

Biomedical Robotics Applications

BETER REHAB & TRUE REHAB

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21 November 2021

Universitatea Tehnica din Cluj-Napoca
Robotics and Nonlinear Control

The need for rehabilitation



- Strokes are a leading cause of disability [WHO]

By Blausen
Medical
Communica-
tions, Inc.

The need for rehabilitation



- Strokes are a leading cause of disability [WHO]
- Parts of the brain stop functioning

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The need for rehabilitation



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- Strokes are a leading cause of disability [WHO]
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- Impact on quality of life of patients

The need for rehabilitation



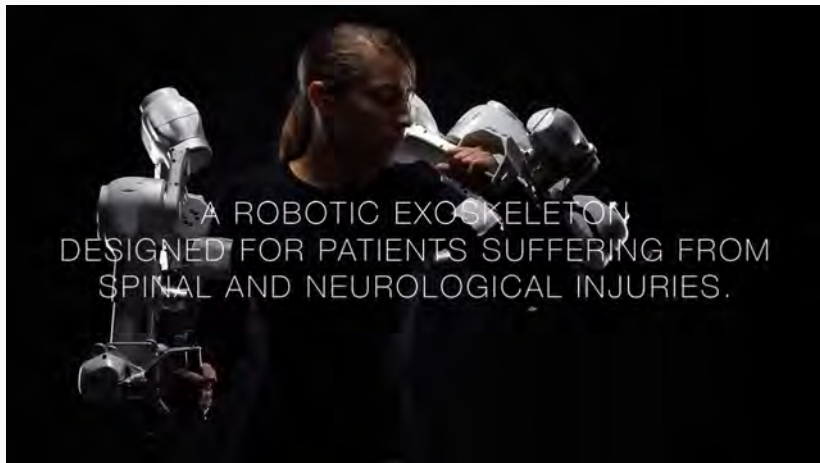
- Strokes are a leading cause of disability [WHO]
- Parts of the brain stop functioning
- Lengthy rehabilitation process
- Impact on quality of life of patients
- Impact on quality of life of **physiotherapists**

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Medical
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Rehabilitation::Traditional rehabilitation



Robotic rehabilitation::Exoskeletons



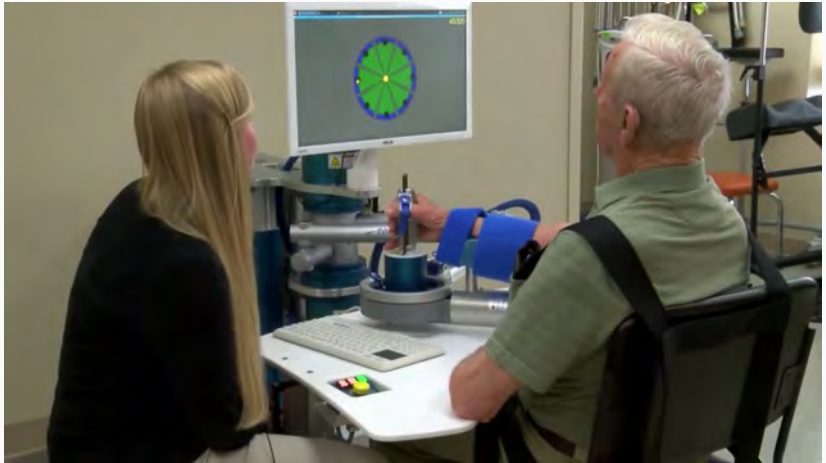
Harmony: Upper-limb Exoskeleton for Stroke Rehabilitation

