CusToM : a Matlab toolbox for musculoskeletal simulation

Charles Pontonnier, Pierre Puchaud

CusToM : a Matlab toolbox for musculoskeletal simulation

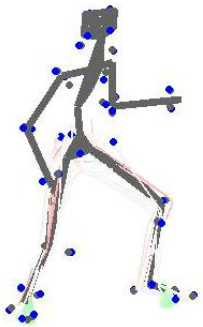
Introduction to musculoskeletal analysis

Charles Pontonnier, Pierre Puchaud

Introduction

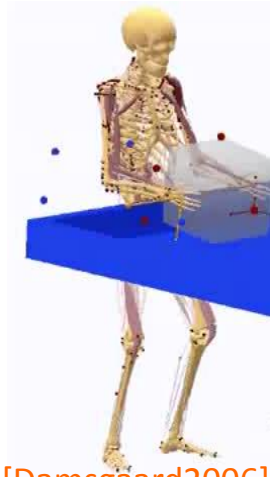
[Pouliquen2015]

[Murai2010]

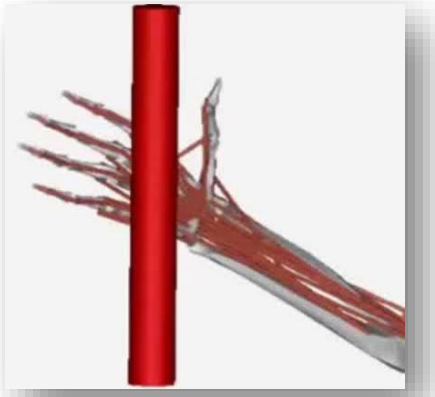


[Plantard2017a]

[Delp2007]

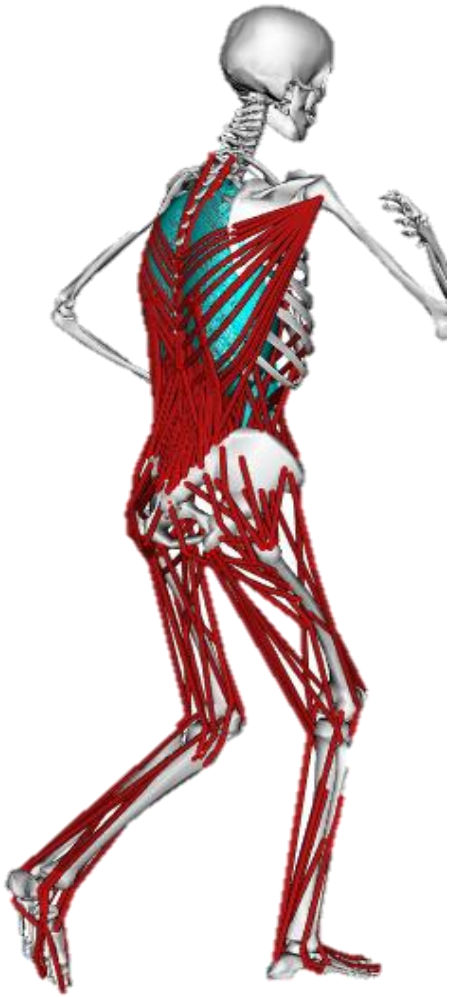


[Damsgaard2006]



[Vignais2014]

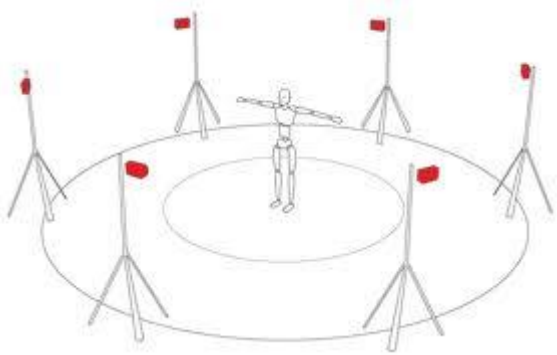
Musculoskeletal analysis

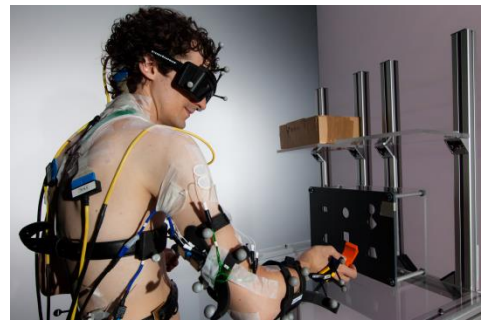


- Angular trajectories
- Joint forces
- Muscle forces

Source : OpenSim

Input data: motion capture (and force platforms)

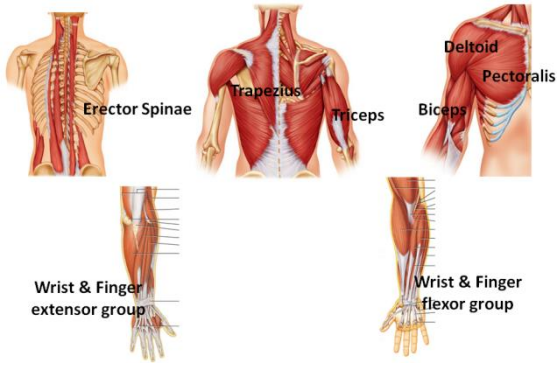


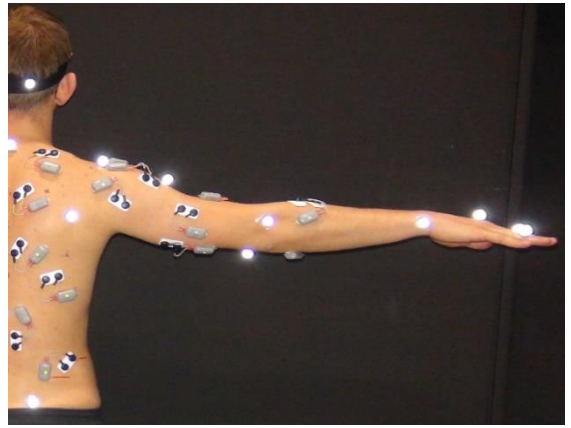


- Reflecting markers
- Infrared cams
- Triangulation

Input data (optional): sEMG

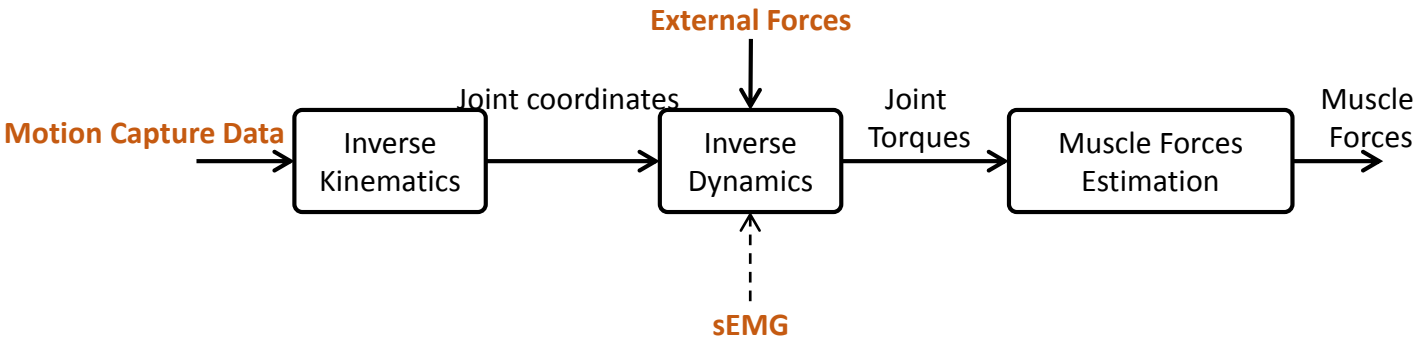
- Measuring electrical activity of muscles
- Classicaly voltage between two points of the muscle chief (bipolar)



