



Lecture «Robot Dynamics»: Kinematic Control

151-0851-00 V

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|-----------|----------------|---|
| lecture: | HG F3 | Tuesday 10:15 – 12:00, every week |
| exercise: | HG D7.1 | Wednesday 8:15 – 10:00, according to schedule |

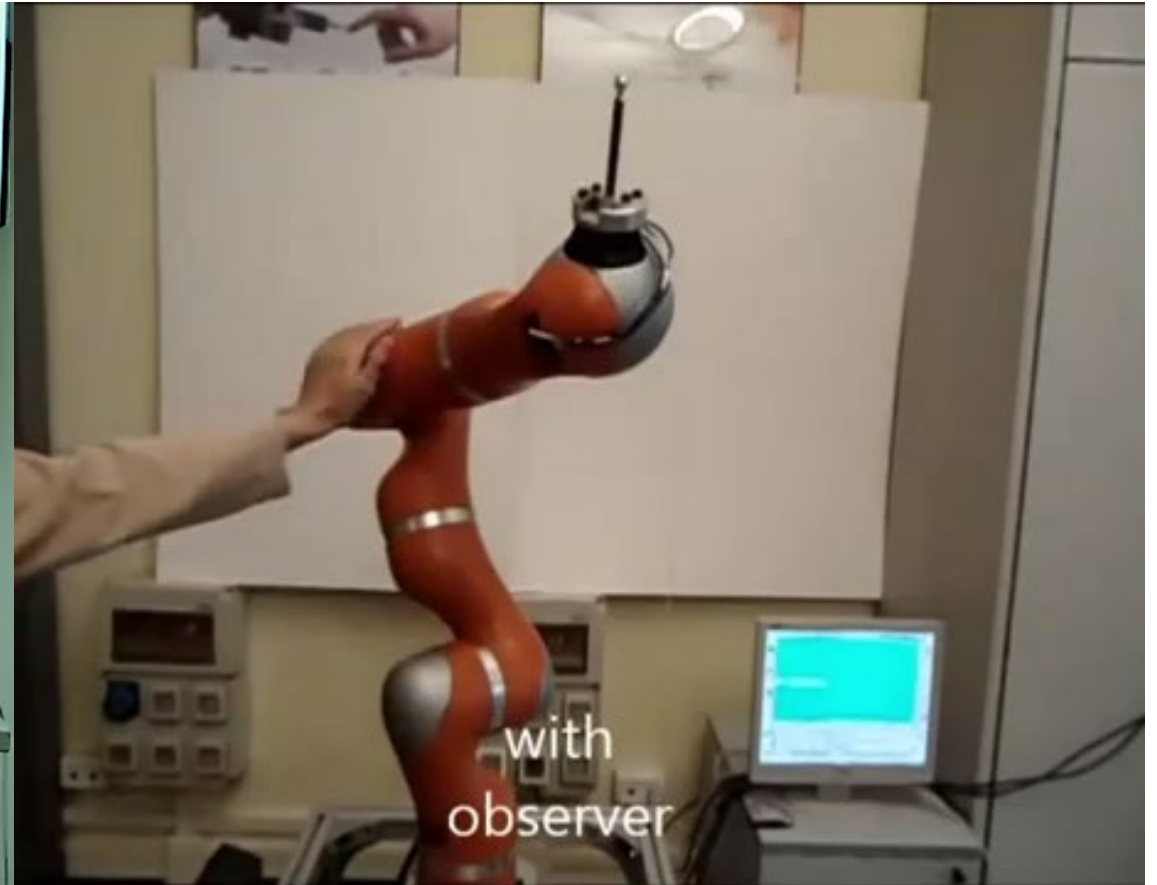
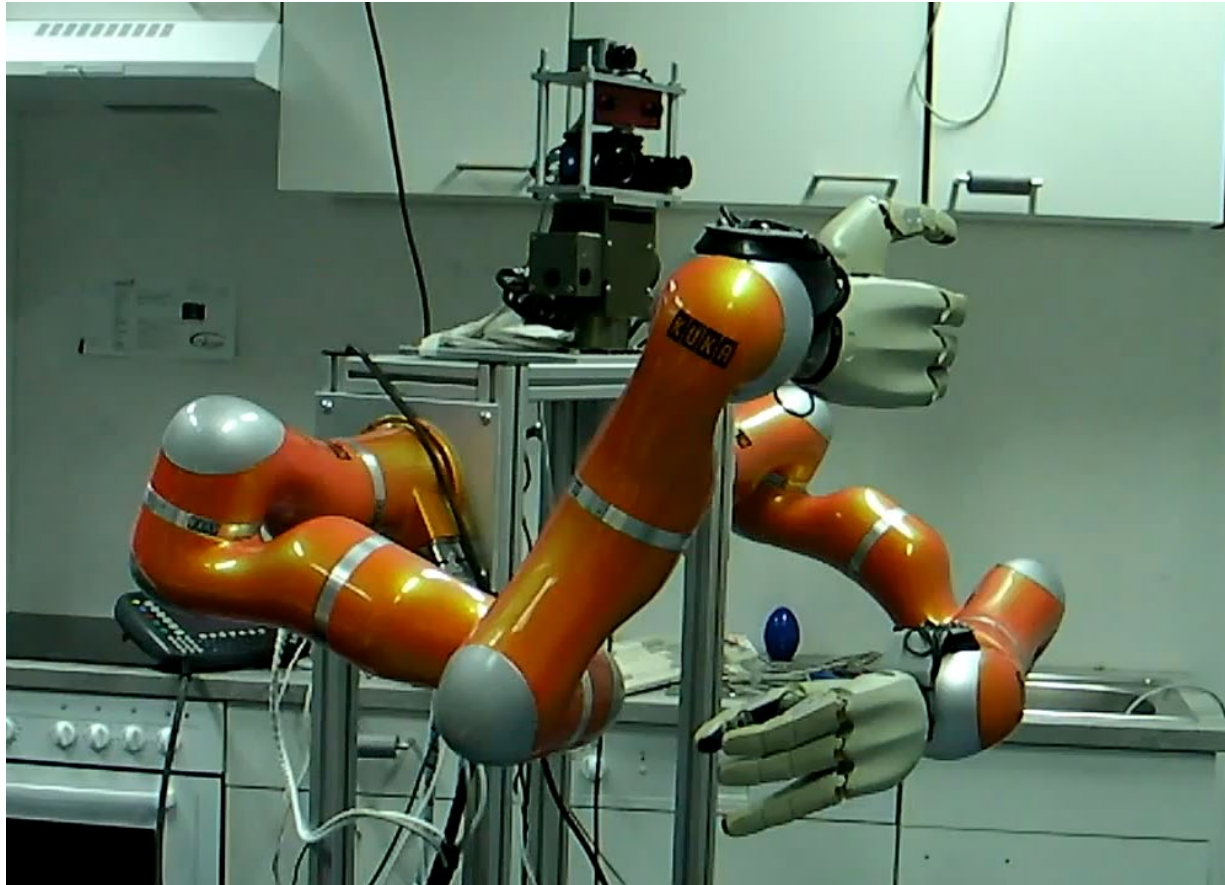
Marco Hutter, Roland Siegwart, and Thomas Stastny

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|------------|---------------------|--|------------|--|--|
| 17.09.2019 | Intro and Outline | Course Introduction; Recapitulation Position, Linear Velocity | | | |
| 24.09.2019 | Kinematics 1 | Rotation and Angular Velocity; Rigid Body Formulation, Transformation | 25.09.2019 | Exercise 1a | Kinematics Modeling the ABB arm |
| 01.10.2019 | Kinematics 2 | Kinematics of Systems of Bodies; Jacobians | 02.10.2019 | Exercise 1a | Differential Kinematics of the ABB arm |
| 08.10.2019 | Kinematics 3 | Kinematic Control Methods: Inverse Differential Kinematics, Inverse Kinematics; Rotation Error; Multi-task Control | 09.10.2019 | Exercise 1b | Kinematic Control of the ABB Arm |
| 15.10.2019 | Dynamics L1 | Multi-body Dynamics | 16.10.2019 | Midterm 1 | Programming kinematics with matlab |
| 22.10.2019 | Dynamics L2 | Floating Base Dynamics | 23.10.2019 | Exercise 2a | Dynamic Modeling of the ABB Arm |
| 29.10.2019 | Dynamics L3 | Dynamic Model Based Control Methods | 30.10.2019 | Exercise 2b | Dynamic Control Methods Applied to the ABB arm |
| 05.11.2019 | Legged Robot | Dynamic Modeling of Legged Robots & Control | 06.11.2019 | Midterm 2 | Programming dynamics with matlab |
| 12.11.2019 | Case Studies 1 | Legged Robotics Case Study | 13.11.2019 | Exercise 3 | Legged robot |
| 19.11.2019 | Rotorcraft | Dynamic Modeling of Rotorcraft & Control | 20.11.2019 | | |
| 26.11.2019 | Case Studies 2 | Rotor Craft Case Study | 27.11.2019 | Exercise 4 | Modeling and Control of Multicopter |
| 03.12.2019 | Fixed-wing | Dynamic Modeling of Fixed-wing & Control | 04.12.2019 | | |
| 10.12.2019 | Case Studies 3 | Fixed-wing Case Study (Solar-powered UAVs - AtlantikSolar, Vertical Take-off and Landing UAVs – Wingtra) | 11.12.2019 | Exercise 5 | Fixed-wing Control and Simulation |
| 17.12.2019 | Summery and Outlook | Summery; Wrap-up; Exam | | Robot Dynamics - Kinematic Control 8.10.2019 2 | |

Outline

- Kinematic control methods
 - Inverse kinematics
 - Singularities, redundancy
 - Multi-task control
 - Iterative inverse differential kinematics
 - Kinematic trajectory control

