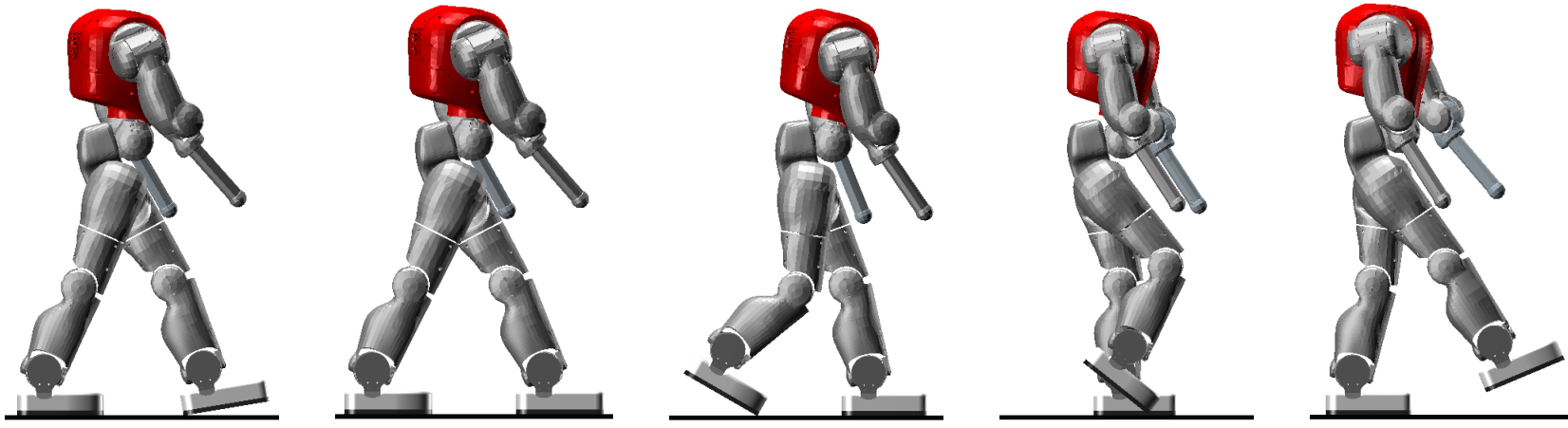


## Motivation & Proposed solution

Bio-inspired controllers are emerging as a promising way to implement dynamic walking. In this study, we implemented the one proposed by Geyer & Herr (2010), relying on reflex-controlled virtual **Hill muscles**. In this model, muscles' state is determined by the length ( $l_{ce}$ ) of their active, contractile element. However, its update rate is governed by a stiff and strongly non-linear state equation, thus requiring a **small integration time step**.



The contributions of this study are:

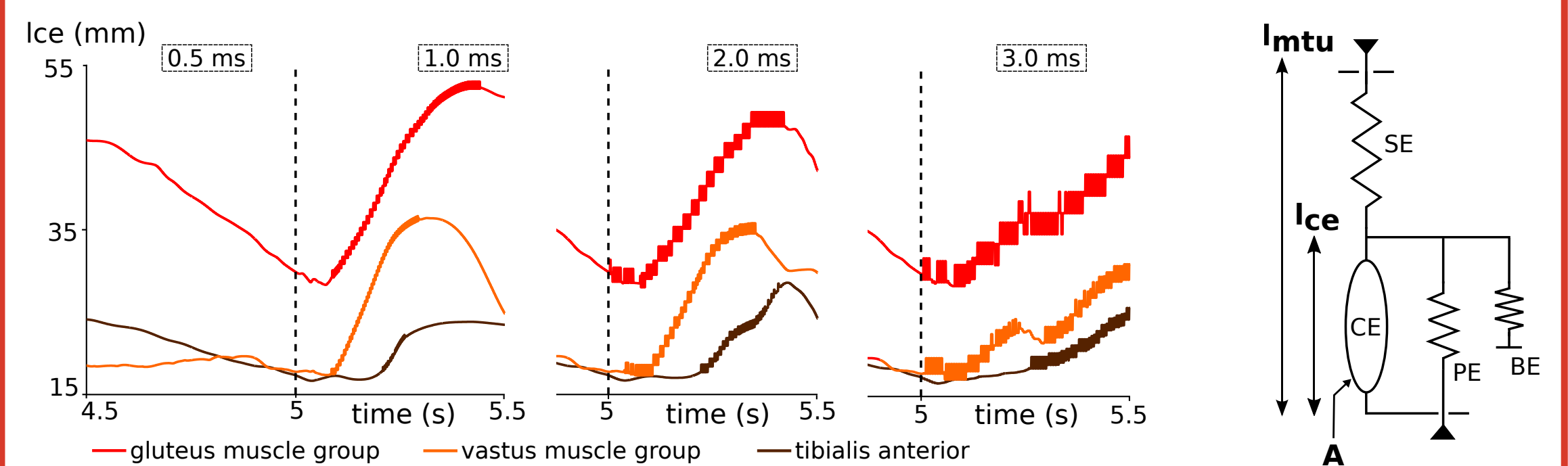
- the presentation of 3 methods to get the **steady-state** value of  $l_{ce}$
- the study of their accuracy and computational cost
- the illustration that the  $l_{ce}$  dynamics can be neglected for **fast muscles**
- a method to mix the steady-state approximation with the full muscle dynamic model for **slow muscles**