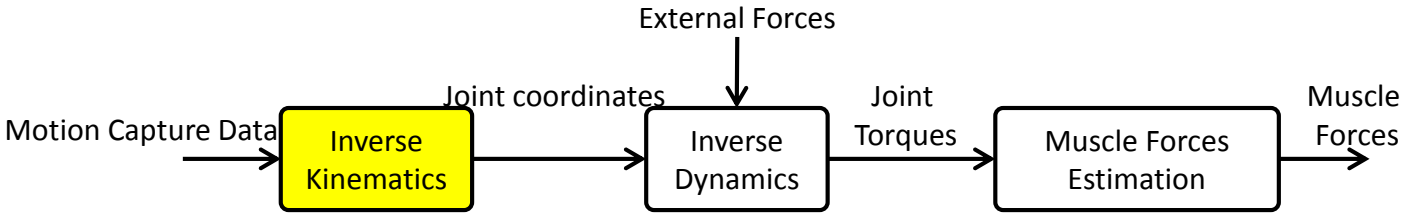


SENIAM, recommendation for electrode placements

Motion analysis (inverse dynamics approach)

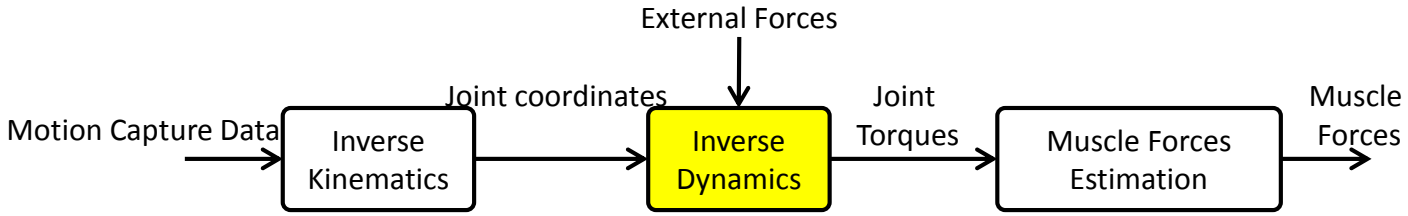


Inverse kinematics



- Joint coordinates computation
- Classically constrained optimization

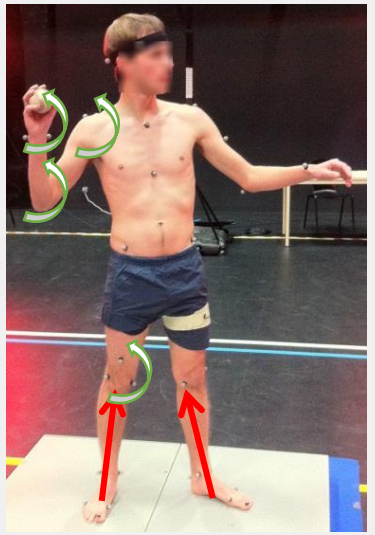
Inverse dynamics



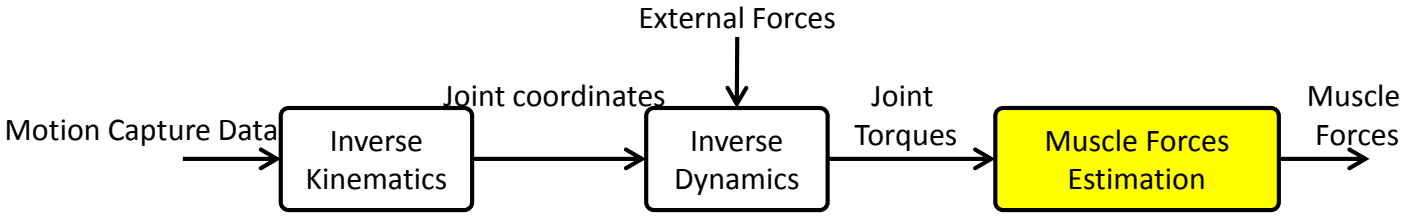
- Joint torques determination
- Classically Newton-Euler algorithm

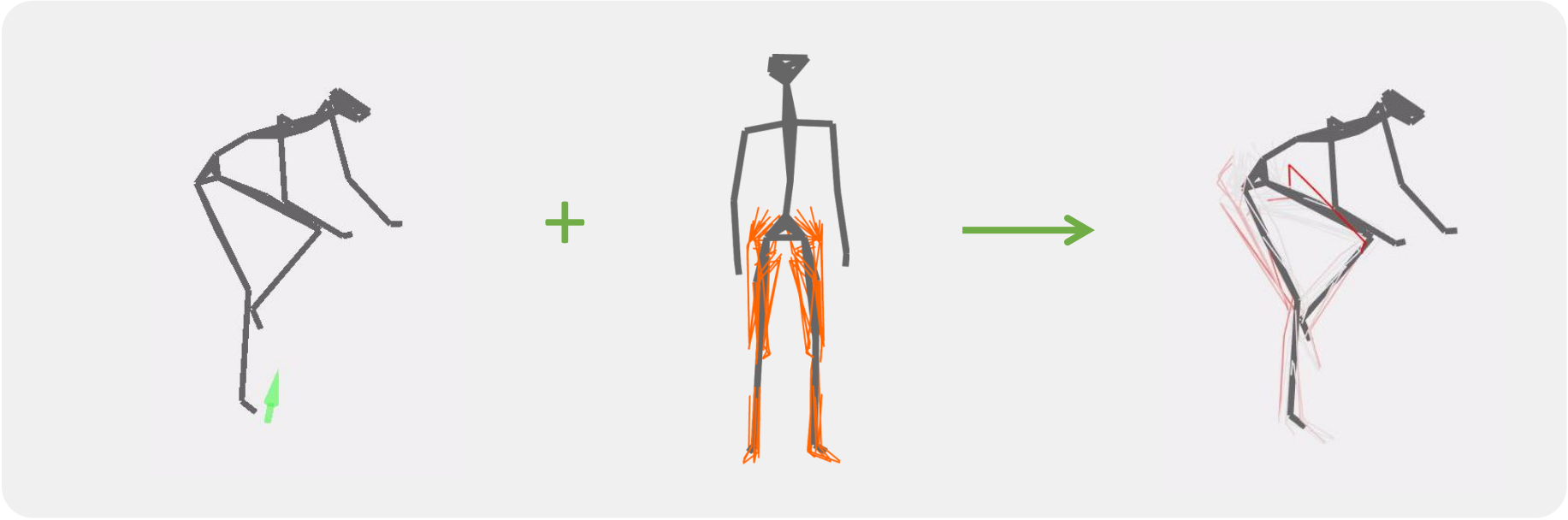
$$f_i = f_i^B - f_i^x + \sum_{j \in \mu(i)} f_i$$