Null-space



Forward kinematics

- Forward kinematics

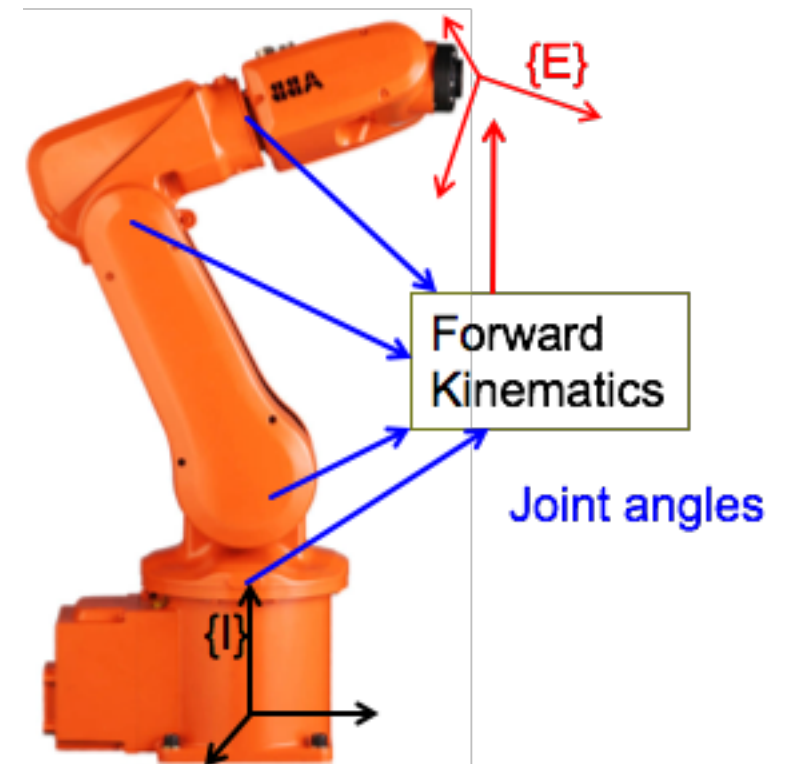
- Description of end-effector configuration (position & orientation) as a function of joint coordinates
- Use the homogeneous transformation matrix

- $$\mathbf{T}_{IE}(\mathbf{q}) = \begin{bmatrix} \mathbf{C}_{IE}(\mathbf{q}) & {}^I\mathbf{r}_{IE}(\mathbf{q}) \\ \mathbf{0}^T & 1 \end{bmatrix} \in \mathbb{R}^{4 \times 4}$$

- Parametrized description

$$\mathbf{x}_e = \begin{pmatrix} \mathbf{r}_e(\mathbf{q}) \\ \phi_e(\mathbf{q}) \end{pmatrix} = f(\mathbf{q})$$

$$\chi_e = \begin{pmatrix} \chi_{eP} \\ \chi_{eR} \end{pmatrix}$$



Inverse kinematics

- Inverse kinematics
 - Description of joint angles as a function of the end-effector configuration

- Use the homogeneous transformation matrix

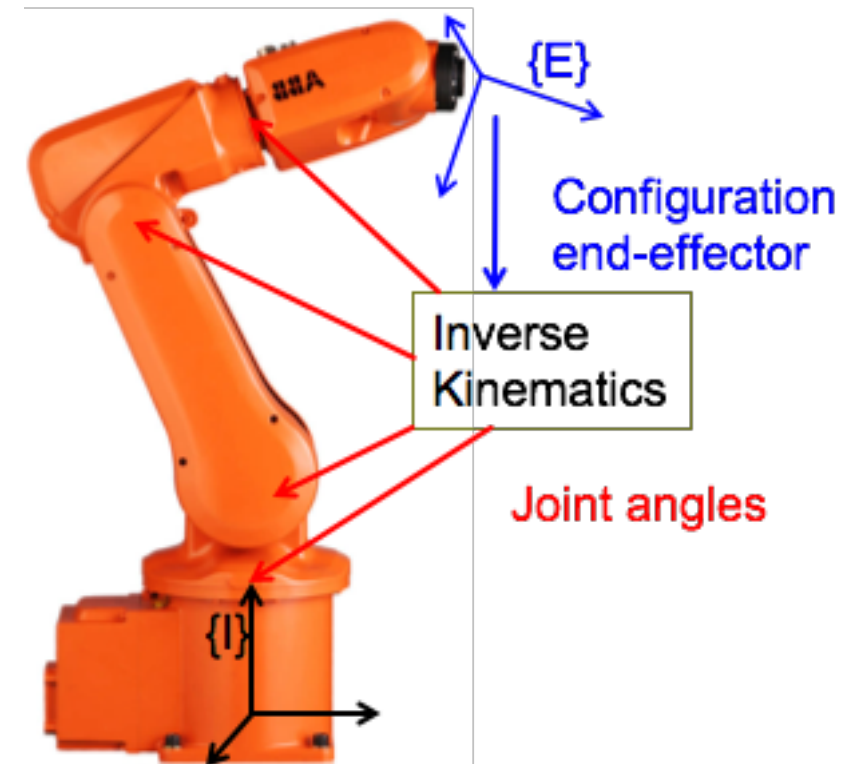
- $$\mathbf{T}_{IE}(\mathbf{q}) = \begin{bmatrix} \mathbf{C}_{IE}(\mathbf{q}) & \mathcal{I}\mathbf{r}_{IE}(\mathbf{q}) \\ \mathbf{0}^T & 1 \end{bmatrix} \in \mathbb{R}^{4 \times 4}$$

- Parametrized description

$$\mathbf{x}_e = \begin{pmatrix} \mathbf{r}_e(\mathbf{q}) \\ \phi_e(\mathbf{q}) \end{pmatrix} = f(\mathbf{q})$$

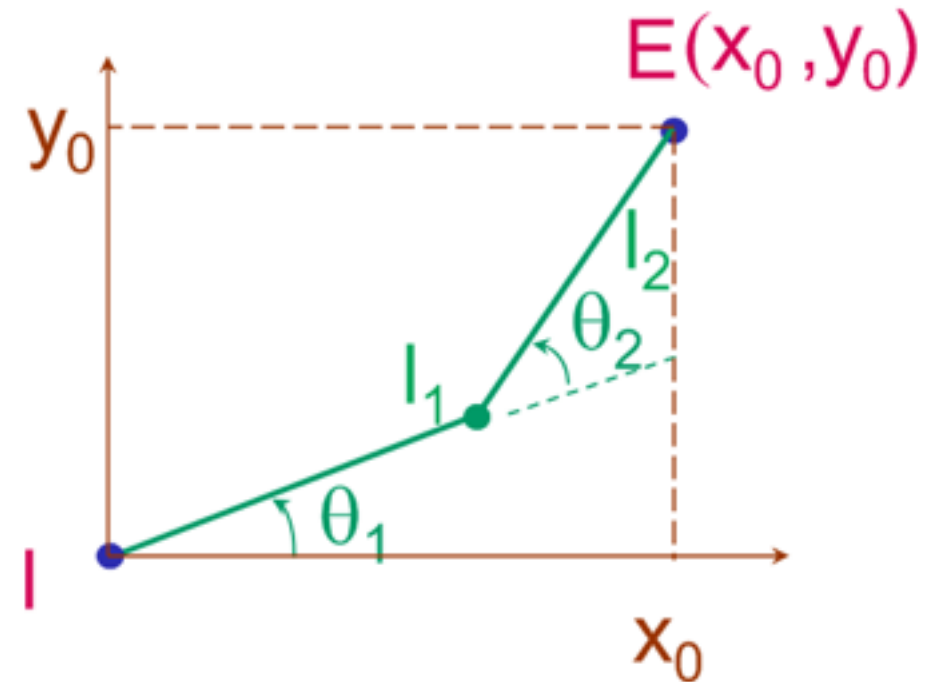
$$\mathbf{q} = \mathbf{f}^{-1}(\mathbf{x}_E)$$

$$\chi_e = \begin{pmatrix} \chi_{eP} \\ \chi_{eR} \end{pmatrix}$$



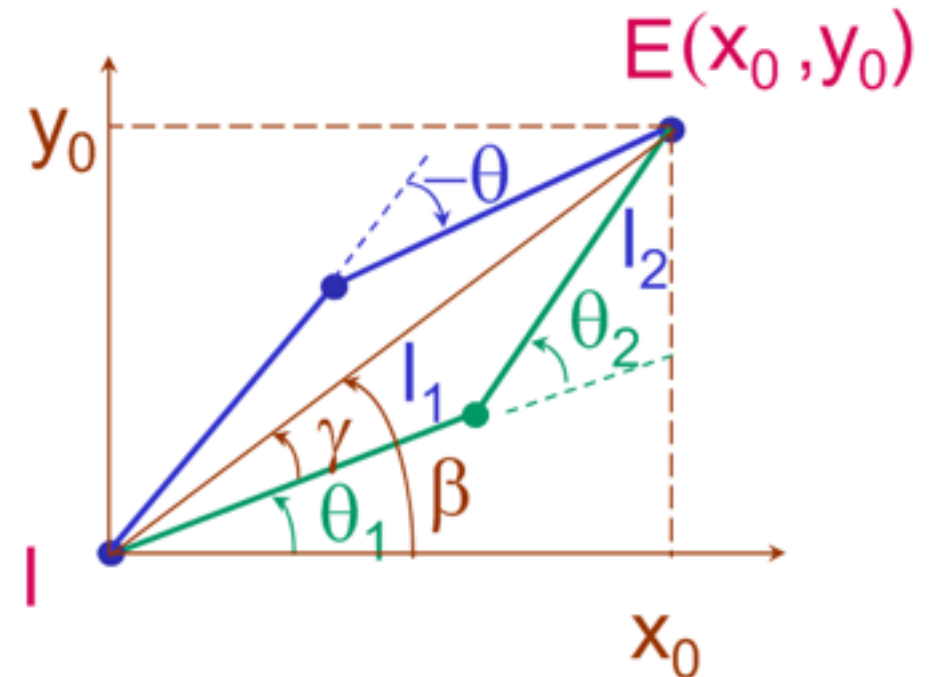
Closed form solutions

- Geometric or Algebra
 - Analytic solutions exist for a large class of mechanisms
 - 3 intersecting neighbouring axes (most industrial robots)



Closed form solutions

- Geometric or Algebraic
 - Analytic solutions exist for a large class of mechanisms
 - 3 intersecting neighbouring axes (most industrial robots)
- Geometric
 - Decompose spatial geometry of manipulator into several plane problems and apply geometric laws



Closed form solutions

- Geometric or Algebraic

- Analytic solutions exist for a large class of mechanisms
 - 3 intersecting neighbouring axes (most industrial robots)

- Geometric

- Decompose spatial geometry of manipulator into several plane problems and apply geometric laws