Robotic rehabilitation::Exoskeletons



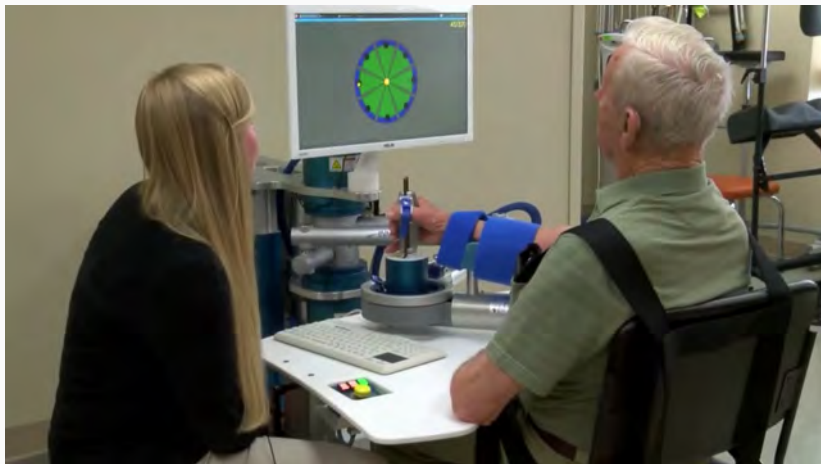
- Very complex mechanics
- Long development times
- Very costly
- Sizing issues
- Long attachment/detachment times (dead time)

Robotic rehabilitation::End-effectors



InMotion Robot

Robotic rehabilitation::End-effectors



InMotion Robot

- Very specific types of motion

Robotic rehabilitation

- Current trend is to mimic 'normal' motion [Díaz et al.]

Robotic rehabilitation

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- Goal of BETER REHAB project: assist patient along intended trajectory

Robotic rehabilitation

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- Newest paradigm: Patient chooses trajectories, physiotherapist assists [Hidler and Sainburg; Lum et al.]
- Goal of BETER REHAB project: assist patient along intended trajectory
- Goal of TRUE REHAB project: patient along muscle optimized trajectory [Caiozzo et al.]

Robotic rehabilitation::Collaborative robotic arms



- Lower development costs
- Much faster attachment
- Patient specific
- Safe

Robotic rehabilitation::Collaborative robotic arms



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Disadvantage

More complex control is necessary

- Intention of motion

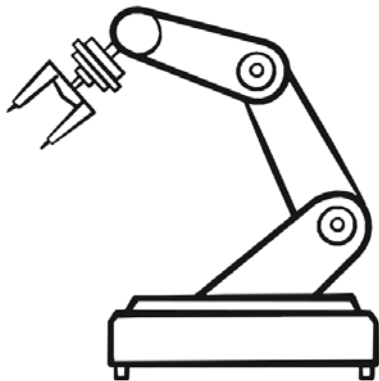
- Intention of motion
- Muscle optimized trajectories

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- Muscle force estimation

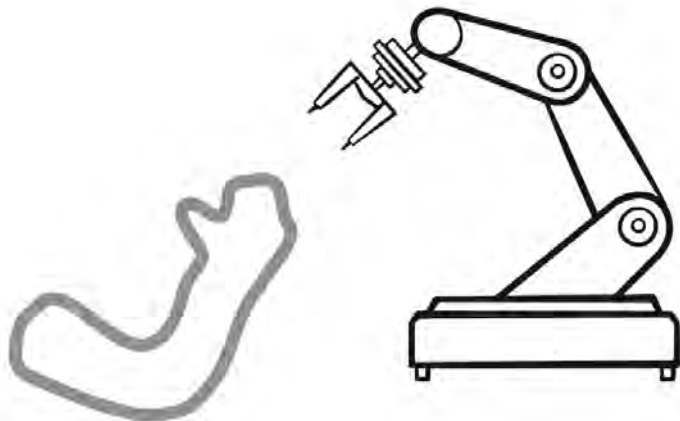
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Intention of motion

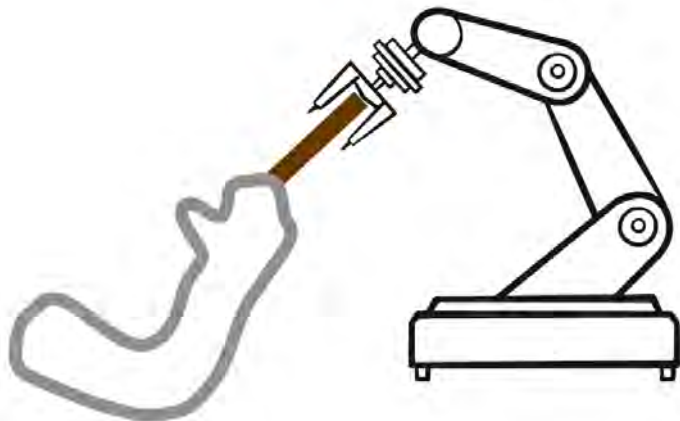
Intention of motion::How?



