

# Optimal control of multibody-systems

## Introduction

Optimal control theory

- motivated by the preservation of resources & time
- leads to temporal boundary value problems

Wide range of applications

- trajectory optimization in robotics
- aerospace engineering (e.g. orbit transfers, ascent & descent trajectories)
- process engineering (e.g. chemical processes)

Our focus is set on multibody-systems (MBS), thus we deal with

- joint formulations & control input interfaces
- optimal trajectory planning for least control-effort maneuvers
- application of energy-momentum (EM) schemes

## Dynamics

Nonsingular *direction-cosine* description of rigid body rotations

Configuration  $\mathcal{B}_t$

$$\bar{\mathbf{x}}(\mathcal{X}, t) = \varphi(t) + \mathbf{R}(t)\mathcal{X}$$

with rotation matrix

$$\mathbf{R} \in \{SO(3) | \mathbf{R} = \mathbf{d}_i \otimes \mathbf{e}_i\}$$

Internal constraints

$$\Phi_b^{\text{rigid}} = \mathbf{d}_i \cdot \mathbf{d}_j - \delta_{ij}$$

Discrete state equations

$$\mathbf{g}_n^h = \begin{bmatrix} (\mathbf{q}_{(n)} - \mathbf{q}_{(n-1)}) - \Delta t \mathbf{v}_{(n-1/2)} \\ \mathbf{M}(\mathbf{v}_{(n)} - \mathbf{v}_{(n-1)}) + \Delta t \left[ \sum_b \nabla_{\mathbf{q}} \Phi_{b,(n-1/2)} \bar{\lambda}_{b,(n)} - \mathbf{f}(\mathbf{q}_{(n)}, \mathbf{q}_{(n-1)}, \bar{\mathbf{u}}_{(n)}) \right] \\ \Phi_{(n)} \end{bmatrix}$$

This advantageous structure enables

- an object-oriented assembly framework for large MBS.
- the intuitive formulation of joints.
- the application of EM-Schemes.

## Optimal Control

Discretization of the time domain  $\Omega = [0, T]$

$$\Omega = \bigcup_n^N \Omega_n, \quad \Omega_n = [t_{n-1}, t_n]$$

Discrete augmented cost functional & running cost

$$J^h = \eta \cdot \Psi|_T + \sum_{n=1}^N \left[ \mathcal{J}^h(\bar{\mathbf{u}}_n) + \mu_n \cdot \mathbf{g}_n^h \right] \Delta t, \quad \mathcal{J}^h = \frac{1}{2} \bar{\mathbf{u}}_n \cdot \bar{\mathbf{u}}_n$$

Discrete necessary conditions of optimality (DNCO)

$$\delta J^h = \delta \mathbf{p} \cdot \left( \nabla_{\mathbf{p}} \tilde{\mathcal{J}}^h + (\nabla_{\mathbf{p}} \mathbf{c})^T \boldsymbol{\nu} \right) + \delta \boldsymbol{\nu} \cdot \mathbf{c}$$