- Algebraic

- Manipulate transformation matrices to get the joint angles

$$\mathbf{T}_{IE} = \mathbf{T}_{01}(\varphi_1) \mathbf{T}_{12}(\varphi_2) \mathbf{T}_{23}(\varphi_3) \mathbf{T}_{34}(\varphi_4) \mathbf{T}_{45}(\varphi_5) \mathbf{T}_{56}(\varphi_6)$$

$$(\mathbf{T}_{01}(\varphi_1) \mathbf{T}_{12}(\varphi_2))^{-1} \mathbf{T}_{IE} = \mathbf{T}_{23}(\varphi_3) \mathbf{T}_{34}(\varphi_4) \mathbf{T}_{45}(\varphi_5) \mathbf{T}_{56}(\varphi_6)$$

$$(\mathbf{T}_{01}(\varphi_1) \mathbf{T}_{12}(\varphi_2) \mathbf{T}_{23}(\varphi_3))^{-1} \mathbf{T}_{IE} = \mathbf{T}_{34}(\varphi_4) \mathbf{T}_{45}(\varphi_5) \mathbf{T}_{56}(\varphi_6)$$

In RD: focus on numerical methods

Inverse Differential Kinematics

- We have seen how Jacobians map velocities from joint space to task-space

- $\mathbf{w}_e = \mathbf{J}_{e0} \dot{\mathbf{q}}$

- In general, we are interested in the inverse problem

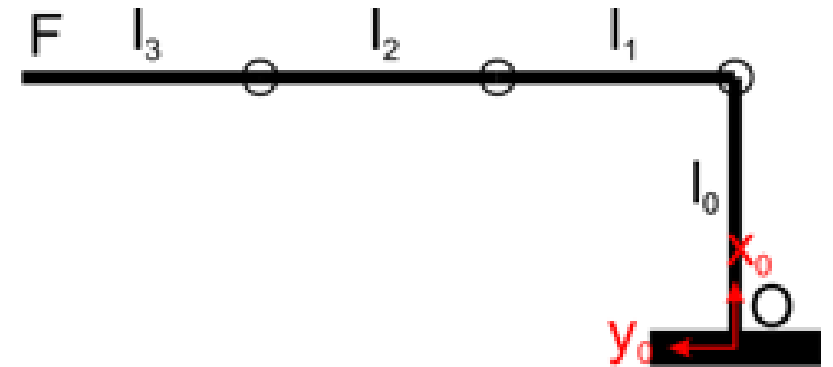
- Simple method: use the pseudoinverse

- $\dot{\mathbf{q}} = \mathbf{J}_{e0}^+ \mathbf{w}_e^*$

- ... however, the Jacobian might be singular!

Singularities

- A singularity is a joint-space configuration \mathbf{q}_s such that $\mathbf{J}_{e0}(\mathbf{q}_s)$ is column-rank deficient
 - the Jacobian becomes badly conditioned
 - small desired velocities \mathbf{w}_e^* produce high joint velocities $\dot{\mathbf{q}}$
- Singularities can be classified into:
 - boundary (e.g. a stretched out manipulator)
 - easy to avoid during motion planning
 - internal
 - harder to prevent, requires careful motion planning

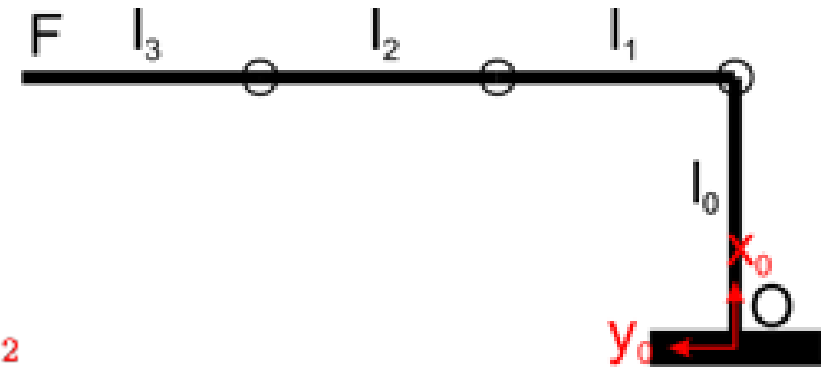


Singularities

- A singularity is a joint-space configuration \mathbf{q}_s such that $\mathbf{J}_{e0}(\mathbf{q}_s)$ is column-rank deficient
 - the Jacobian becomes badly conditioned
 - small desired velocities \mathbf{w}_e^* produce high joint velocities $\dot{\mathbf{q}}$
- Use a damped version of the Moore-Penrose pseudo inverse

$$\dot{\mathbf{q}} = \mathbf{J}_{e0}^T (\mathbf{J}_{e0} \mathbf{J}_{e0}^T + \lambda^2 \mathbf{I})^{-1} \mathbf{w}_e^* \quad \min \quad \|\mathbf{w}_e^* - \mathbf{J}_{e0} \dot{\mathbf{q}}\|^2 + \lambda^2 \|\dot{\mathbf{q}}\|^2$$

$$\lambda > 0, \lambda \in \mathbb{R}$$



Redundancy

- A kinematic structure is redundant if the dimension of the task-space is smaller than the dimension of the joint-space
 - E.g. the human arm has 7DoF (three in the shoulder, one in the elbow, and three in the wrist)
 - $\mathbf{q} \in \mathbb{R}^7$
 - $\mathbf{w} \in \mathbb{R}^6$
 - $\mathbf{J}_{e0} \in \mathbb{R}^{6 \times 7}$
- Redundancy implies infinite solutions

$$\dot{\mathbf{q}} = \mathbf{J}_{e0}^+ \mathbf{w}_e^* + \mathbf{N} \dot{\mathbf{q}}_0$$

$$\mathbf{N} = \mathcal{N}(\mathbf{J}_{e0})$$

$$\mathbf{J}_{e0} \mathbf{N} = \mathbf{0}$$

$$\mathbf{J}_{e0} (\mathbf{J}_{e0}^+ \mathbf{w}_e^* + \mathbf{N} \dot{\mathbf{q}}_0) = \mathbf{w}_e^*$$

 - One way to compute the nullspace projection matrix
 - $\mathbf{N} = \mathbf{I} - \mathbf{J}_{e0}^+ \mathbf{J}_{e0}$

Multi-task control

- Manipulation (as well as locomotion!...) is a complex combination of high level tasks
 - track a desired position
 - ensure kinematic constraints
 - reach a desired end-effector orientation
- Break down the complexity into smaller tasks
 - Two methods
 - Multi-task with equal priority
 - Multi-task with Prioritization

$$task_i := \{J_i, w_i^*\}$$



Multi-task control

Equal priority

- Assume that t tasks have been defined

- The generalised velocity is given by

$$\dot{q} = \underbrace{\begin{bmatrix} \mathbf{J}_1 \\ \vdots \\ \mathbf{J}_{n_t} \end{bmatrix}}_{\bar{\mathbf{J}}}^+ \underbrace{\begin{pmatrix} \mathbf{w}_1^* \\ \vdots \\ \mathbf{w}_{n_t}^* \end{pmatrix}}_{\bar{\mathbf{w}}}$$

The pseudo inversion will try to solve all tasks at the same time in an optimal way

- It is possible to weigh some tasks higher than others

-

$$\bar{\mathbf{J}}^{+W} = (\bar{\mathbf{J}}^T \mathbf{W} \bar{\mathbf{J}})^{-1} \bar{\mathbf{J}}^T \mathbf{W} \quad \mathbf{W} = \text{diag}(w_1, \dots, w_m)$$

Multi-task control

Prioritization

- Instead of solving all tasks at once, we can use consecutive nullspace projection to ensure a strict priority
- We already saw that $\dot{\mathbf{q}} = \mathbf{J}_1^+ \mathbf{w}_1^* + \mathbf{N}_1 \dot{\mathbf{q}}_0$
- The solution for task 2 should not violate the one found for task 1
 - $\mathbf{w}_2 = \mathbf{J}_2 \dot{\mathbf{q}} = \mathbf{J}_2 (\mathbf{J}_1^+ \mathbf{w}_1^* + \mathbf{N}_1 \dot{\mathbf{q}}_0)$

- This can be solved for $\dot{\mathbf{q}}_0$

$$\dot{\mathbf{q}}_0 = (\mathbf{J}_2 \mathbf{N}_1)^+ (\mathbf{w}_2^* - \mathbf{J}_2 \mathbf{J}_1^+ \mathbf{w}_1^*)$$

- Back substituting yields

$$\dot{\mathbf{q}} = \mathbf{J}_1^+ \mathbf{w}_1^* + \mathbf{N}_1 (\mathbf{J}_2 \mathbf{N}_1)^+ (\mathbf{w}_2^* - \mathbf{J}_2 \mathbf{J}_1^+ \mathbf{w}_1^*)$$

- The iterative solution for T tasks is then given by

$$\dot{\mathbf{q}} = \sum_{i=1}^{n_T} \mathbf{N}_i \dot{\mathbf{q}}_i, \quad \text{with} \quad \dot{\mathbf{q}}_i = (\mathbf{J}_i \mathbf{N}_i)^+ \left(\mathbf{w}_i^* - \mathbf{J} \sum_{k=1}^{i-1} \mathbf{N}_k \dot{\mathbf{q}}_k \right)$$

Multi-task control

Example - single task

- 3DoF planar robot arm with unitary link lengths

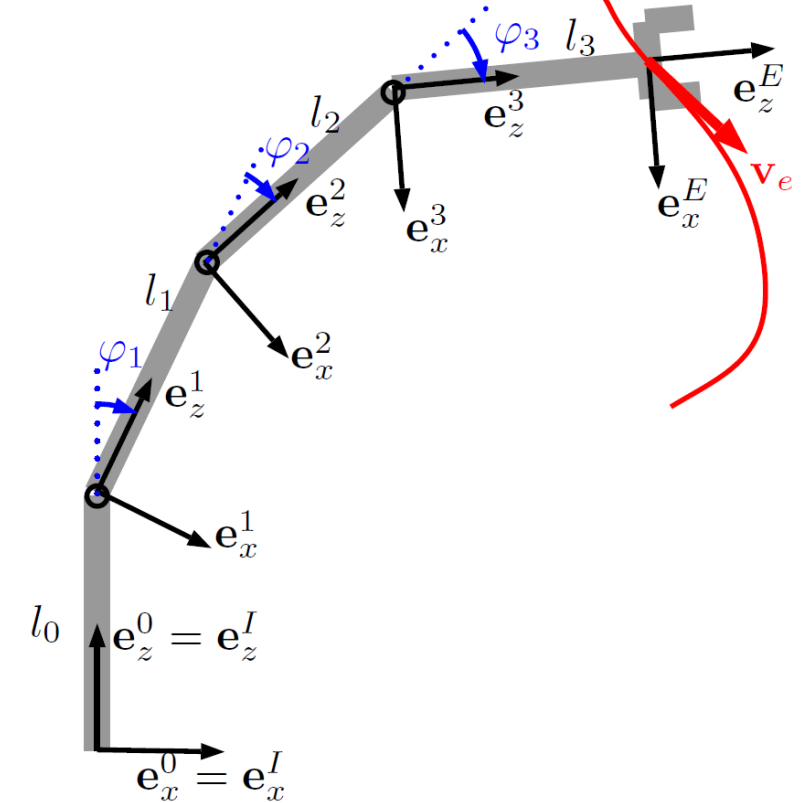
- Find the generalised velocities, given