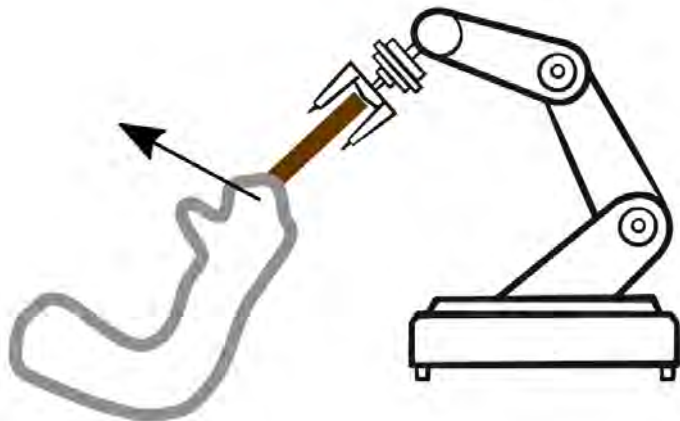
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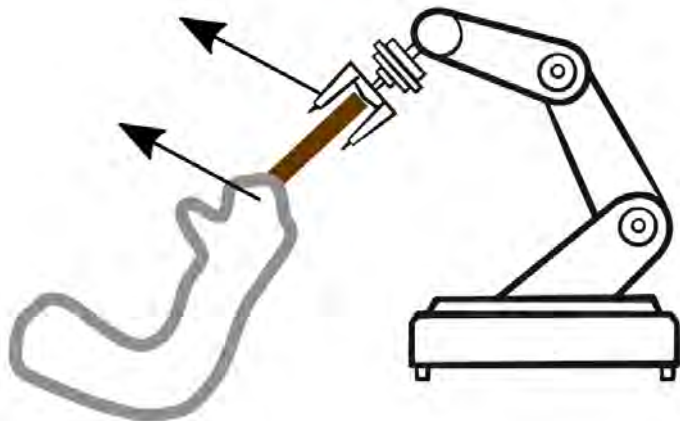
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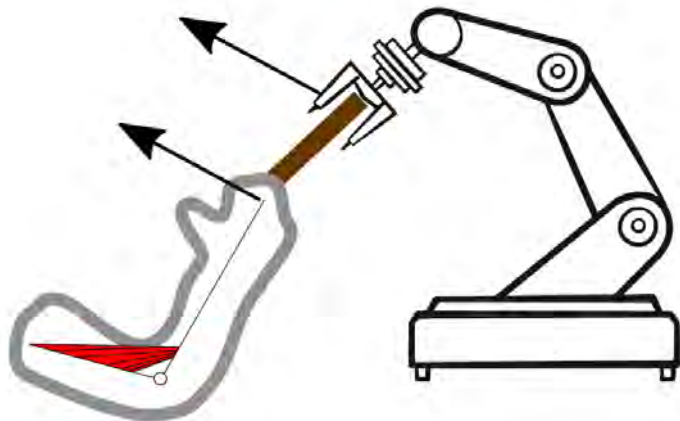
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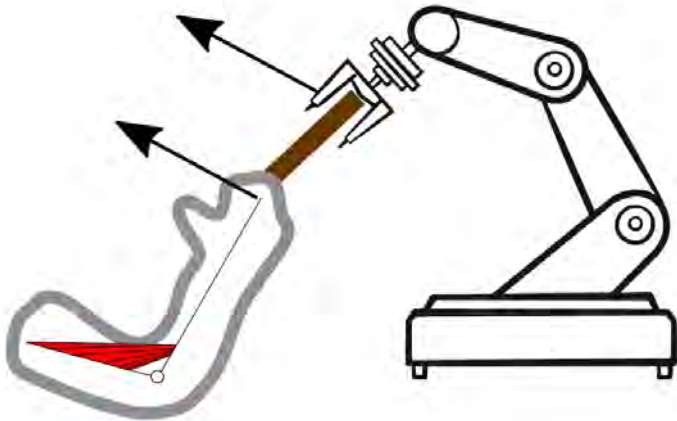
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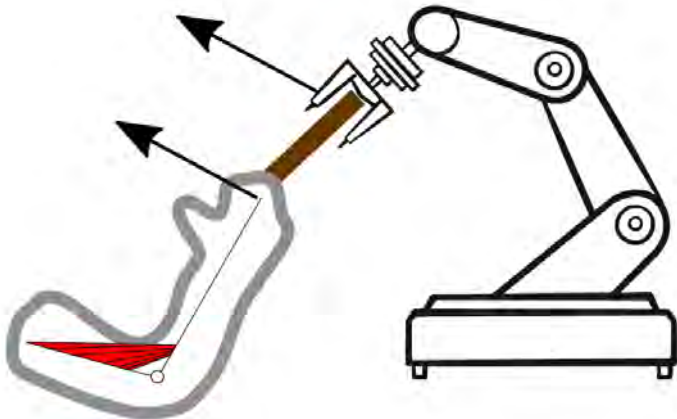