External forces measures



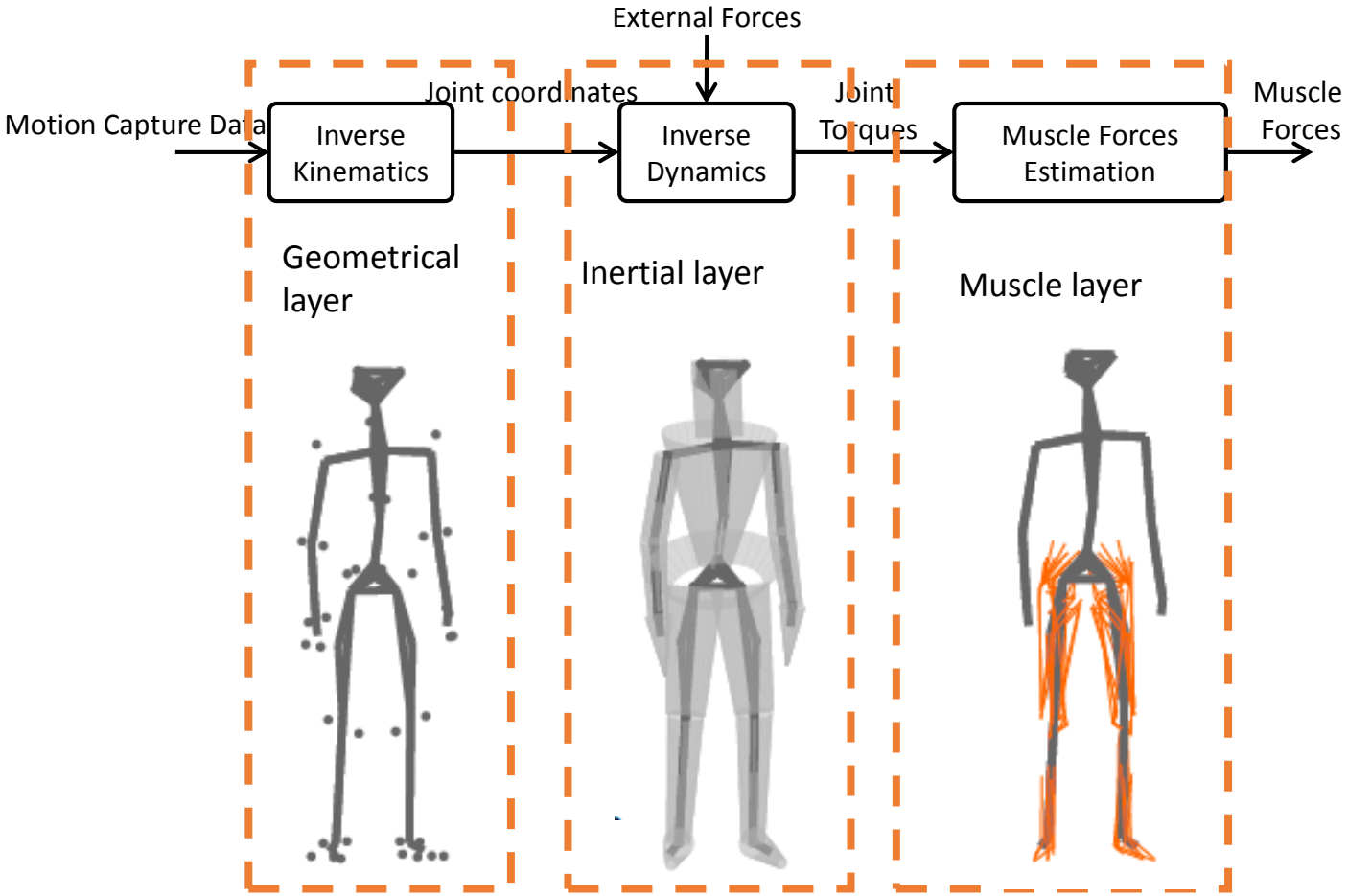
Muscle forces estimation



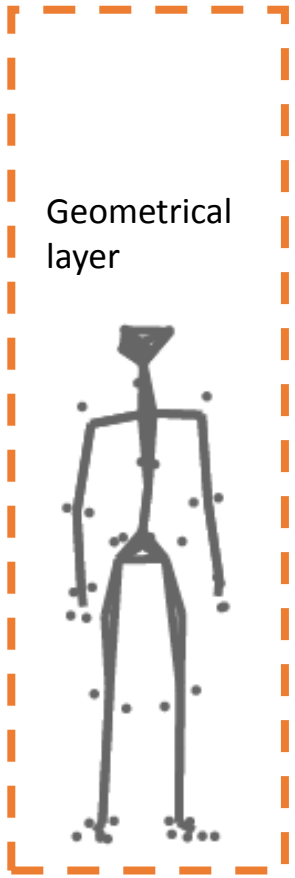


- Classically through non-linear constrained optimization

Musculoskeletal model

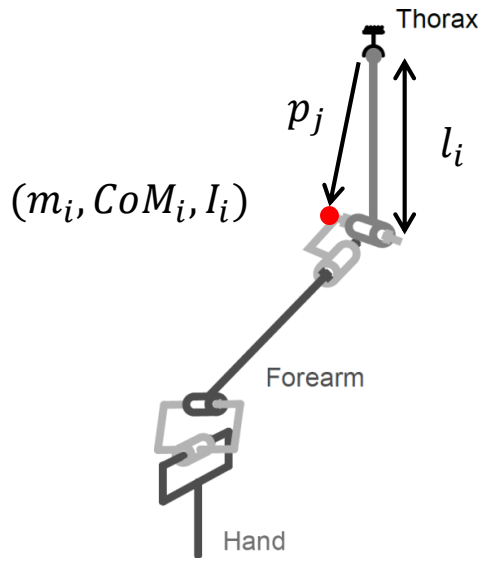


Geometrical layer

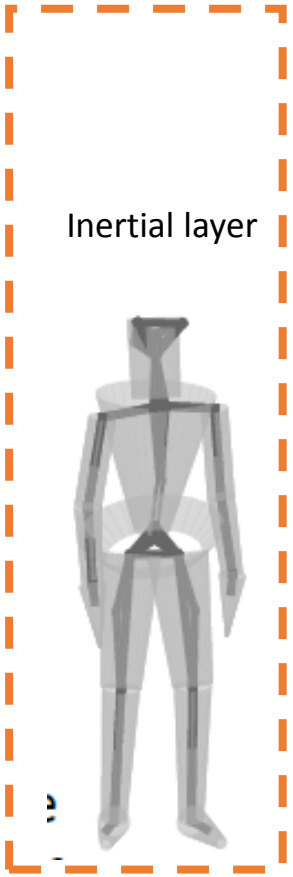


Oestoarticular model

- Polyarticulated rigid body system
- Kinematics joints
- Geometrical properties

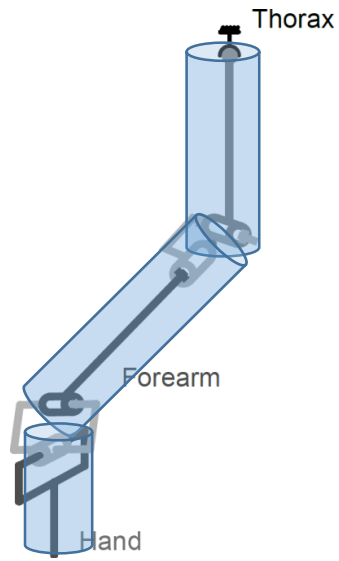


Inertial layer



Oestoarticular model

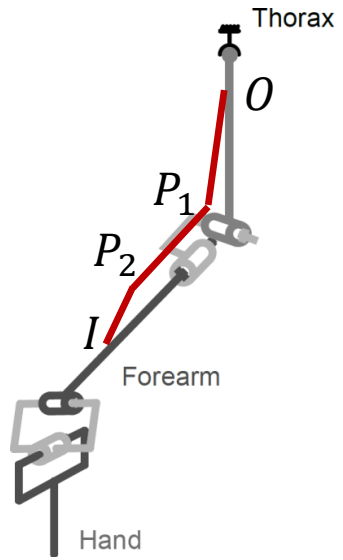
- Inertial properties (mass, center of mass, inertia matrix)



Muscle layer

Muscle model

- Muscular topology



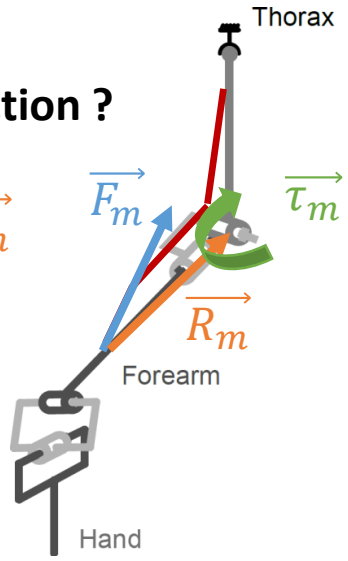
Muscle layer

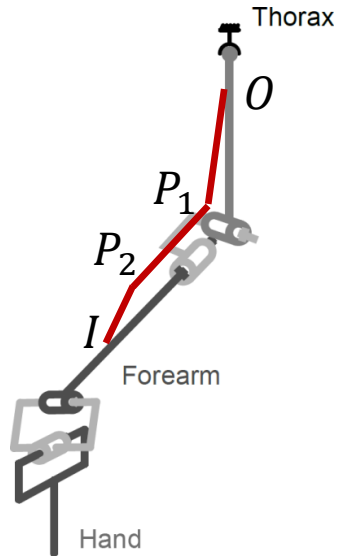
Muscle model

- Muscular topology

Muscle action ?