- $\mathbf{q}_t = (\pi/6, \pi/3, \pi/3)^T$
 ${}^0\dot{\mathbf{r}}_{E,t}^* = (1, 1)^T$

- $$\mathbf{r}_E = \begin{pmatrix} \sin(\varphi_1) + \sin(\varphi_1 + \varphi_2) + \sin(\varphi_1 + \varphi_2 + \varphi_3) \\ 0 \\ 1 + \cos(\varphi_1) + \cos(\varphi_1 + \varphi_2) + \cos(\varphi_1 + \varphi_2 + \varphi_3) \end{pmatrix} = \begin{pmatrix} s_1 + s_{12} + s_{123} \\ 0 \\ 1 + c_1 + c_{12} + c_{123} \end{pmatrix}$$

$$\mathbf{J}_E = \begin{bmatrix} +c_1 + c_{12} + c_{123} & +c_{12} + c_{123} & +c_{123} \\ 0 & 0 & 0 \\ -s_1 - s_{12} - s_{123} & -s_{12} - s_{123} & -s_{123} \end{bmatrix}$$

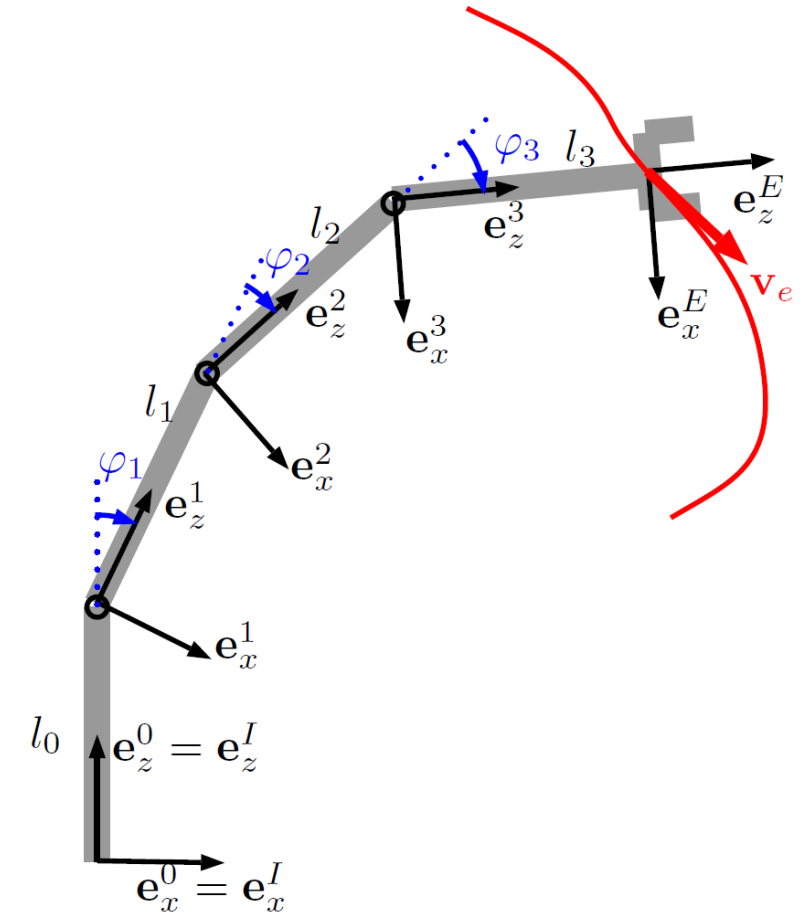
$$\mathbf{q}_t = (\pi/6, \pi/3, \pi/3)^T \quad \mathbf{J}_E = \frac{1}{2} \begin{bmatrix} +\sqrt{3} + 0 - \sqrt{3} & 0 - \sqrt{3} & -\sqrt{3} \\ 0 & 0 & 0 \\ -1 - 2 - 1 & -2 - 1 & -1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 0 & -\sqrt{3} & -\sqrt{3} \\ 0 & 0 & 0 \\ -4 & -3 & -1 \end{bmatrix}$$



Multi-task control

Example - single task

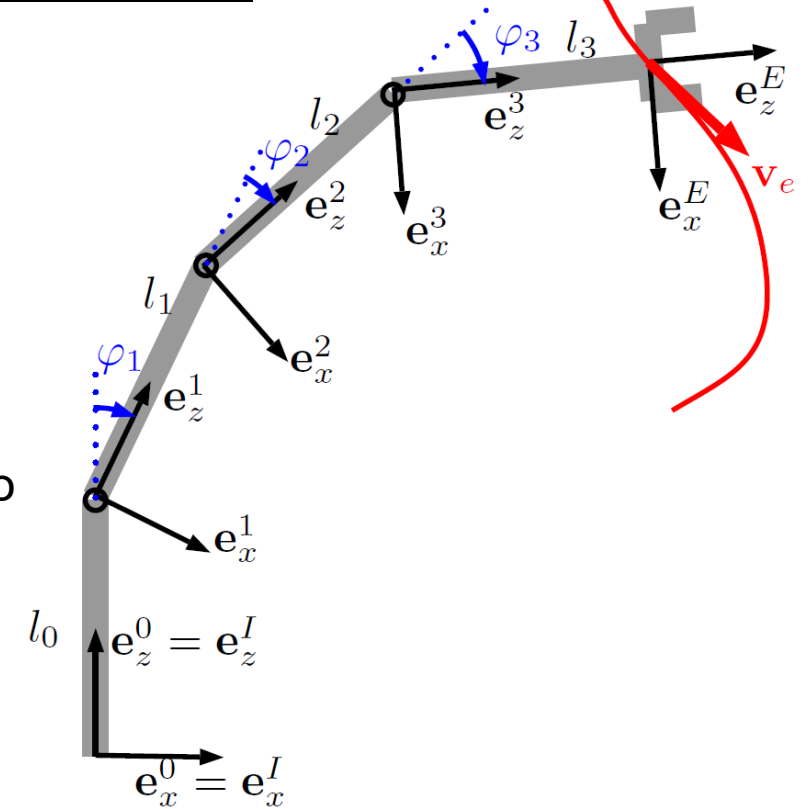
- 3DoF planar robot arm with unitary link lengths
 - Find the generalised velocities, given
 - $\mathbf{q}_t = (\pi/6, \pi/3, \pi/3)^T$ ${}^0\dot{\mathbf{r}}_{E,t}^* = (1, 1)^T$



Multi-task control

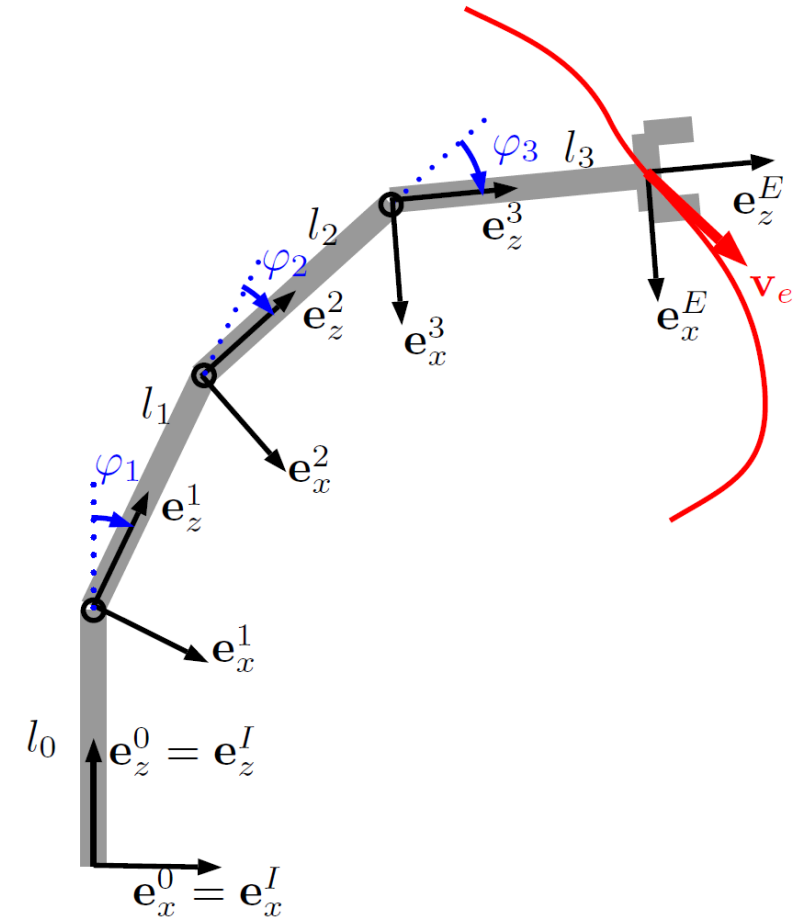
Example - stacked task

- 3DoF planar robot arm with unitary link lengths
 - Find the generalised velocities, given
 - $\mathbf{q}_t = (\pi/6, \pi/3, \pi/3)^T$ ${}^0\dot{\mathbf{r}}_{E,t}^* = (1, 1)^T$
 - Additionally, we want to fulfill a second task with the same priority as the first, namely that the first and **second** joint velocities are zero