## Time step numerical issue

The length  $l_{ce}$  is updated with a time integration of its derivative  $\dot{l}_{ce} = f(l_{ce}, l_{mtu}, A)$ .

- $l_{mtu}$ : total muscle-tendon unit length
- $A$ : activation provided by the motor neuron

Due to the stiffness and non-linearity of  $f(\cdot)$ , exceeding a **critical time step** value generates **numerical oscillations**.

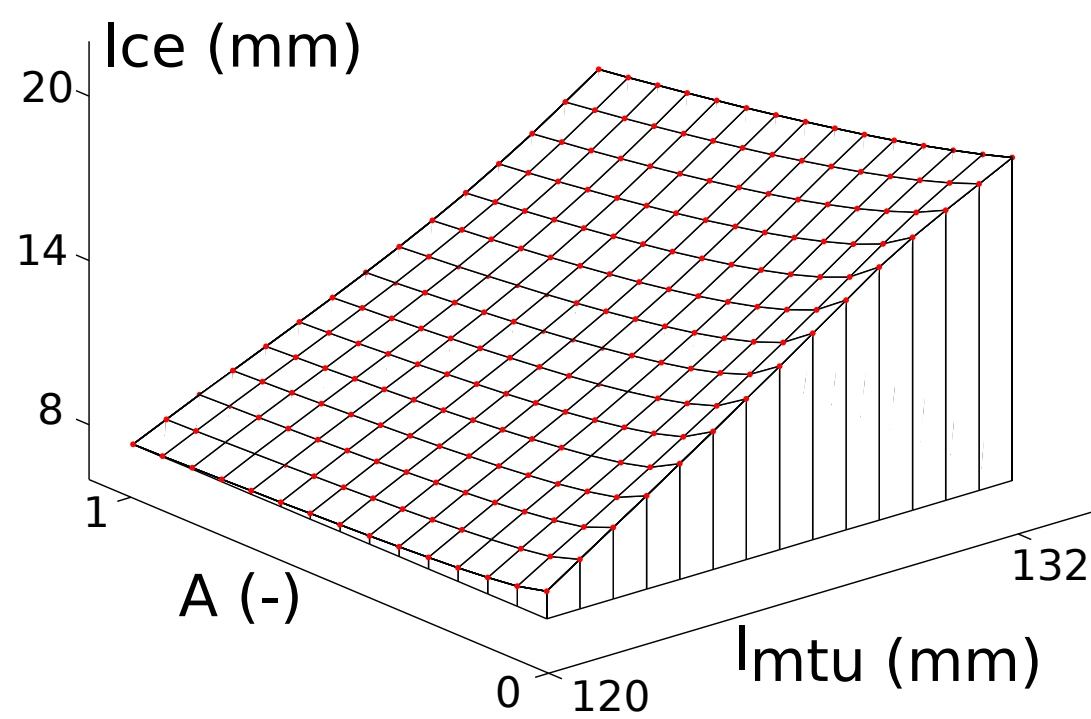


Control rules are **impacted by**  $l_{ce}$  in the feedback loop, causing the walker to fall.

