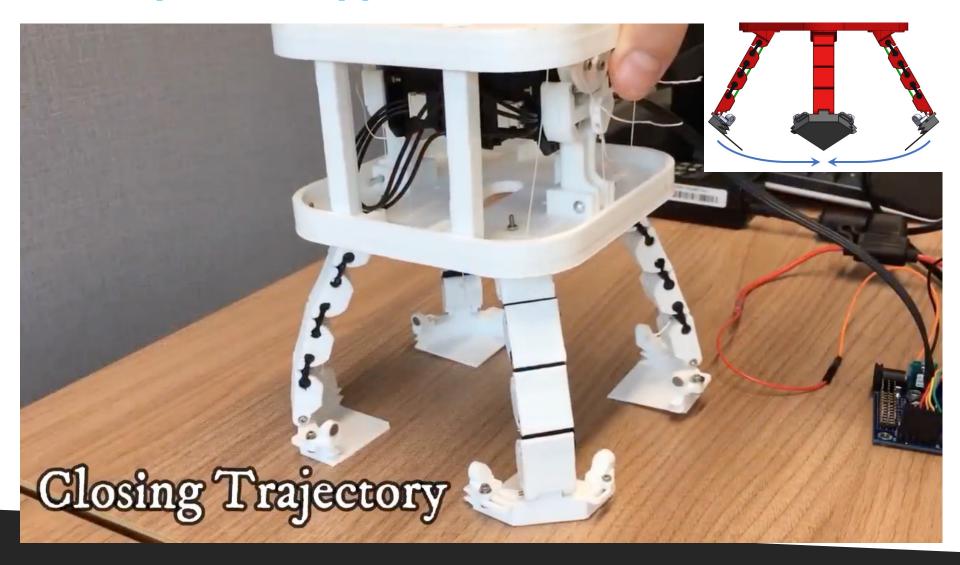
Salvietti et al., Robosoft 2020



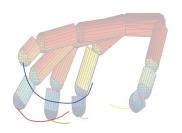
Interlocking Teeth Adjustable Magnet

Quad Spatula

The Quad-Spatula Gripper



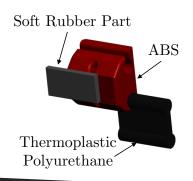
Joint stiffness design



Design/select fingertip trajectories

$$\mathbf{k}_{q_k} = \mathbf{Q}_k^{-1} \mathbf{T}_k^{\mathrm{T}} \delta \mathbf{f}_k$$

Compute relative joint stiffness ratio





$$k_i = f(\rho_i)$$

infill percentage density $\boldsymbol{\rho}$

Built the soft module with the K computed



assemble the fingers

EXPERIMENTS

Grasping ability

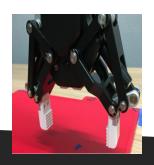


Slices	Strips	Cubes/Irregular
Tomato	Smoked Salmon	Cherry Tomato
Cucumber	Kale	Olives
Pepperoni	Avocado	Tofu Cubes
Cold-Cut Ham		Pineapple Cubes
		Chicken Cubes
		Sausages
		Chicken Nuggets
		Broccoli



Grasping ability

- no ingredient is identical to the other in its set
- all ingredients vary significantly by size and shape
- ingredients can possess one or a combination of properties and at different scales such as sliminess, stickiness, deformability and brittleness

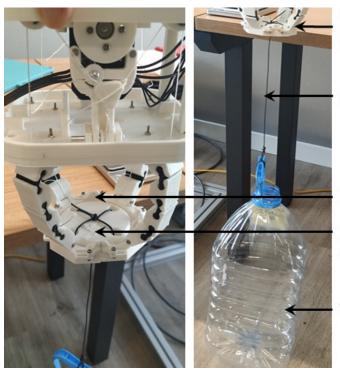






Ingredient	Parallel Gripper	QS-Gripper	QS-Gripper with magnets
Cherry tomato	80%	100%	100%
Tomato slices	40%	100%	100%
Olives	100%	100%	100%
Tofu	90%	100%	100%
Pineapple cubes	90%	100%	100%
Dinosaur Kale Strips	90%	100%	100%
Cucumber slices	60%	90%	100%
Pepperoni slices	80%	100%	100%
Avocado strips	70%	100%	100%
Ham slices	70%	100%	100%
Broccoli	80%	100%	100%
Salmon Strips	70%	90%	100%
Chicken cubes	20%	60%	90%
Chicken Nuggets	90%	100%	100%
Sausage	70%	100%	100%
Total success rate	73.3%	96%	99.3%

Evaluation Interlocking with Magnetic Alignment in the QS



Quad-Spatula Gripper (with magnetic alignment)

Load Bearing Cable

The QS-Gripper was able to achieve 58% successful interlocks without- and 80% with magnetic alignment.

Interlocking

Disc to Transfer Load onto Quad Spatulas

Water Jug

The QS-Gripper was able to hold approximately 335 grams withoutand 735 grams with magnetic alignment.

Evaluation of Uncertainty in Object Positioning



Evaluation of Uncertainty in Object Positioning