Intention of motion::How?



EMG can help us predict the intention of a person

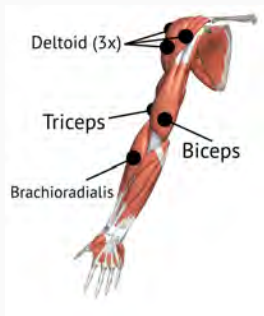
Intention of motion::How?



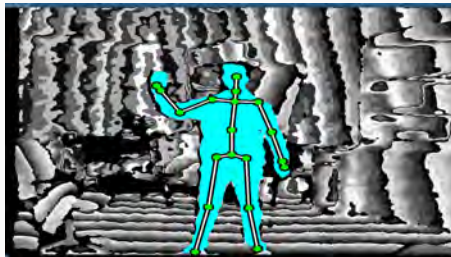
EMG can help us predict the intention of a person

The robot can then assist in the right direction by the desired amount

Intention of motion::Measurements



Intention of motion::Measurements

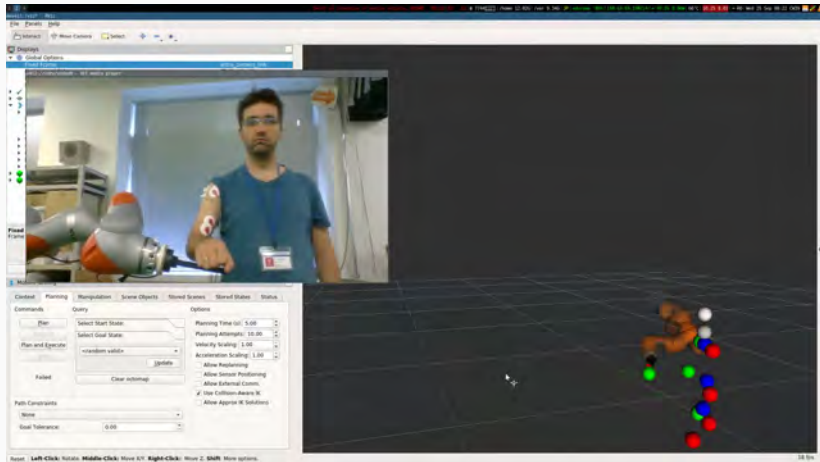


Intention of motion::Prediction

The screenshot displays a software interface for motion planning, likely ROS MoveIt!. It is divided into several sections:

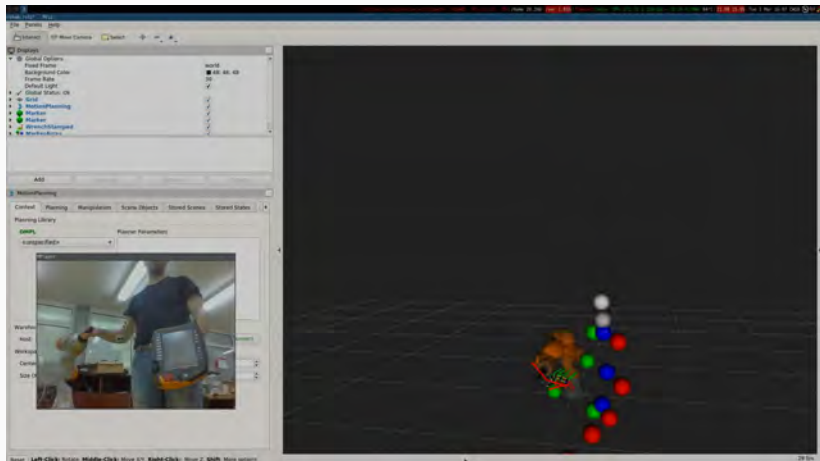
- Top Left:** A video feed showing a person in a blue shirt with sensors on their arm, interacting with a robot arm in a real-world environment.
- Bottom Left:** A control panel with tabs for "Command", "Planning", "Manipulation", "Scene Objects", "Stored States", and "Status". The "Planning" tab is active, showing:
 - Commands:** "Plan" button.
 - Plan and Execute:** "Plan and Execute" button, "random valid" dropdown, and "jupyter" button.
 - Path Constraints:** "None" dropdown and "Goal Tolerance" set to "0.05".
 - Options:** "Planning Time (s): 5.00", "Planning Attempts: 10.00", "Velocity Scaling: 1.00", "Acceleration Scaling: 1.00".
 - Advanced Options:** "Allow Postponing", "Allow Inverse Positioning", "Allow External Control", "Use Collision-Aware IK", and "Allow Approx IK Solutions".
- Right Side:** A 3D simulation environment showing a robot arm model and a sequence of colored spheres (red, blue, green, white) representing a planned path or trajectory.
- Bottom:** A status bar with mouse controls: "Reset", "Left-Click: Rotate", "Middle-Click: Move X/Y", "Right-Click: Move Z", "Shift: Move options", and a "18 fps" indicator.

Intention of motion::Prediction



*43rd International Conference of the IEEE Engineering in
Medicine and Biology Society*

Intention of motion::Experimental validation



Experimental
validation

Experimental validation

Quantifying the human-robot force interaction

Experimental validation

Quantifying the human-robot force interaction

IEEE Robotics Automation Magazine

Muscle optimized trajectories

Operation principle

Calculate trajectories for maximizing/minimizing force production of specific muscle forces

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- Increase efficiency of rehabilitation

Operation principle

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- Train specific muscles

Operation principle

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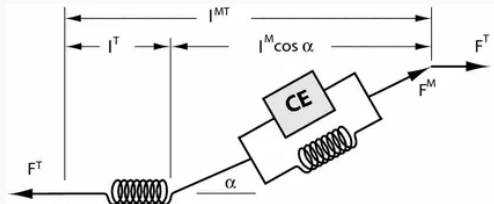
- Increase efficiency of rehabilitation
- Train specific muscles
- Minimize load on sensitive areas (e.g. surgery)

Muscle forces::Estimation

Relationship between muscle activation (EMG) and force is not known

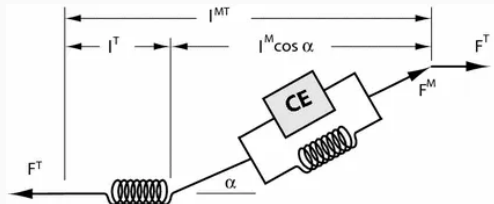
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Muscle forces::Estimation

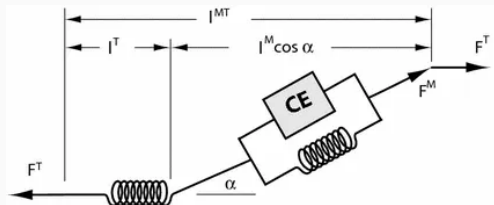
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Several parameters that we should identify [Thelen]

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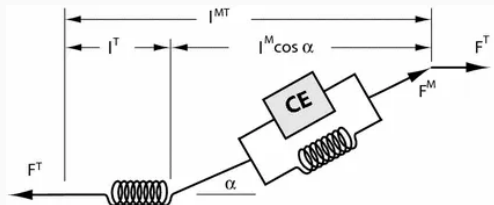


Several parameters that we should identify [Thelen]

- Maximum isometric force

Muscle forces::Estimation

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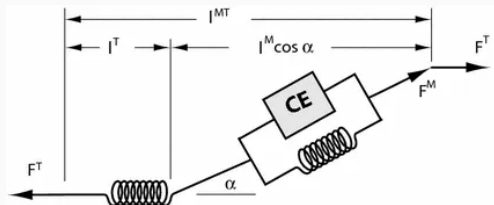