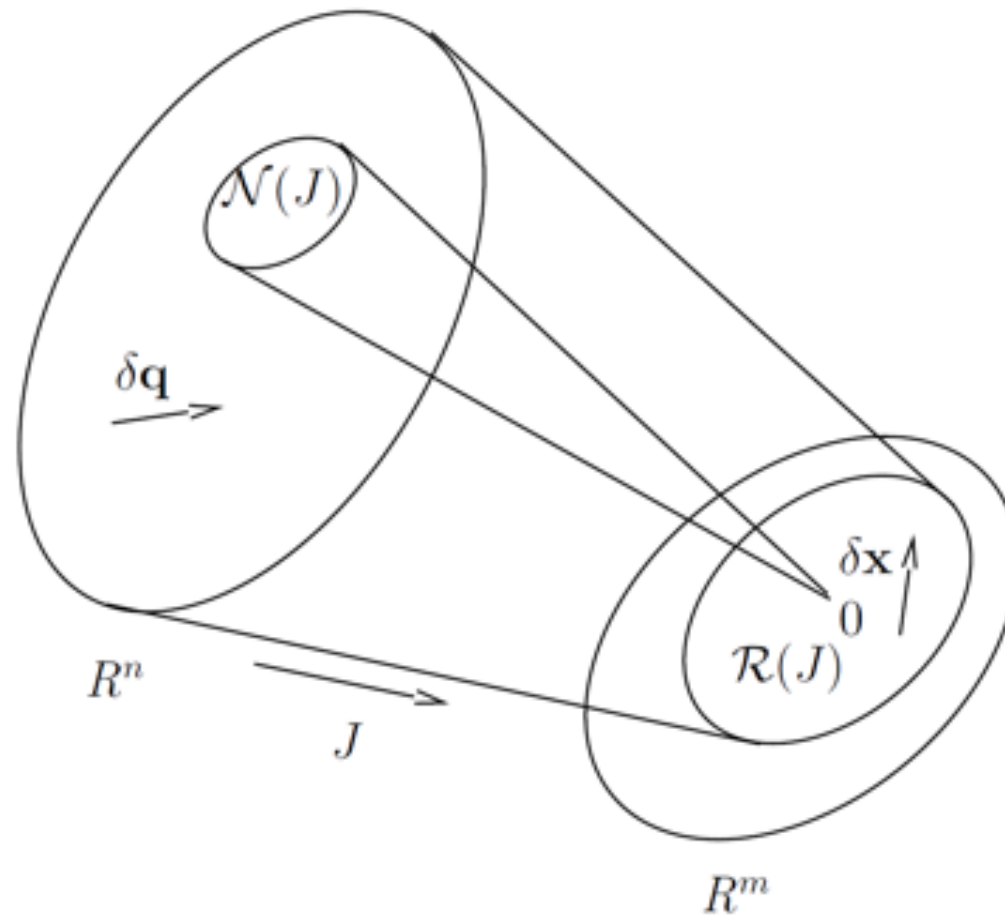


Multi-task control

Example - stacked task



Mapping associated with the Jacobian



Numerical solutions

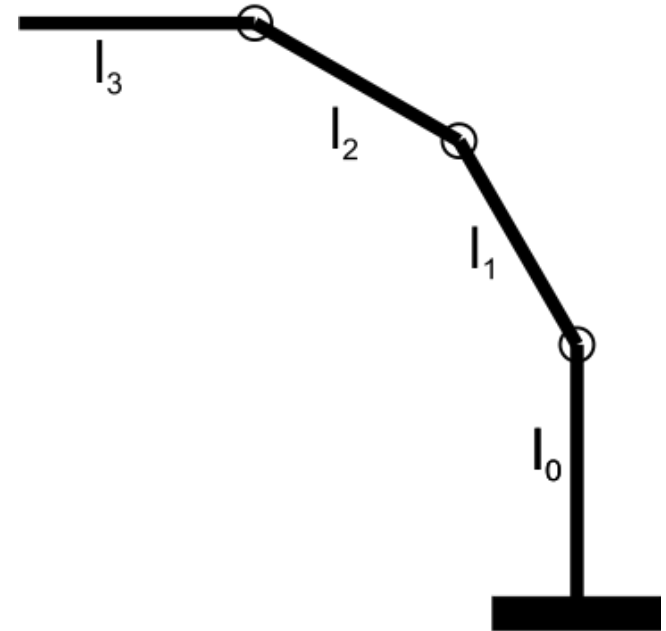
Inverse differential kinematics

- Jacobians map joint-space velocities to end-effector velocities
 - $\dot{\chi}_e = \mathbf{J}_{eA}(\mathbf{q})\dot{\mathbf{q}}$ $\Delta\chi_e = \mathbf{J}_{eA}(\mathbf{q}) \cdot \Delta\mathbf{q}$
- We can use this to iteratively solve the inverse kinematics problem
 - target configuration χ_e^* , initial joint space guess \mathbf{q}^0
 1. $\mathbf{q} \leftarrow \mathbf{q}^0$ ▷ start configuration
 2. while $\|\chi_e^* - \chi_e(\mathbf{q})\| \geq \text{tol}$ do ▷ while the solution is not reached
 3. $\mathbf{J}_{eA} \leftarrow \mathbf{J}_{eA}(\mathbf{q}) = \frac{\partial \chi_e}{\partial \mathbf{q}}(\mathbf{q})$ ▷ evaluate Jacobian
 4. $\mathbf{J}_{eA}^+ \leftarrow (\mathbf{J}_{eA}(\mathbf{q}))^+$ ▷ compute the pseudo inverse
 5. $\Delta\chi_e \leftarrow \chi_e^* - \chi_e(\mathbf{q})$ ▷ find the end-effector configuration error vector
 6. $\mathbf{q} \leftarrow \mathbf{q} + \mathbf{J}_{eA}^+ \Delta\chi_e$ ▷ updated the generalized coordinates

Inverse kinematics

Three-link arm example

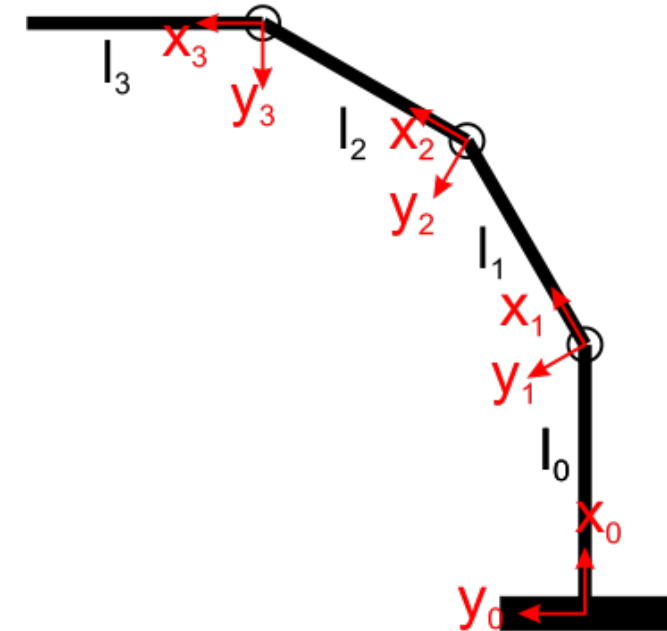
- Determine end-effector Jacobian



Inverse kinematics

Three-link arm example

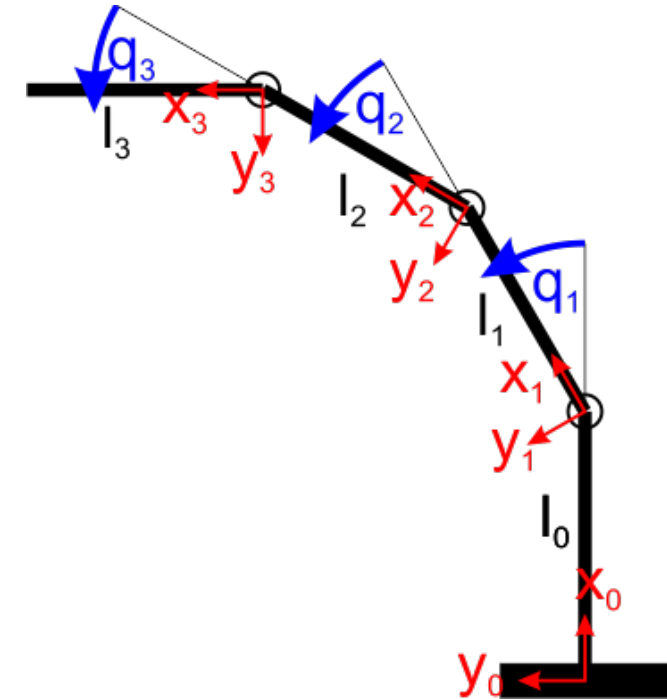
- Determine end-effector Jacobian
 1. Introduce coordinate frames



Inverse kinematics

Three-link arm example

- Determine end-effector Jacobian
 1. Introduce coordinate frames
 2. Introduce generalized coordinates



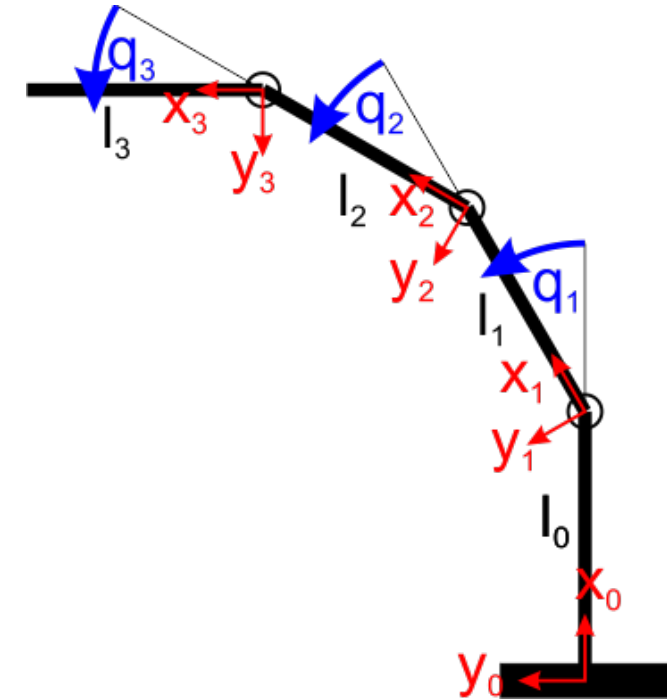
Inverse kinematics

Three-link arm example

- Determine end-effector Jacobian

1. Introduce coordinate frames
2. Introduce generalized coordinates
3. Determine end-effector position

$${}^0\mathbf{r}_{0E}(\mathbf{q}) = \begin{bmatrix} l_0 + l_1 \cos(q_1) + l_2 \cos(q_1 + q_2) + l_3 \cos(q_1 + q_2 + q_3) \\ l_1 \sin(q_1) + l_2 \sin(q_1 + q_2) + l_3 \sin(q_1 + q_2 + q_3) \\ 0 \end{bmatrix}$$