$$\overrightarrow{\tau_m} = \overrightarrow{F_m} \times \overrightarrow{R_m}$$

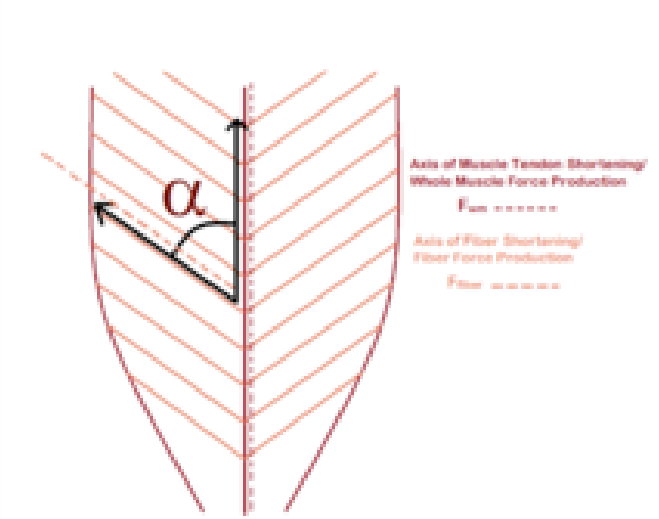


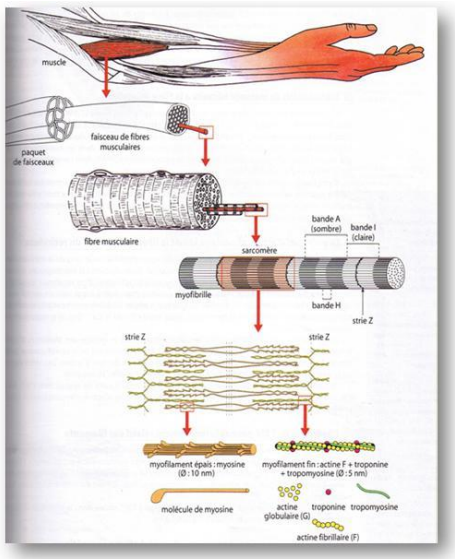


Muscle layer

Muscle model

- Force generation behavior

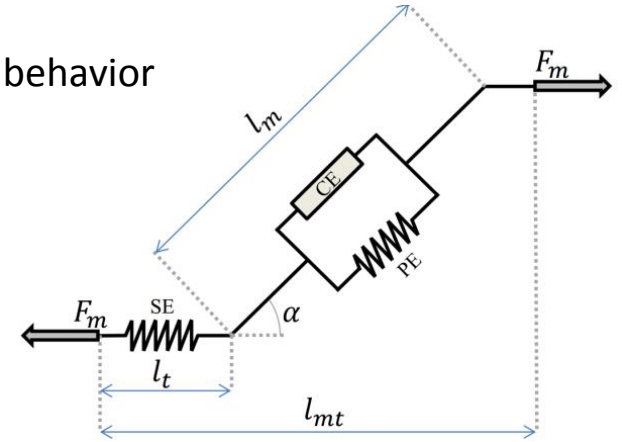




Muscle layer

Muscle model

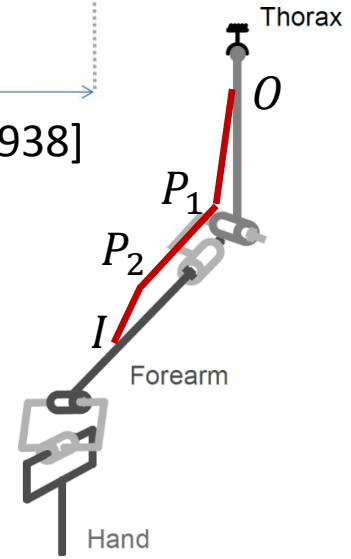
- Force generation behavior



Hill model [Hill1938]

$$F_{m,j} = \left[f_p\left(\bar{l}_{m,j}\right) + f_a\left(a_j, \bar{l}_{m,j}, \dot{\bar{l}}_{m,j}\right) \right] F_{o,j}$$

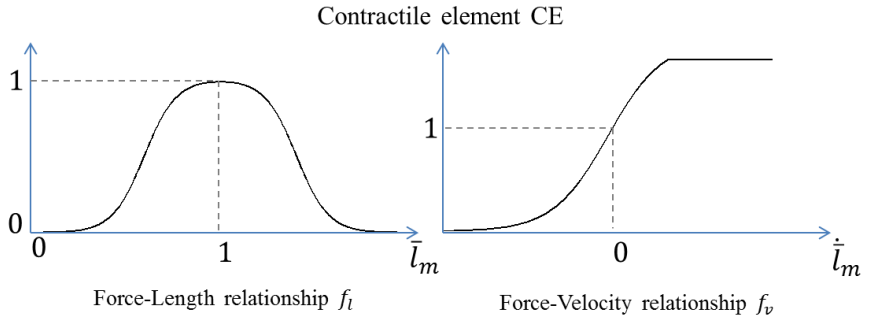
Visco-elastic law

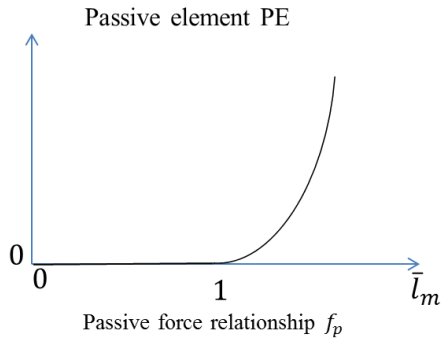


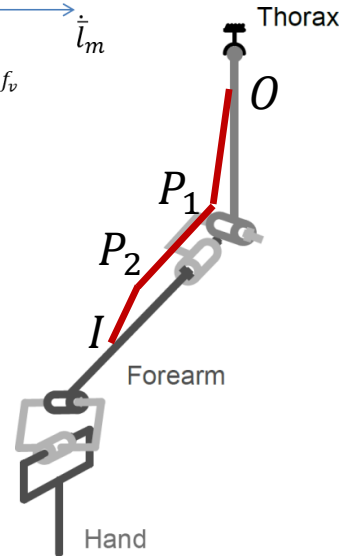
Muscle layer

Muscle model

- Force generation behavior







$$F_{m,j} = \left[f_p\left(\bar{l}_{m,j}\right) + f_a\left(a_j, \bar{l}_{m,j}, \dot{\bar{l}}_{m,j}\right) \right] F_{o,j}$$

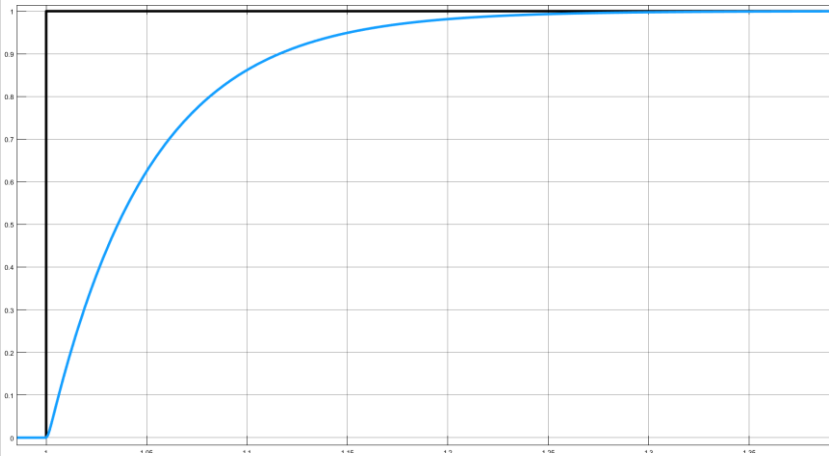
Muscle layer

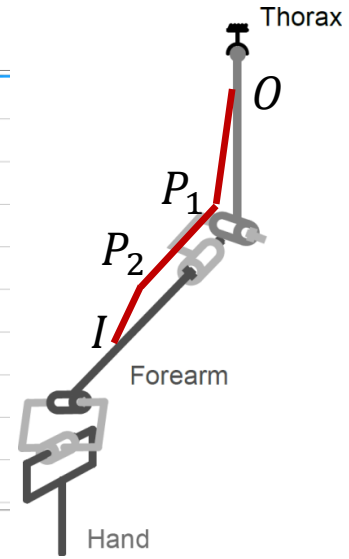
Muscle model

$$\dot{e}_j = (u_j - e_j)/\tau_{ne}$$

- Activation dynamics

$$\dot{a}_j = \begin{cases} (e_j - a_j)/\tau_{act} & , \quad e_j \geq a_j \\ (e_j - a_j)/\tau_{deact} & , \quad e_j < a_j \end{cases}$$