## Steady-state computation - Three methods to get $l_{ce}$ without integration

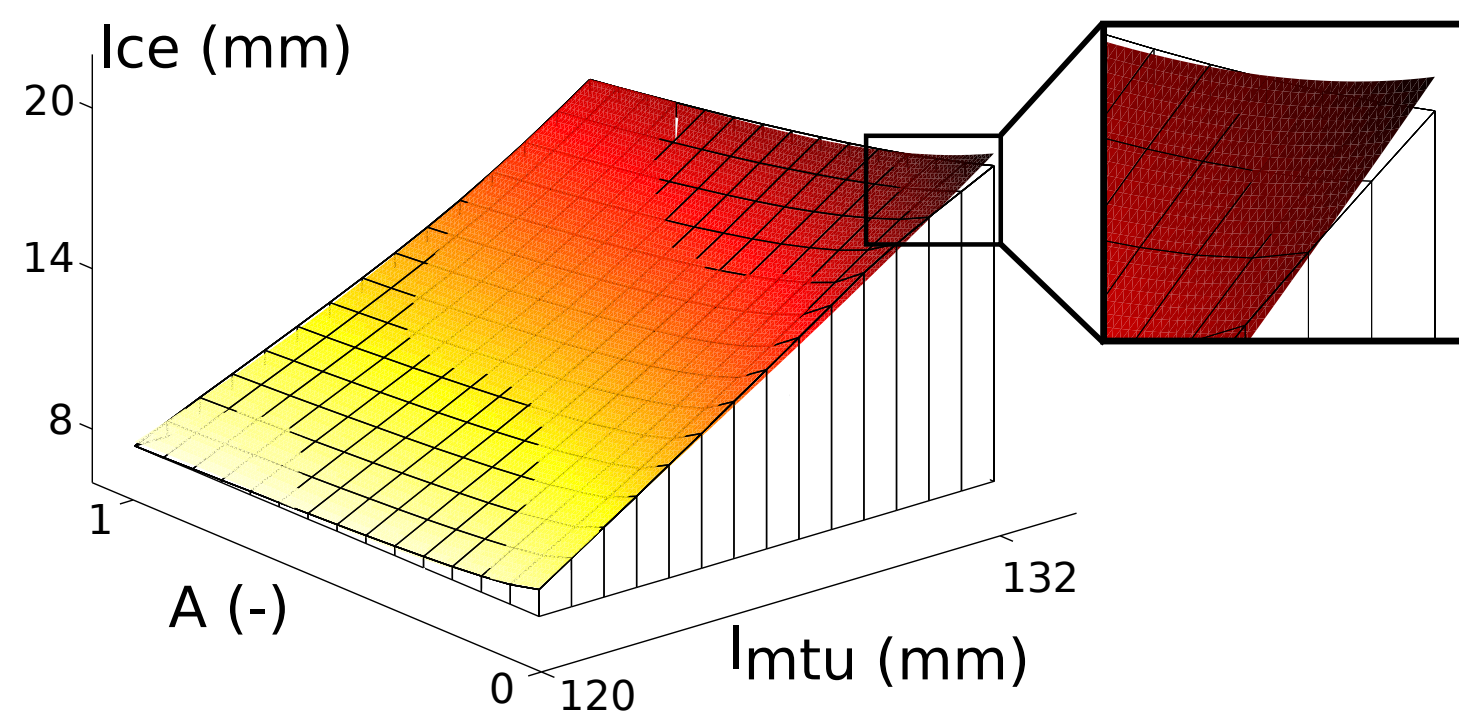
Neglecting the muscle dynamics:  $l_{ce}$  is always at steady-state, i.e.  $\dot{l}_{ce} = 0$ . Then, the problem is to solve  $f(l_{ce}, l_{mtu}, A) = 0$ , i.e. to find  $l_{ce} = g(l_{mtu}, A)$ .