Inverse kinematics

Three-link arm example

■ Determine end-effector Jacobian

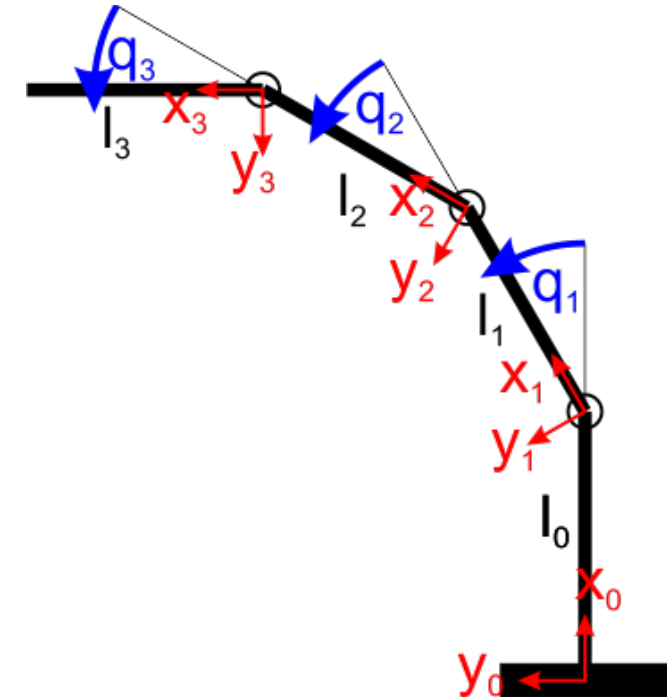
1. Introduce coordinate frames
2. Introduce generalized coordinates
3. Determine end-effector position

$${}^0\mathbf{r}_{0E}(\mathbf{q}) = \begin{bmatrix} l_0 + l_1 \cos(q_1) + l_2 \cos(q_1 + q_2) + l_3 \cos(q_1 + q_2 + q_3) \\ l_1 \sin(q_1) + l_2 \sin(q_1 + q_2) + l_3 \sin(q_1 + q_2 + q_3) \\ 0 \end{bmatrix}$$

4. Compute the Jacobian

$${}^0\mathbf{J}_{eP} = \frac{\partial}{\partial \mathbf{q}} {}^0\mathbf{r}_{0E}(\mathbf{q})$$

$$= \begin{bmatrix} -l_1 \sin(q_1) - l_2 \sin(q_1 + q_2) - l_3 \sin(q_1 + q_2 + q_3) & -l_2 \sin(q_1 + q_2) - l_3 \sin(q_1 + q_2 + q_3) & -l_3 \sin(q_1 + q_2 + q_3) \\ l_1 \cos(q_1) + l_2 \cos(q_1 + q_2) + l_3 \cos(q_1 + q_2 + q_3) & l_2 \cos(q_1 + q_2) + l_3 \cos(q_1 + q_2 + q_3) & l_3 \cos(q_1 + q_2 + q_3) \\ 0 & 0 & 0 \end{bmatrix}$$



Inverse kinematics

Three-link arm example

- Iterative inverse kinematics to find desired configuration
 - $\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eP}^+(\mathbf{r}_{goal} - \mathbf{r}^i)$

Inverse kinematics

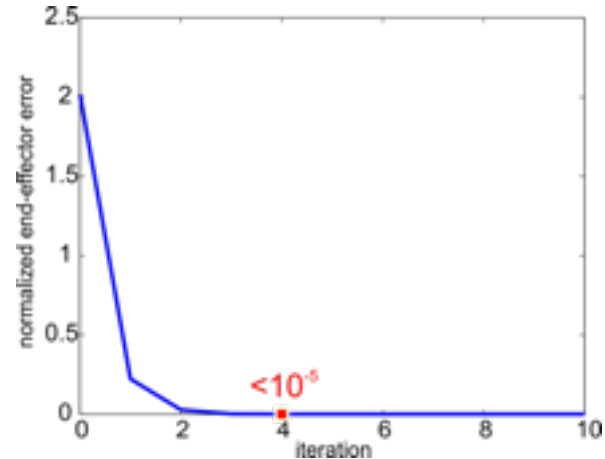
Three-link arm example

- Iterative inverse kinematics to find desired configuration

- $\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eP}^+(\mathbf{r}_{goal} - \mathbf{r}^i)$

- start value

$$\mathbf{q}^0 = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$



With zero start point

$$\mathbf{q}_0 = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Inverse kinematics

Three-link arm example

- Iterative inverse kinematics to find desired configuration

- $\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eP}^+(\mathbf{r}_{goal} - \mathbf{r}^i)$

- start value

$$\mathbf{q}^0 = \begin{bmatrix} \pi/2 \\ 0 \\ 0 \end{bmatrix}$$



Inverse kinematics

Three-link arm example

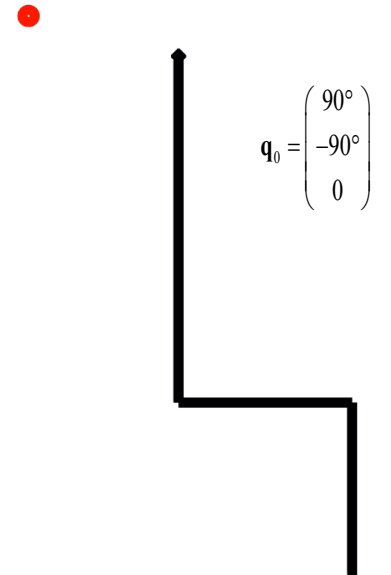
- Iterative inverse kinematics to find desired configuration

- $\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eP}^+(\mathbf{r}_{goal} - \mathbf{r}^i)$

- start value

$$\mathbf{q}^0 = \begin{bmatrix} \pi/2 \\ -\pi/2 \\ 0 \end{bmatrix}$$

- Same goal position, multiple solutions
 - joint-space bigger than task-space, redundant system



Inverse kinematics

Iterative methods

- Let's have a closer look at the joint update rule

- $\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eA}^+ \Delta \chi$

- Two main issues

- Scaling

- if the current error is too large, the error linearization implemented by the Jacobian is not accurate enough
 - use a scaling factor $0 < k < 1$ $\mathbf{q}^{i+1} = \mathbf{q}^i + k \mathbf{J}_{eA}^+ \Delta \chi$
 - unfortunately, this will lead to slower convergence

Inverse kinematics

Iterative methods

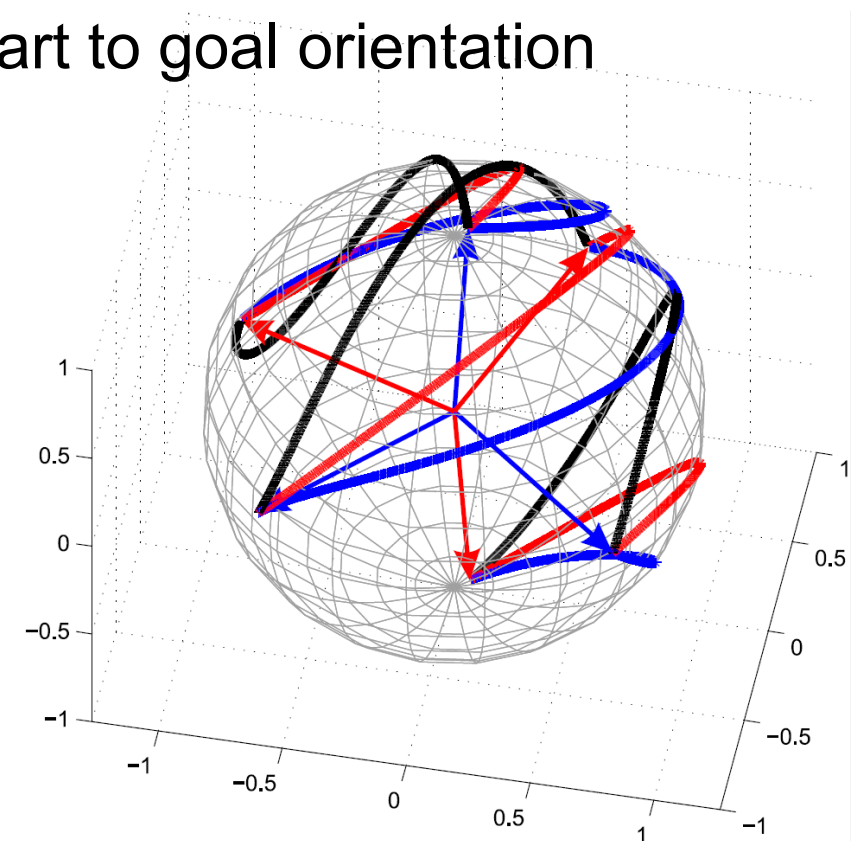
- Let's have a closer look at the joint update rule
 - $\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eA}^+ \Delta\chi$
- Two main issues
 - Singular configurations
 - When the Jacobian is rank-deficient, the inversion becomes a badly conditioned problem
 - Use the damped pseudoinverse (Levenberg-Marquardt)
 - $\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eA}^T (\mathbf{J}_{eA} \mathbf{J}_{eA}^T + \lambda^2 \mathbf{I})^{-1} \Delta\chi$
 - Use the transpose of the Jacobian
 - $\mathbf{q}^{i+1} = \mathbf{q}^i + \alpha \mathbf{J}_{eA}^T \Delta\chi$
- For a detailed explanation, check “Introduction to Inverse Kinematics with Jacobian Transpose, Pseudoinverse and Damped Least Squares methods”, **Samuel Buss**, 2009

Inverse differential kinematics

Orientation error

- 3D rotations are defined in the Special Orthogonal group $SO(3)$
- The parametrization affects convergence from start to goal orientation

$$\mathbf{q}^{i+1} = \mathbf{q}^i + \mathbf{J}_{eA}^+ \Delta\chi$$



Inverse differential kinematics

Orientation error

- 3D rotations are defined in the Special Orthogonal group $SO(3)$
- The parametrization affects convergence from start to goal orientation
 - Rotate along shortest path in $SO(3)$: use rotational vectors which parametrize rotation from start to goal