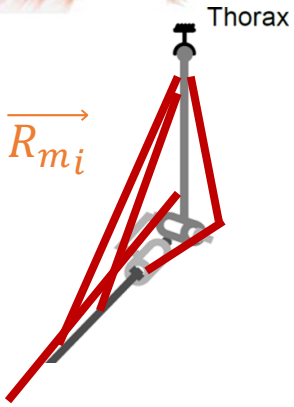


Muscle layer

Muscle model

- Motor control

$$\vec{\tau} = \sum_i \overrightarrow{F_{m_i}} \times \overrightarrow{R_{m_i}}$$



Muscular redundancy

Muscle layer

Muscle model

- Motor control

  → minimising a function representing the motor control to define a muscle recruitment law

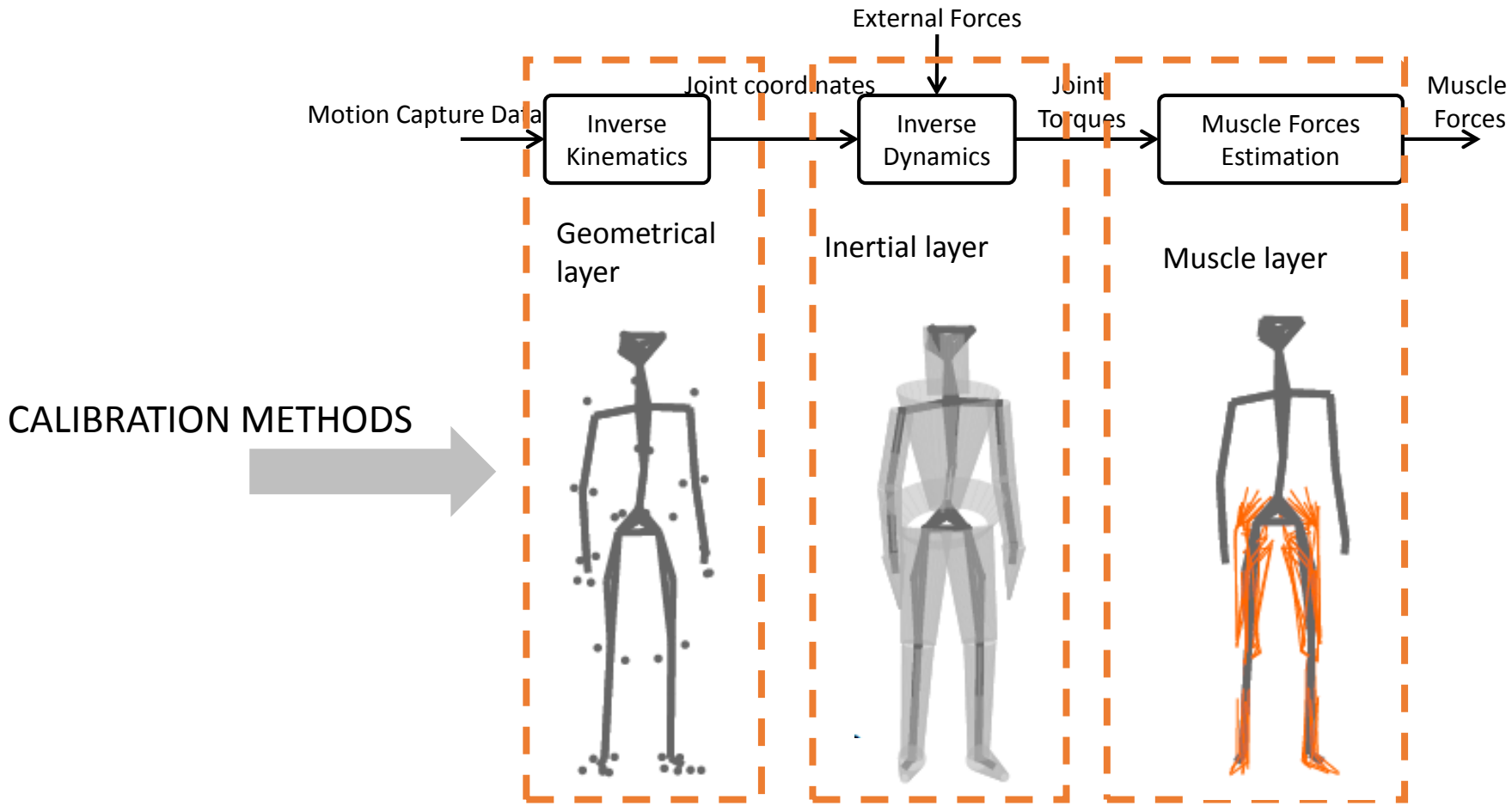
$$\min f(F) \xrightarrow{\text{classically}} f(F) = \sum_n \left(\frac{F_{m_i}}{F_{max_i}}\right)^p$$

s.t. $\vec{\tau} = \sum_i \overrightarrow{F_{m_i}} \times \overrightarrow{R_{m_i}}$

$F_{min_i} < F_{m_i} < F_{max_i}$

The more p is high, the more muscles acts in synergy
The more p is low, the more powerful muscles are preferably activated

Muscle layer



- Classically optimization under constraints

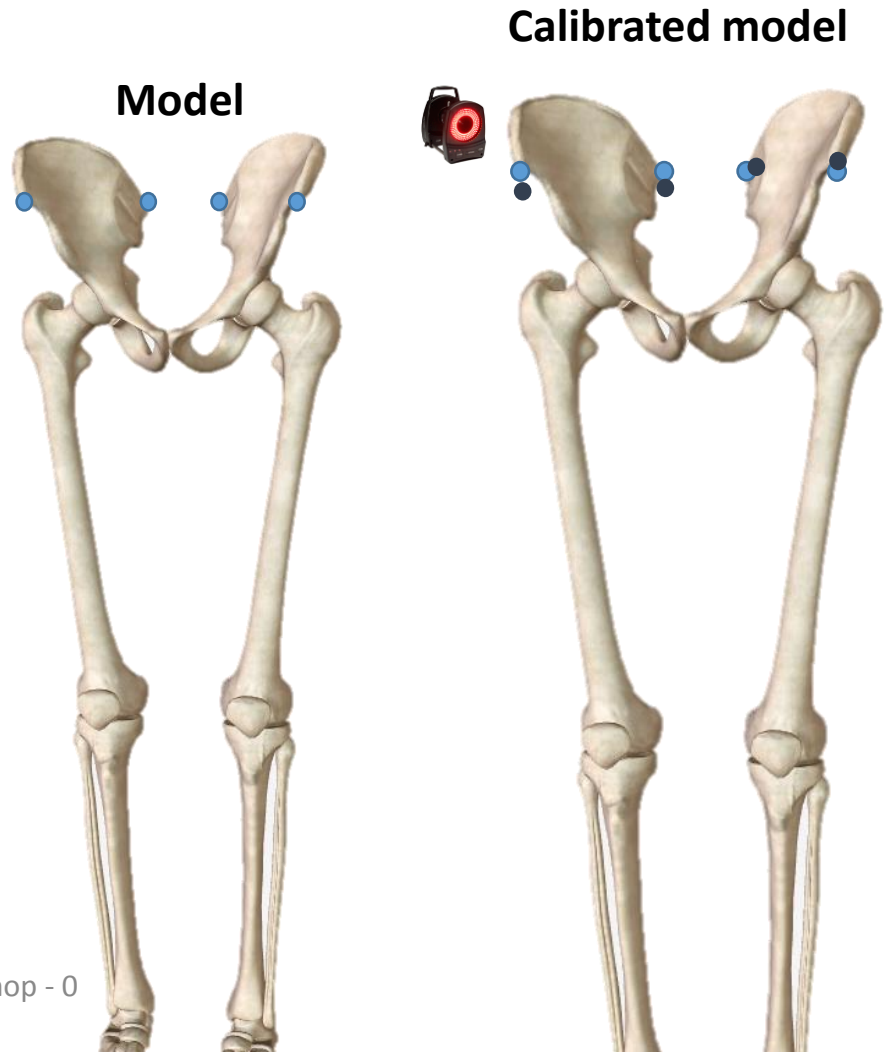
Geometrical calibration

Init: Anthropometrics tables [Winter 1955]
Marker-based [AnyBody, OpenSim]