Several parameters that we should identify [Thelen]

- Maximum isometric force
- Tendon slack length

Muscle forces::Estimation

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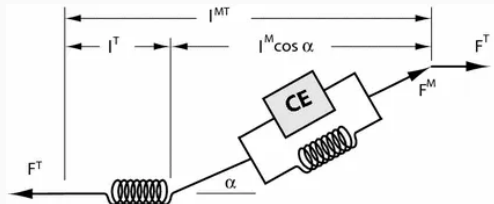


Several parameters that we should identify [Thelen]

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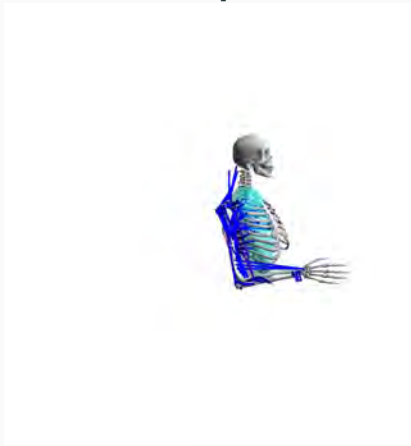
Several parameters that we should identify [Thelen]

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- Pennation angle
- Fiber length

Muscle parameters::Identification

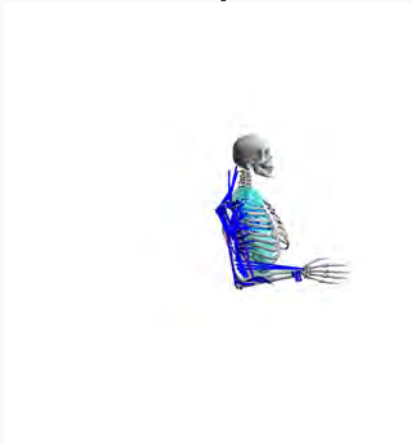
Muscle parameters::Identification

Musculoskeletal modelling [Reed et al.; Blana et al.]



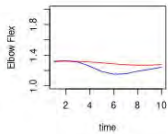
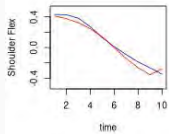
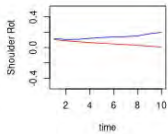
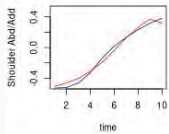
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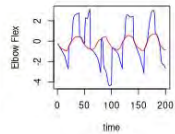
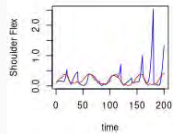
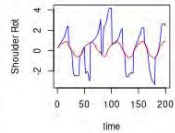
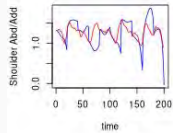
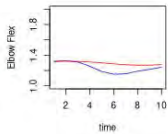
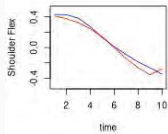
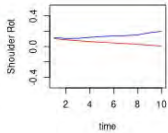
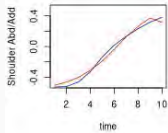


Global Search Optimization methods [Kennedy and Eberhart;
Falisse et al.]

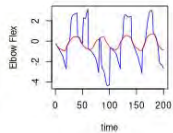
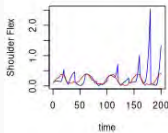
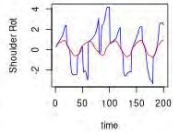
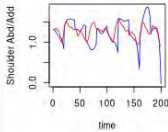
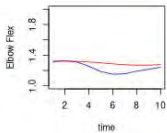
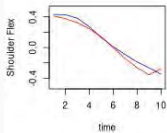
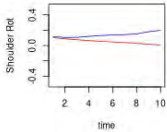
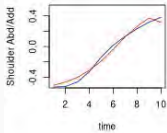
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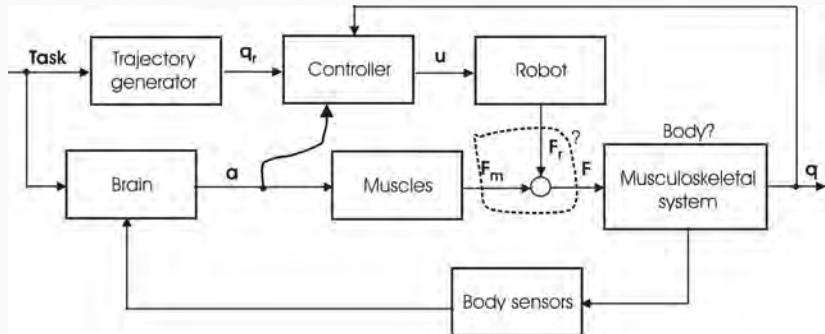
Muscle parameters::Identification