### Look Up Table (LUT) interpolation



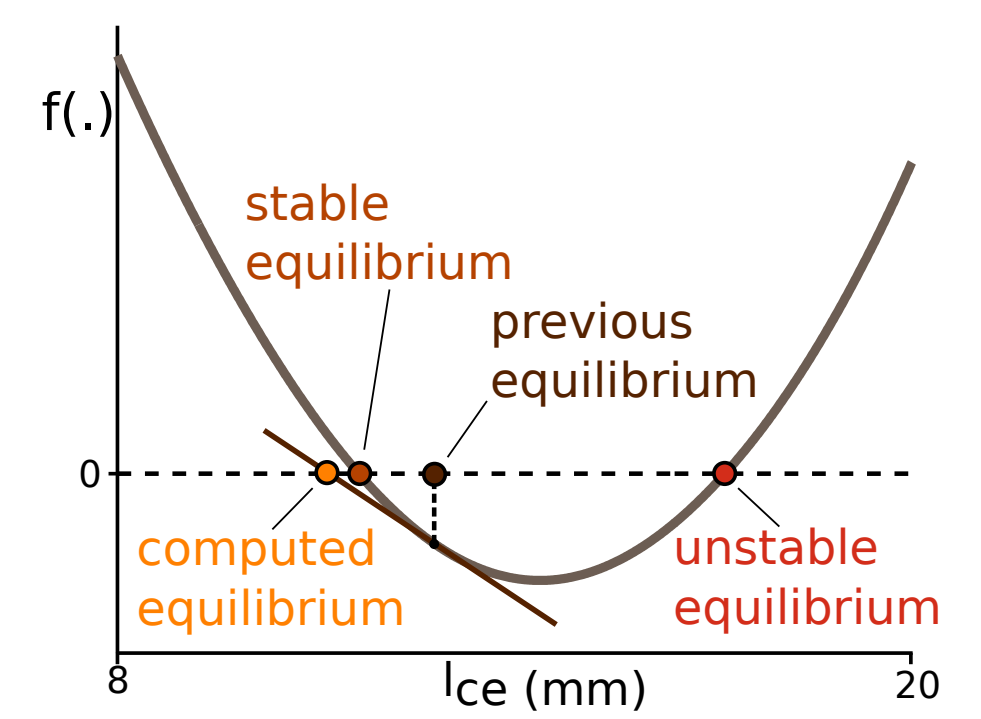
- generated off-line
- cost depends on the mesh refinement
- accuracy depends on the mesh refinement

### Third-order polynomial (TOP) approximation



- generated off-line
- computationally efficient
- accuracy depends on the LUT to fit