with

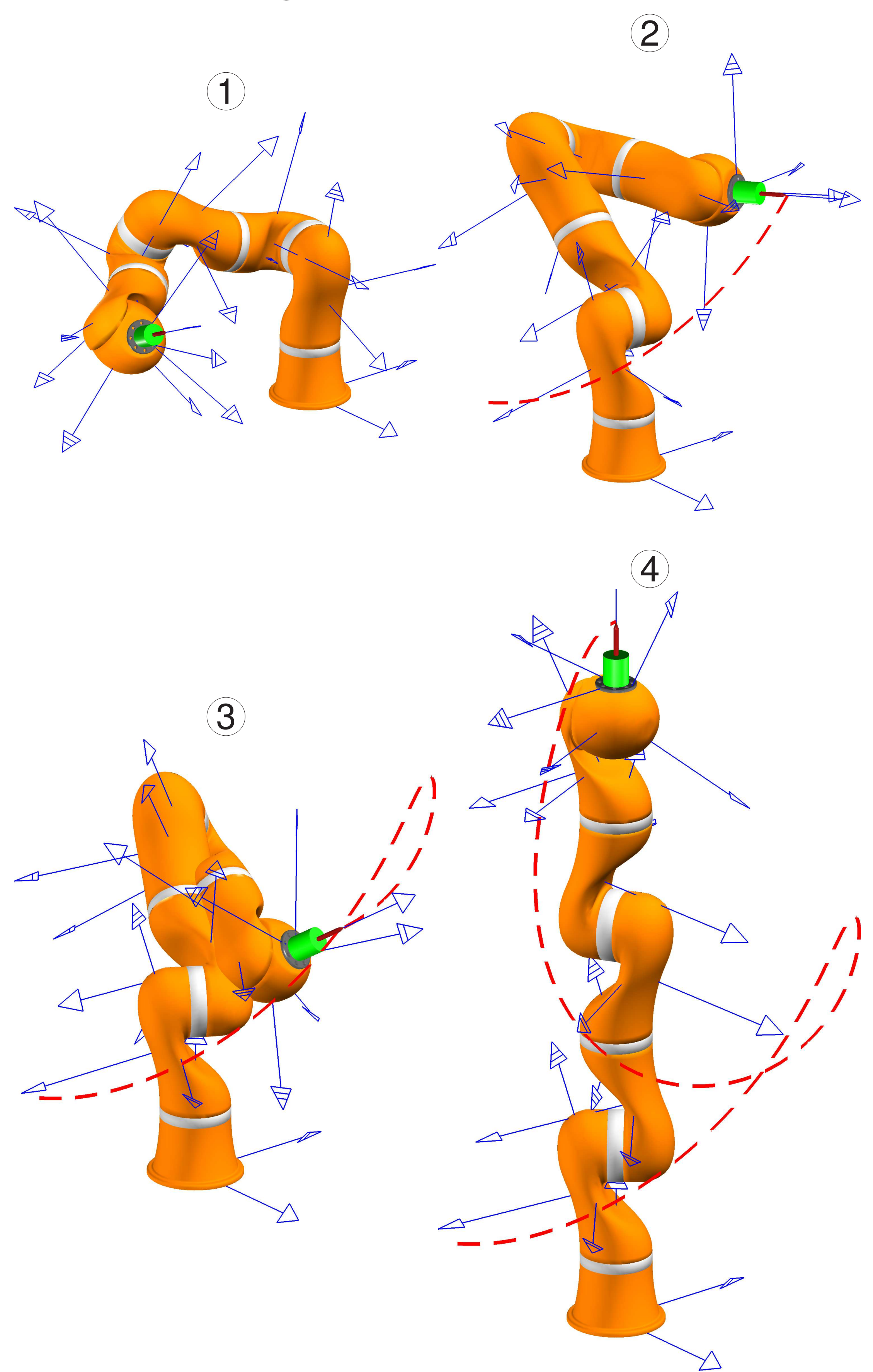
$$\mathbf{c} = \{ \mathbf{g}_1^h, \dots, \mathbf{g}_n^h, \dots, \tilde{\mathbf{g}}_N, \Psi \}$$

$$\boldsymbol{\nu} = \{ \mu_1, \dots, \mu_n, \dots, \tilde{\mu}_N, \eta \}$$

$$\mathbf{p} = \{ \mathbf{x}_1, \bar{\mathbf{u}}_1, \dots, \mathbf{x}_n, \bar{\mathbf{u}}_n, \dots, \tilde{\mathbf{x}}_N, \bar{\mathbf{u}}_N \}, \quad \mathbf{x}_n = \{ \mathbf{q}_n, \mathbf{v}_n, \bar{\lambda}_n \}$$

## Numerical example

We consider the following MBS of a fully actuated 7-axis manipulator. The initial and terminal state ① & ④ of the MBS as well as the provided time of 10s to occupy state ④ are predefined. The time domain  $\Omega$  has been subdivided into  $N = 160$  equidistant subdomains. The solution of the DNCO provide the illustrated candidate solution for a least-control-effort maneuver, together with the required servo-torques to achieve this trajectory. The solution features a trajectory, where the manipulator makes perfect use of its inertia to reach the final state within the given time.



Optimal Trajectory (Least Control Effort)

The servo-torques as well as the delta net energy  $\Delta E_{\text{net}}^{(n,n+1)} = E_{\text{net}}^{n+1} - E_{\text{net}}^n$  are shown below.