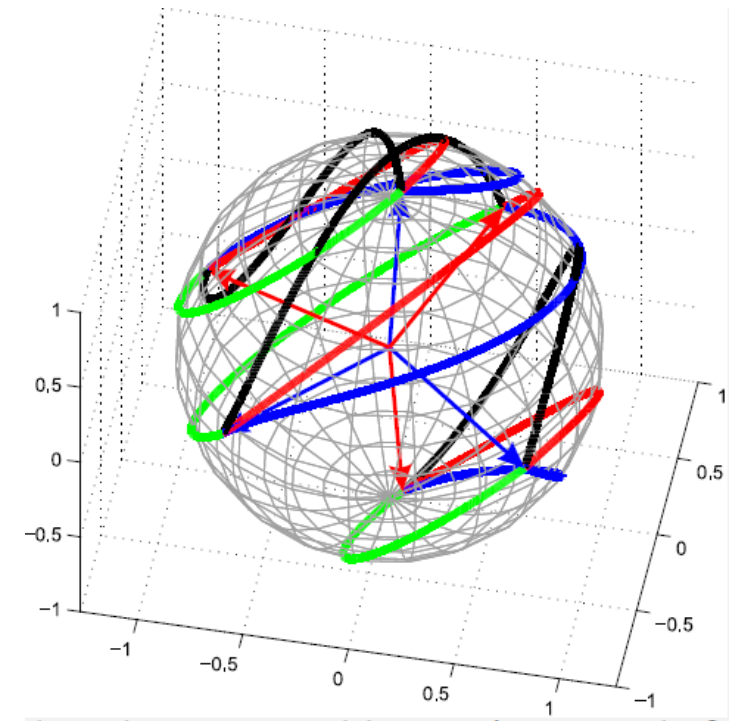
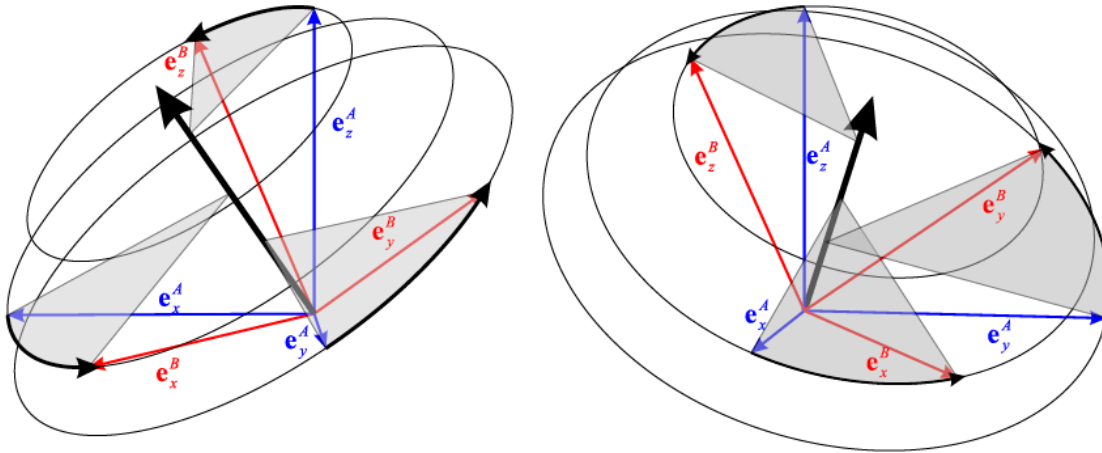
$$\Delta\chi_{rotvec} = \Delta\varphi \quad \implies \quad \mathbf{C}_{gS}(\Delta\varphi) = \mathbf{C}_{gI}(\varphi^*)\mathbf{C}_{SI}^T(\varphi^t)$$

- The update law for rotations will then be

$$\mathbf{q} \leftarrow \mathbf{q} + k_{PR}\mathbf{J}_{e0R}^+ \Delta\varphi$$

This is NOT the difference between rotation vectors, but the rotation vector extracted from the relative rotation between start and goal

Rotation with rotation vector and angle



Trajectory control

Position

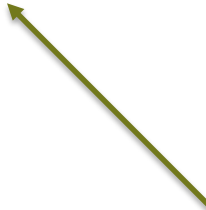
- Consider a planned desired motion of the end effector
 - $\mathbf{r}_e^*(t)$
 $\dot{\mathbf{r}}_e^*(t)$
- Let's see how to kinematically control the end-effector position
 - Feedback term
 - $\Delta \mathbf{r}_e^t = \mathbf{r}_e^*(t) - \mathbf{r}_e(\mathbf{q}^t)$
 - We can design a nonlinear stabilizing controller law
 - $\dot{\mathbf{q}}^* = \mathbf{J}_{e0p}^+(\mathbf{q}^t) \cdot (\dot{\mathbf{r}}_e^*(t) + k_{pp} \Delta \mathbf{r}_e^t)$

Trajectory control

Orientation

- Derivation more involved
- Final control law similar to the position case

$$\dot{\mathbf{q}} = \mathbf{J}_{e0R}^+ (\omega(t)_e^* + k_{PR}\Delta\varphi)$$



Note that we are not using the analytical Jacobian since we are dealing with angular velocities and rotational vectors



Floating Base Kinematics

151-0851-00 V

| | | |
|--------------|----------|--|
| lecture: | CAB G11 | Tuesday 10:15 – 12:00, every week |
| exercise: | HG E1.2 | Wednesday 8:15 – 10:00, according to schedule (about every 2nd week) |
| office hour: | LEE H303 | Friday 12.15 – 13.00 |

Marco Hutter, Roland Siegwart, and Thomas Stastny

Floating Base Systems

Kinematics

- Generalized coordinates

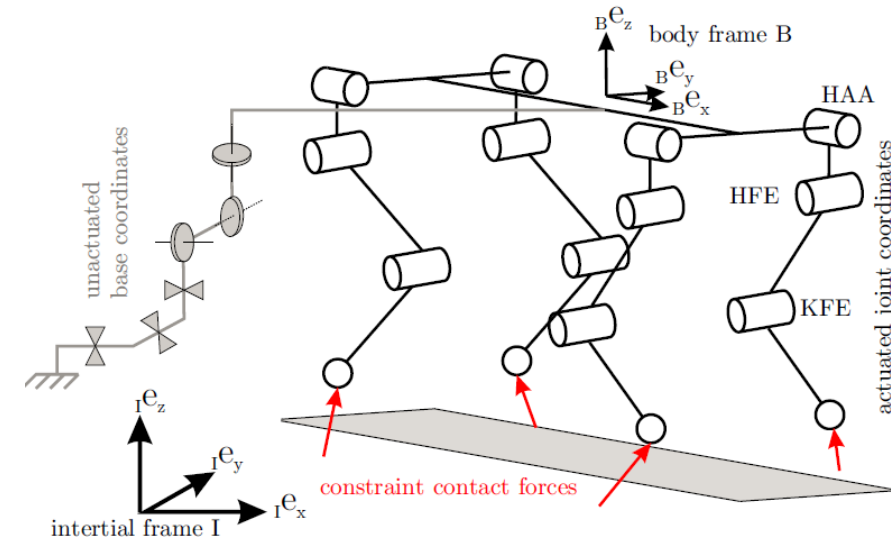
$$\mathbf{q} = \begin{pmatrix} \mathbf{q}_b \\ \mathbf{q}_j \end{pmatrix} \quad \text{with} \quad \mathbf{q}_b = \begin{pmatrix} \mathbf{q}_{b_P} \\ \mathbf{q}_{b_R} \end{pmatrix} \in \mathbb{R}^3 \times SO(3)$$

- Generalized velocities and accelerations?

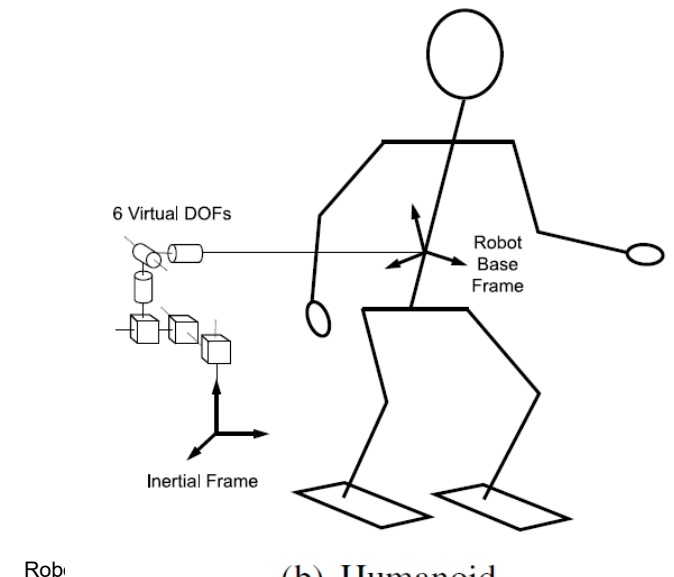
- Time derivatives $\dot{\mathbf{q}}, \ddot{\mathbf{q}}$ depend on parameterization

$$\text{Often} \quad \mathbf{u} = \begin{pmatrix} I\mathbf{v}_B \\ \boxed{\mathbf{E}}\boldsymbol{\omega}_{IB} \\ \dot{\varphi}_1 \\ \vdots \\ \dot{\varphi}_{n_j} \end{pmatrix} \in \mathbb{R}^{6+n_j} = \mathbb{R}^{n_u} \quad \dot{\mathbf{u}} = \begin{pmatrix} I\mathbf{a}_B \\ \boxed{\mathbf{E}}\dot{\boldsymbol{\psi}}_{IB} \\ \ddot{\varphi}_1 \\ \vdots \\ \ddot{\varphi}_{n_j} \end{pmatrix} \in \mathbb{R}^{6+n_j}$$

$$\text{Linear mapping} \quad \mathbf{u} = \mathbf{E}_{fb} \cdot \dot{\mathbf{q}}, \text{ with } \mathbf{E}_{fb} = \begin{bmatrix} \mathbb{I}_{3 \times 3} & 0 & 0 \\ 0 & \mathbf{E}_{\chi_R} & 0 \\ 0 & 0 & \mathbb{I}_{n_j \times n_j} \end{bmatrix}$$