### Newton-Raphson scheme (NRS)

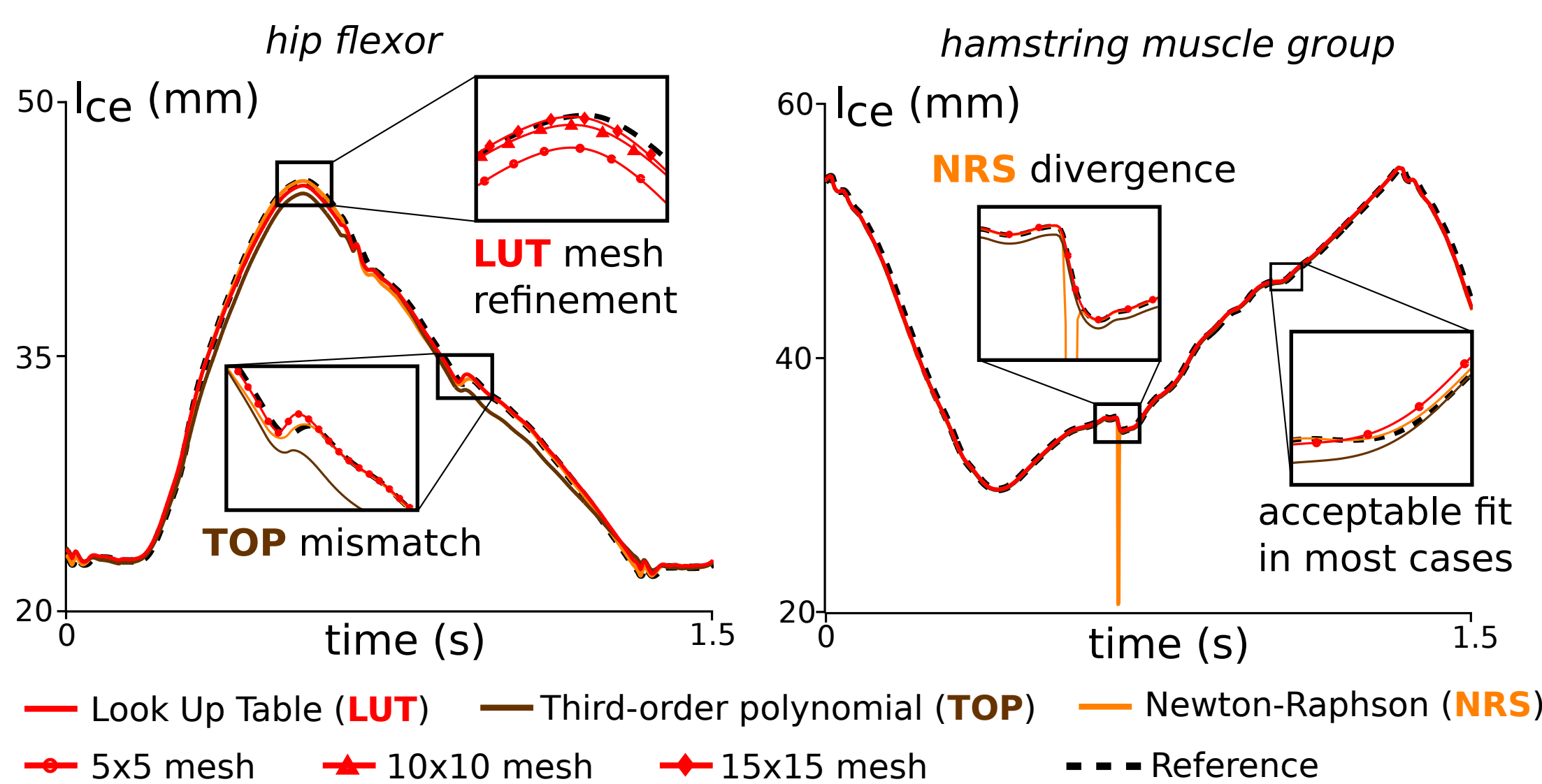


- no pre-process computation
- convergence to a stable equilibrium not guaranteed
- more than one iteration might be needed

## Results

We compare the three steady-state  $l_{ce}$  profiles with a reference based on the muscle dynamics.

- **LUT**: refining the mesh increases the accuracy.
- **TOP**: accuracy deteriorated for some muscles at some phases of the gait.
- **NRS**: excellent fit with only one iteration, except when the gradient of  $f(\cdot)$  is close to 0.



The **LUT** approximation **never diverges** and is **accurate**, provided a good mesh refinement.

We replaced the full dynamics model by these steady-state approximations.