Functional Optimization

$$\min_{k,\Delta p} \sum_{f}^{N_f} \sum_{m}^{N_m} ||\{X_{exp,m}(t_f)\} - \{X_{mod,m}^{global}(q(t_f), k, \Delta p)\}||_2^2$$

[Puchaud2018, Muller2015]



Inertial calibration

Init: anthropometrics tables [Dumas 2007, De Leva 1994]

Optimisation

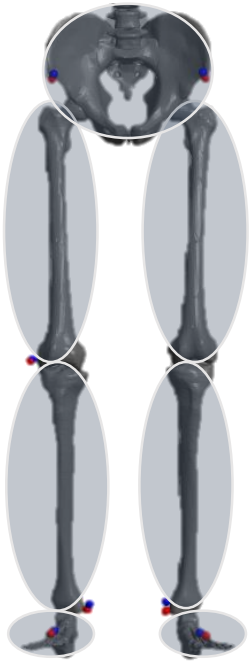
A. Muller 2017

Minimizing dynamics residuals

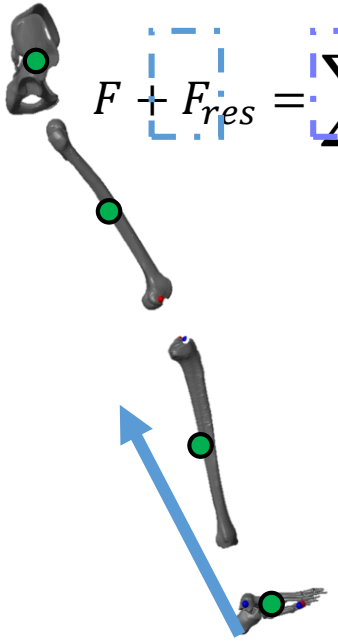
[Muller2017]

$$\min_{p} \sum_{f}^{N_f} \sum_{i}^{6} F_{res,i}(t_f)$$

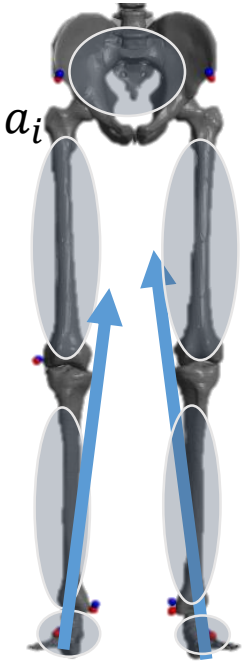
Model



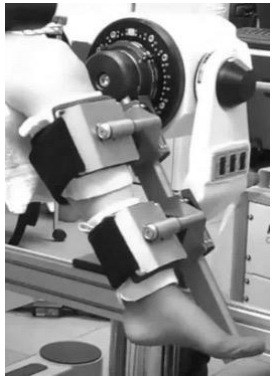
Inverse Dynamics



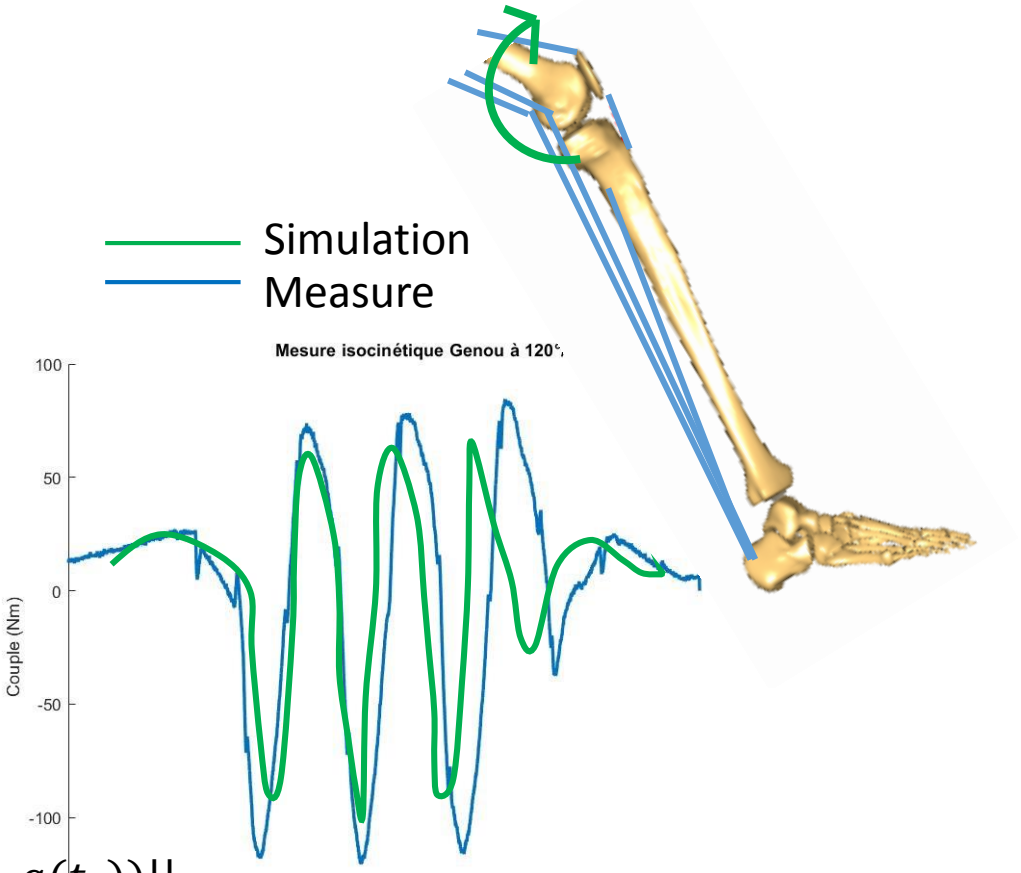
Calibrated model



[Muller2017, Haering 2017]



Optimization



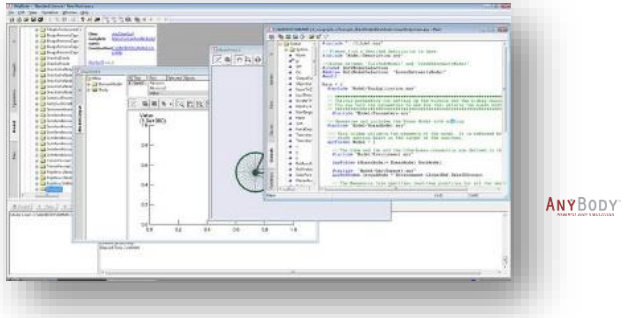
Finding muscular parameters taht makes the resulting torque fitting the experimental data

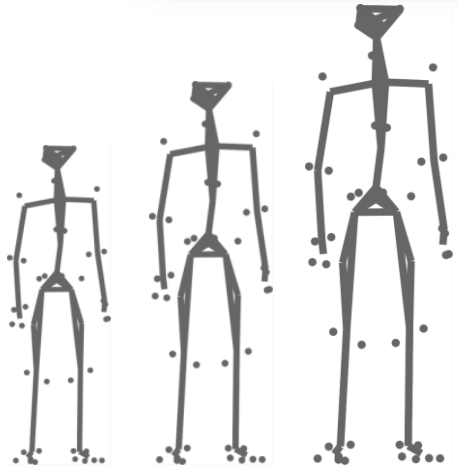
$$\min_{P_{muscle}} \sum_{f}^{N_f} \sum_{i}^{3} ||C_{exp,i}(t_f) - C_{sim,i}(q(t_f), a(t_f))||$$

Mains issues

- Computational cost (optimization)
- Editing, assembling models
- Subject specific models
- Running multiple simulations