(a) Quadruped



(b) Humanoid

Floating Base Systems

Differential kinematics

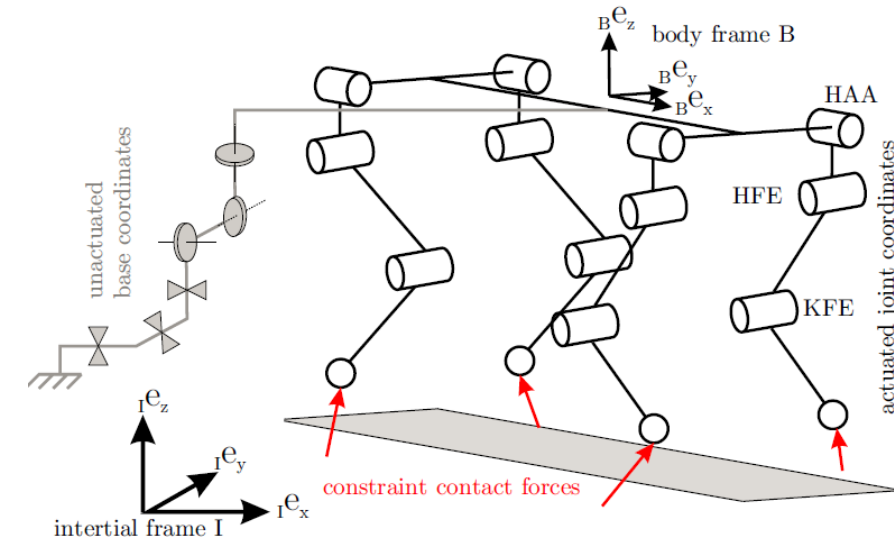
- Position of an arbitrary point on the robot

$$\mathcal{I}\mathbf{r}_{IQ}(\mathbf{q}) = \underbrace{\mathcal{I}\mathbf{r}_{IB}(\mathbf{q})}_{\mathcal{I}\mathbf{r}_{IB}(\mathbf{q}_b)} + \underbrace{\mathbf{C}_{IB}(\mathbf{q})}_{\mathbf{C}_{IB}(\mathbf{q}_b)} \cdot \underbrace{{}^B\mathbf{r}_{BQ}(\mathbf{q})}_{{}^B\mathbf{r}_{BQ}(\mathbf{q}_j)}$$

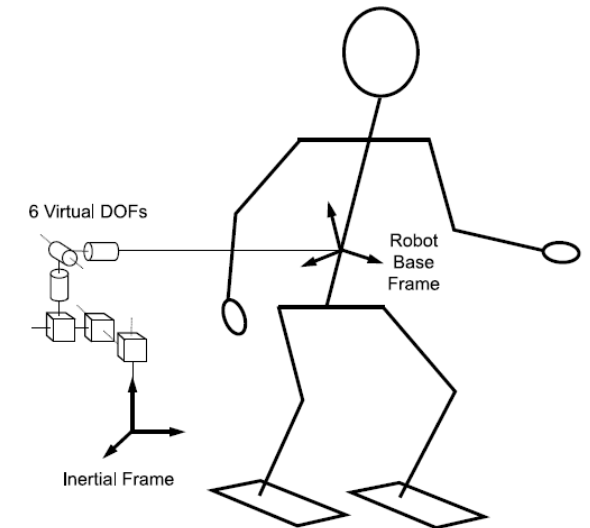
- Velocity of this point

$$\begin{aligned}\mathcal{I}\mathbf{v}_Q &= \mathcal{I}\mathbf{v}_B + \dot{\mathbf{C}}_{IB} \cdot {}^B\mathbf{r}_{BQ} + \mathbf{C}_{IB} \cdot {}^B\dot{\mathbf{r}}_{BQ} \\ &= \mathcal{I}\mathbf{v}_B + \mathbf{C}_{IB} \cdot [{}^B\boldsymbol{\omega}_{IB}]_{\times} \cdot {}^B\mathbf{r}_{BQ} + \mathbf{C}_{IB} \cdot {}^B\dot{\mathbf{r}}_{BQ} \\ &= \mathcal{I}\mathbf{v}_B - \mathbf{C}_{IB} \cdot [{}^B\mathbf{r}_{BQ}]_{\times} \cdot {}^B\boldsymbol{\omega}_{IB} + \mathbf{C}_{IB} \cdot {}^B\dot{\mathbf{r}}_{BQ} \\ &= \mathcal{I}\mathbf{v}_B - \mathbf{C}_{IB} \cdot [{}^B\mathbf{r}_{BQ}]_{\times} \cdot {}^B\boldsymbol{\omega}_{IB} + \mathbf{C}_{IB} \cdot {}^B\mathbf{J}_{P_{q_j}}(\mathbf{q}_j) \cdot \dot{\mathbf{q}}_j \\ &= \underbrace{\begin{bmatrix} \mathbb{I}_{3 \times 3} & -\mathbf{C}_{IB} \cdot [{}^B\mathbf{r}_{BQ}]_{\times} & \mathbf{C}_{IB} \cdot {}^B\mathbf{J}_{P_{q_j}}(\mathbf{q}_j) \end{bmatrix}}_{\mathcal{I}\mathbf{J}_Q(\mathbf{q})} \cdot \mathbf{u} \quad \text{with} \quad \mathbf{u} = \begin{pmatrix} \mathcal{I}\mathbf{v}_B \\ {}^B\boldsymbol{\omega}_{IB} \\ \dot{\varphi}_1 \\ \vdots \\ \dot{\varphi}_{n_j} \end{pmatrix}\end{aligned}$$

$$\begin{aligned}[{}^B\boldsymbol{\omega}_{IB}]_{\times} &= \mathbf{C}_{BI} [{}^I\boldsymbol{\omega}_{IB}]_{\times} \mathbf{C}_{BI}^T \\ &= \mathbf{C}_{IB}^T \dot{\mathbf{C}}_{IB} \mathbf{C}_{IB}^T \mathbf{C}_{IB} = \mathbf{C}_{IB}^T \dot{\mathbf{C}}_{IB}\end{aligned}$$



(a) Quadruped



Robot

(b) Humanoid

Contact Constraints

- A contact point C_i is not allowed to move:

$$\mathcal{I}\mathbf{r}_{IC_i} = \text{const}, \quad \mathcal{I}\dot{\mathbf{r}}_{IC_i} = \mathcal{I}\ddot{\mathbf{r}}_{IC_i} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

- Constraint as a function of generalized coordinates:

$$\mathcal{I}\mathbf{J}_{C_i}\mathbf{u} = \mathbf{0}, \quad \mathcal{I}\mathbf{J}_{C_i}\dot{\mathbf{u}} + \mathcal{I}\dot{\mathbf{J}}_{C_i}\mathbf{u} = \mathbf{0}$$

- Stack of constraints

$$\mathbf{J}_c = \begin{bmatrix} \mathbf{J}_{C_1} \\ \vdots \\ \mathbf{J}_{C_{n_c}} \end{bmatrix} \in \mathbb{R}^{3n_c \times n_n}$$

Contact Constraint

Wheeled vehicle simple example

- Contact constraints

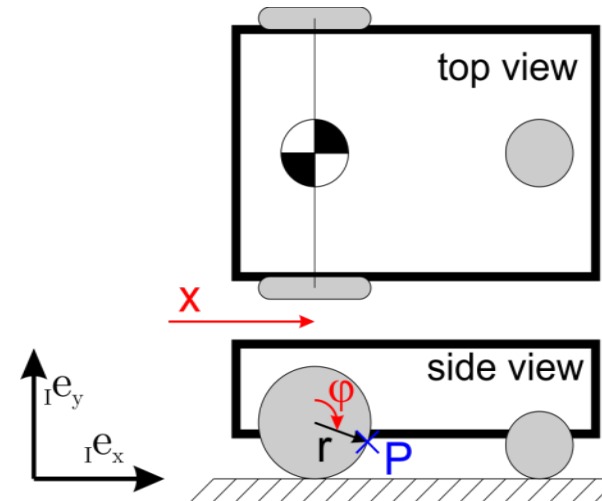
- Point on wheel ${}_{\mathcal{I}}\mathbf{r}_{IP} = \begin{pmatrix} x + r \sin(\varphi) \\ r + r \cos(\varphi) \\ 0 \end{pmatrix}$

- Jacobian ${}_{\mathcal{I}}\mathbf{J}_P = \begin{bmatrix} 1 & r \cos(\varphi) \\ 0 & -r \sin(\varphi) \\ 0 & 0 \end{bmatrix}$

- Contact constraints

$${}_{\mathcal{I}}\dot{\mathbf{r}}_{IP}|_{\varphi=\pi} = {}_{\mathcal{I}}\mathbf{J}_P|_{\varphi=\pi} \dot{\mathbf{q}} = \begin{bmatrix} 1 & -r \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{pmatrix} \dot{x} \\ \dot{\varphi} \end{pmatrix} = \mathbf{0}$$

=> Rolling condition $\dot{x} - r\dot{\varphi} = 0$



$$\mathbf{q} = \begin{pmatrix} x \\ \varphi \end{pmatrix} \quad \begin{array}{l} \text{Un-actuated base} \\ \text{Actuated joints} \end{array}$$

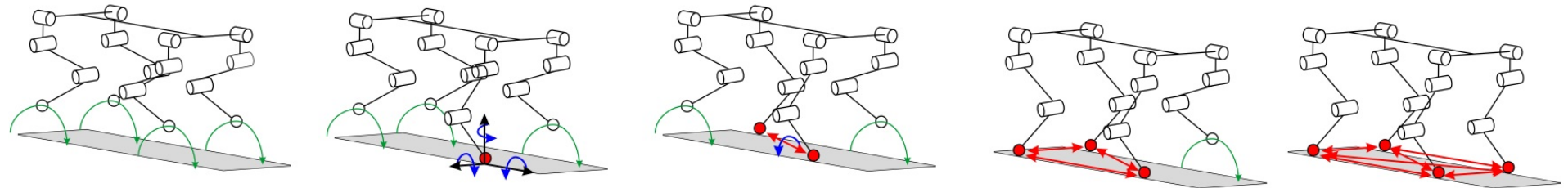
Properties of Contact Jacobian

- Contact Jacobian tells us, how a system can move.
 - Separate stacked Jacobian $\mathbf{J}_c = [\mathbf{J}_{c,b} \quad \mathbf{J}_{c,j}] = \begin{bmatrix} \frac{\partial \mathbf{r}_c}{\partial \mathbf{q}_b} & \frac{\partial \mathbf{r}_c}{\partial \mathbf{q}_j} \end{bmatrix} \in \mathbb{R}^{n_c \times (n_b + n_j)}$

relation between base motion and constraints
 - Base is fully controllable if $\text{rank}(\mathbf{J}_{c,b}) = 6$
 - Nr of kinematic constraints for joint actuators: $\text{rank}(\mathbf{J}_c) - \text{rank}(\mathbf{J}_{c,b})$
- Generalized coordinates DON'T correspond to the degrees of freedom
 - Contact constraints!
- Minimal coordinates (= correspond to degrees of freedom)
 - Require to switch the set of coordinates depending on contact state (=> never used)

Quadrupedal Robot with Point Feet

- Floating base system with 12 actuated joint and 6 base coordinates (18DoF)



| | | | | | |
|----------------------|---|---|---|---|----|
| Total constraints | 0 | 3 | 6 | 9 | 12 |
| Internal constraints | 0 | 0 | 1 | 3 | 6 |
| Uncontrollable DoFs | 6 | 3 | 1 | 0 | 0 |