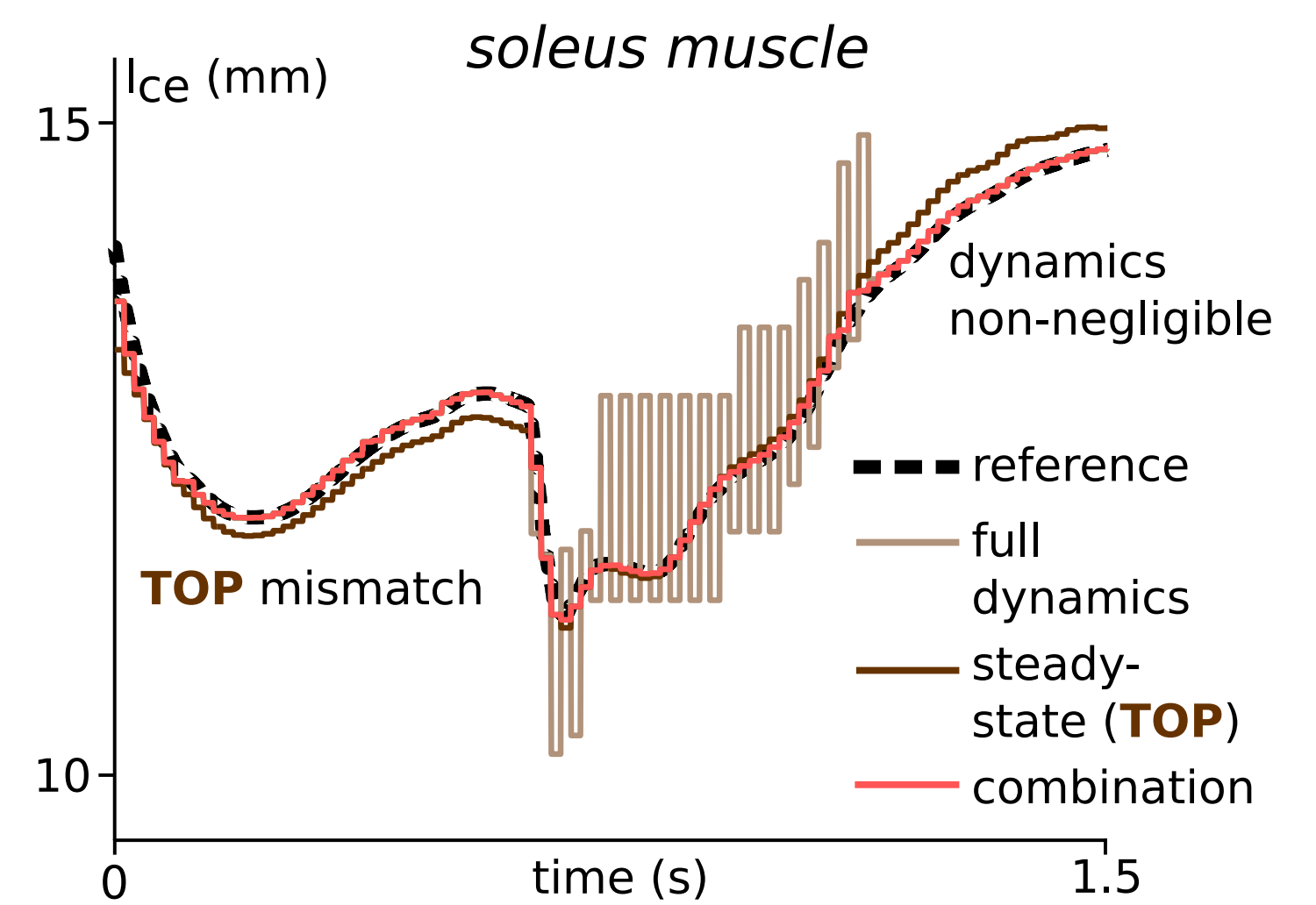
- Any of these approximations preserved the dynamic walking gait.
- However, the pattern was more jerky than with the original model.
- Re-optimizing the controller led to retrieve natural and smooth gaits.

Controllers were able to cope with **3 ms time steps**, instead of **0.5 ms before**.

## Slow dynamics muscles case

At some phases of the gait, the **soleus muscle dynamics cannot be neglected**, i.e. all steady-state approximations diverge from the actual  $l_{ce}$  value.

- Reference computed with a 0.5 ms time step, the others with a 10 ms one.
- Neither using the full dynamics model nor a steady-state approximation provided a correct fit.
- Combining these two signals led to a correct fit. Combination is done by saturating the  $l_{ce}$  determined by the full dynamics to the steady-state value.



This **signal combination** highly improves the accuracy for **slow dynamics** muscles computed with **large time steps**.

## Acknowledgment & Reference

This work is supported by FP7 under Grant 611832 (WALK-MAN), by UCLouvain (IMMC11/13) and by F.R.S.-FNRS (Ph.D. scholarship to N. Van der Noot).

H. Geyer and H. Herr. A muscle-reflex model that encodes principles of legged mechanics produces human walking dynamics and muscle activities. *Neural Systems and Rehabilitation Engineering, IEEE Transactions on*, 18(3):263 - 273, 2010.