Heavy, expert software





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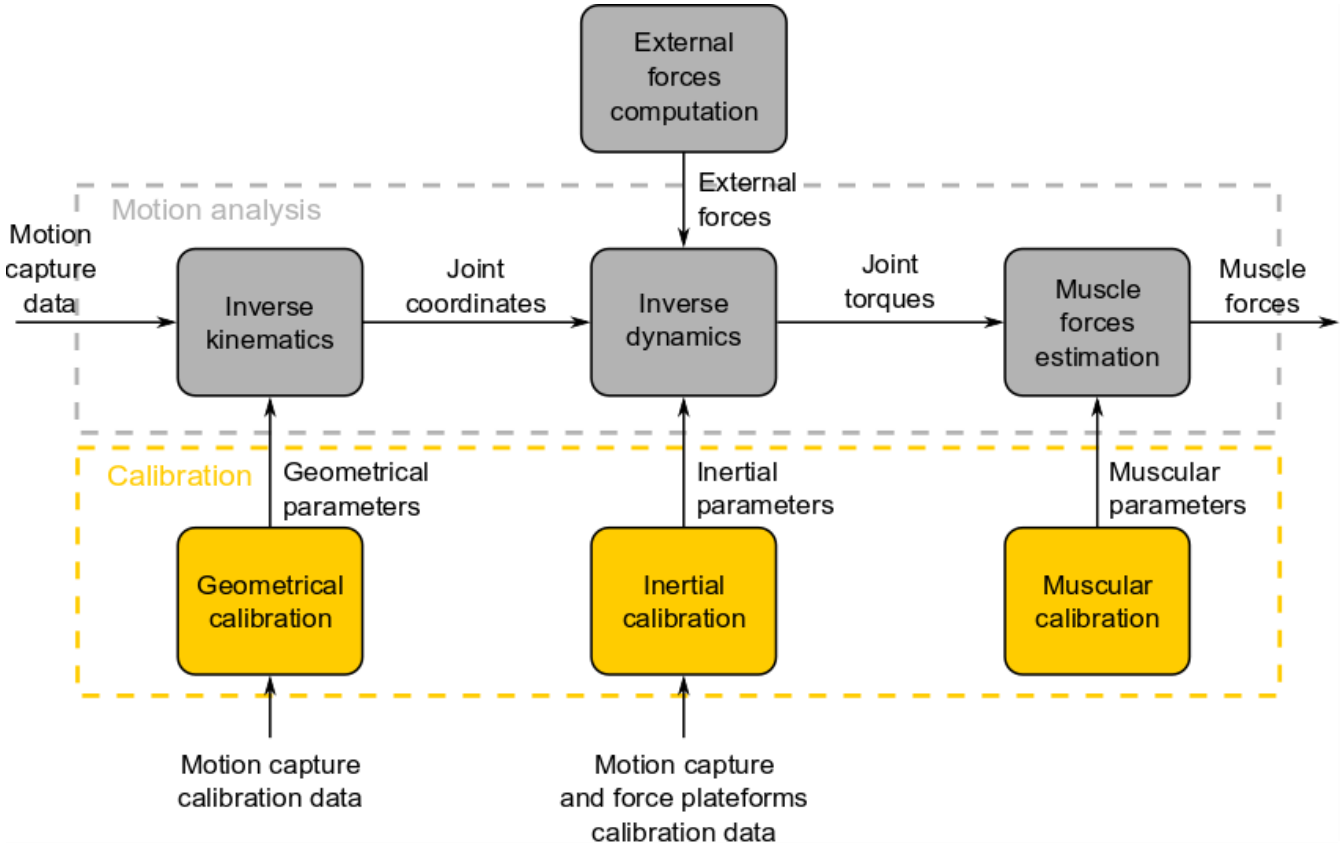
Introduction to CusToM

Charles Pontonnier, Pierre Puchaud

CusToM



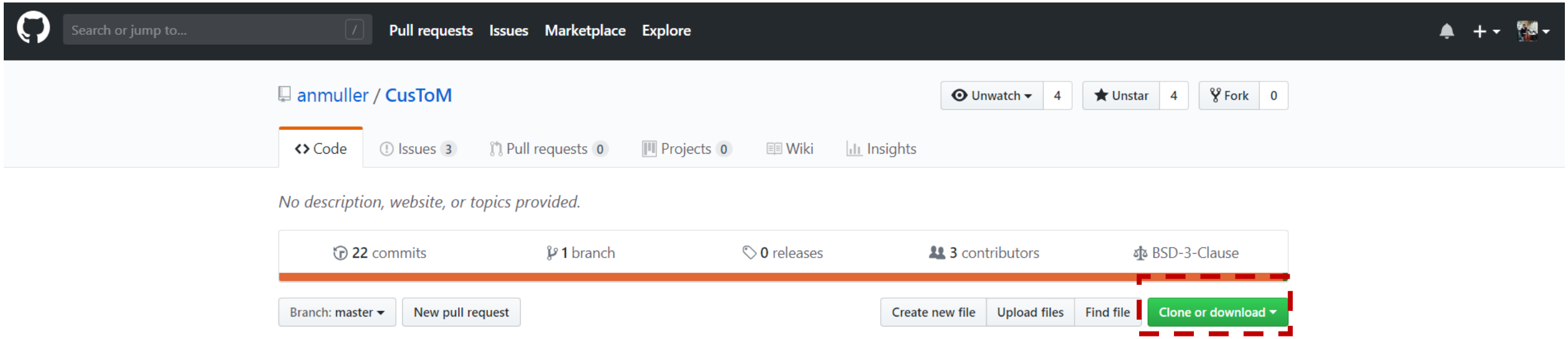
https://github.com/anmuller/CusToM

Please cite:
Muller, A., Pontonnier, C., Puchaud, P., Dumont, G., (2018). **CusToM : *a Matlab toolbox for musculoskeletal simulation*, in review.** Journal of Open Source Software.