Preparing for *Computer Methods in Biomechanics and Biomedical Engineering*

Robot control

Control::Scheme



Control::Robot model



- No torque control available on the UR5

Control::Practical problems

- No torque control available on the UR5
- Black box controller

Control::Practical problems

- No torque control available on the UR5
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- Uncertainty in motor parameters

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Alternative Controller

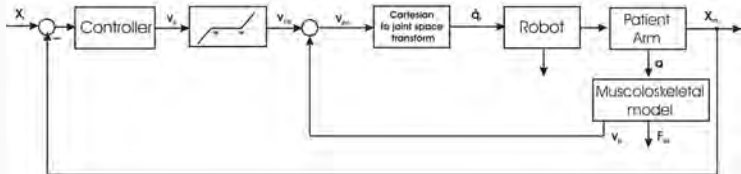
Sliding mode controller can be implemented for the outer position feedback loop (cartesian coordinates) - based on the linearised closed loop model from v_{co} to x_m

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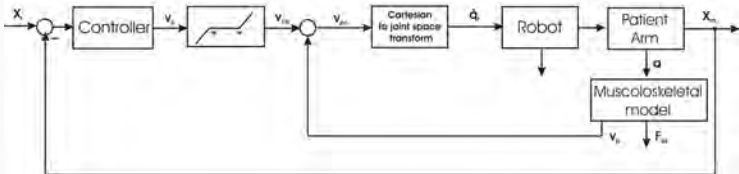


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Controller identification

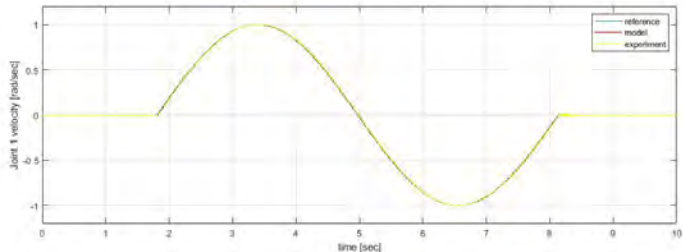
We have successfully identified the structure and parameters of the internal controllers of the robot

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Motor parameters identification

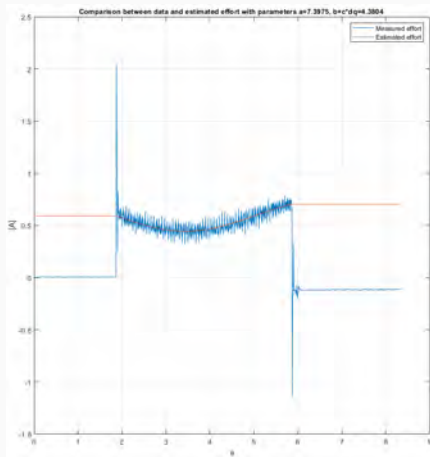
- Current-torque relationship
- Static and dynamic friction

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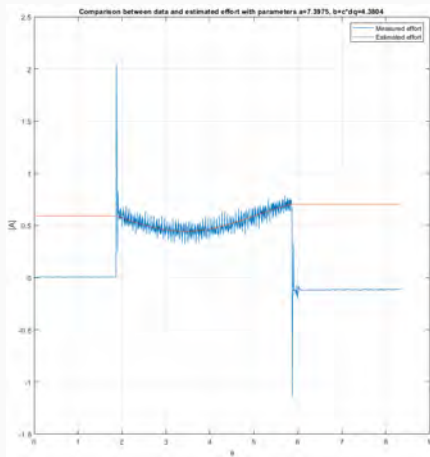
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Preparing for *ICONS 2022*



Funding

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Questions?

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