Run on Matlab ®

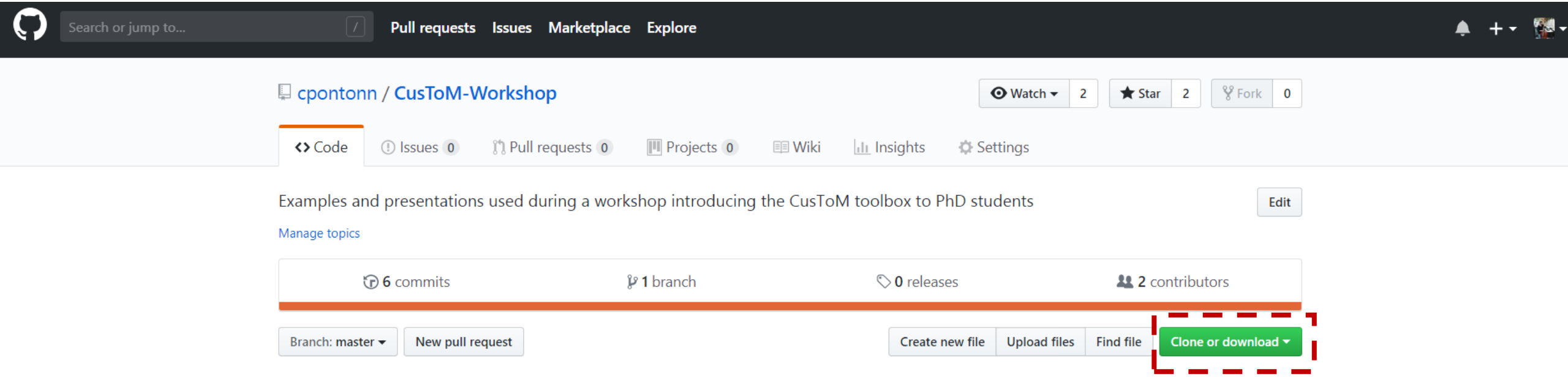
CusToM

https://github.com/anmuller/CusToM



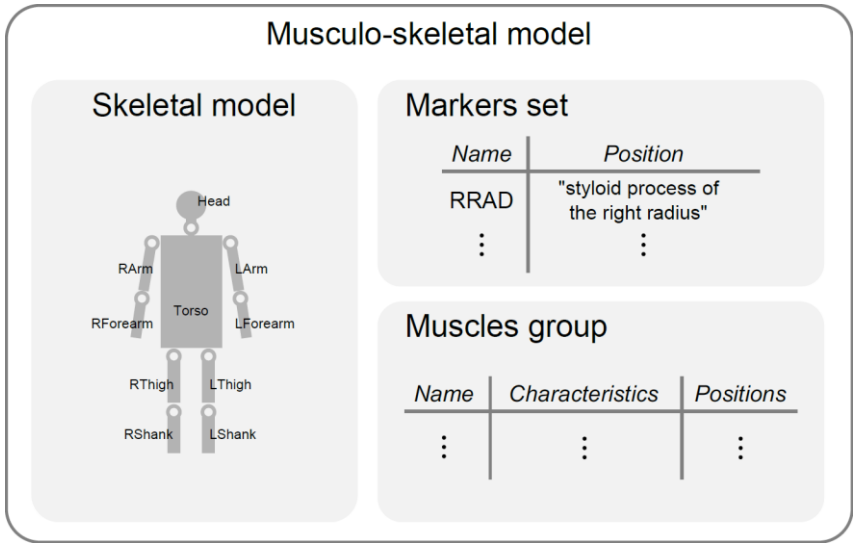
And for the workshop…

https://github.com/cpontonn/CusToM-Workshop

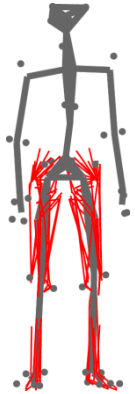


Until you're done with CusToM

- **Modularity**



- **Model calibration**

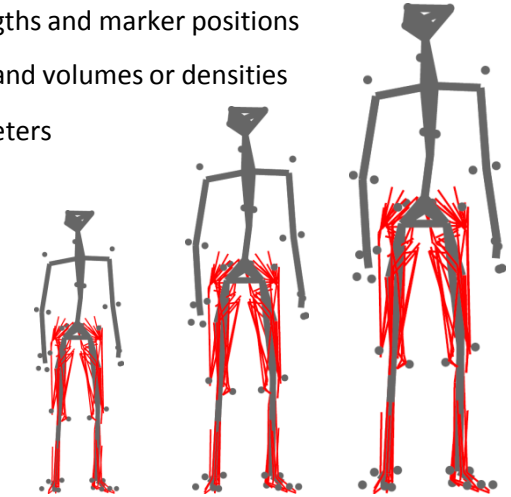


Geometrical: segment lengths and marker positions

Inertial: segment masses and volumes or densities

Muscular: Muscle parameters

Use of motion capture; external forces, isokinetic ergometer data



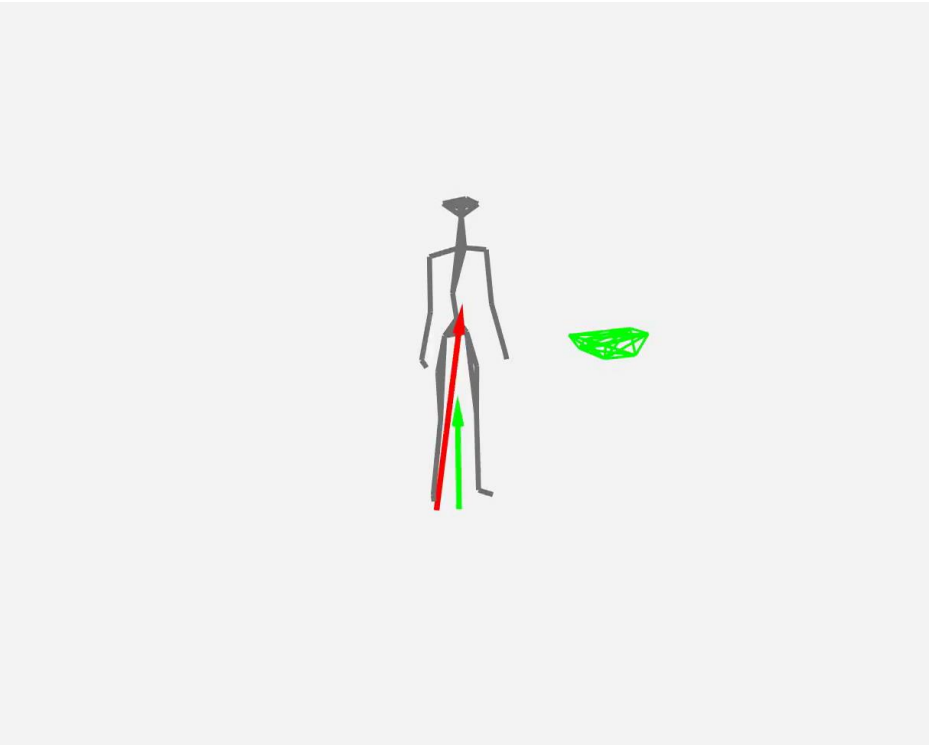
Force prediction

Contact forces prediction

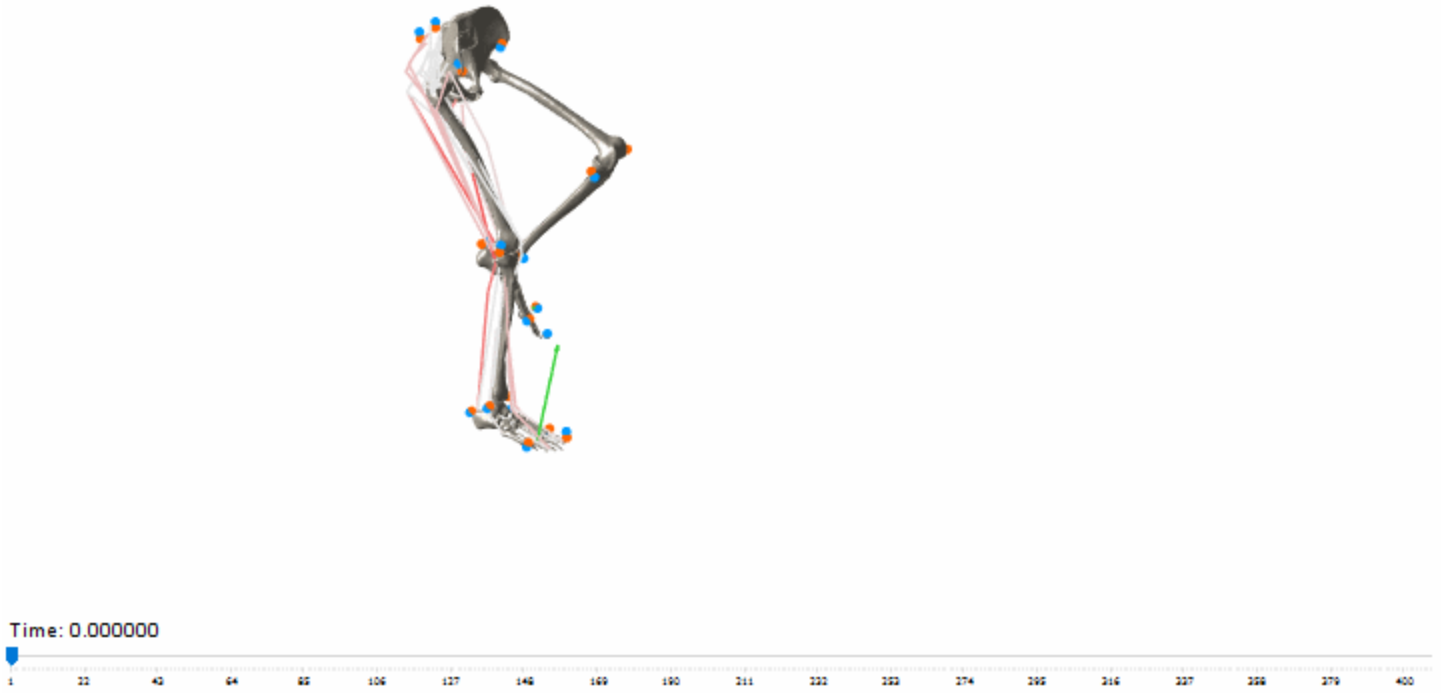
$$\min_{(\boldsymbol{\alpha},\boldsymbol{\beta},\boldsymbol{\gamma})} \quad \sum_{i=1}^{2N_f} \left(\alpha_i^2 + \beta_i^2 + \gamma_i^2\right)$$

$$\text{t.q.} \quad \boldsymbol{M_s}(\boldsymbol{q})\ddot{\boldsymbol{q}} + \boldsymbol{C_s}(\boldsymbol{q},\dot{\boldsymbol{q}}) + \boldsymbol{G_s}(\boldsymbol{q}) + \boldsymbol{\lambda_s} + \boldsymbol{E_s} = \boldsymbol{0};$$

$$\forall i \in [\![1, 2N_f]\!], (\alpha_i, \beta_i, \gamma_i) \in [-1, 1]^3$$



Vizualisation (thanks to GIBBON)



Now, let’s learn how it works

Lesson #1: Kinematics and geometrical calibration
Lesson #2: Muscle forces estimation
Lesson #3: Force prediction and post-processing
Lesson #4: XSENS handling and model edition

References (work in and with CusToM)

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