

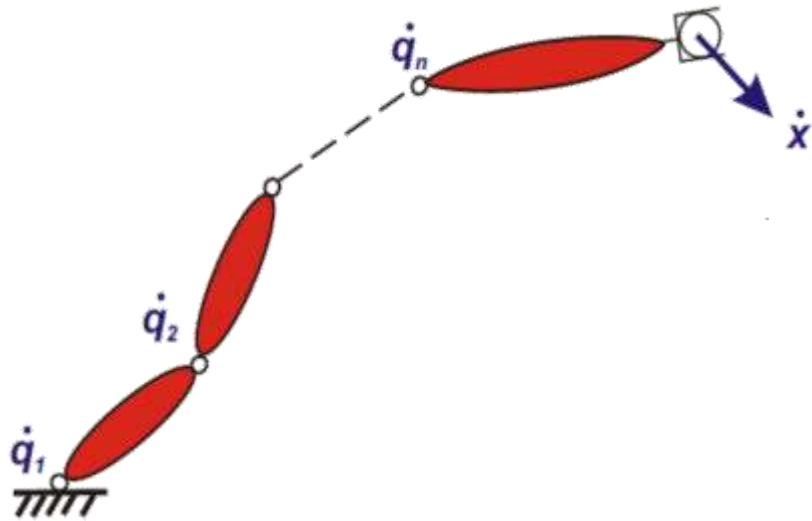
Intelligent Robot Control

Lecture 3: Control of Redundant Robots

Tadej Petrič

tadej.petric@ijs.si

Robot manipulators



- Robot is seen as (open) kinematic chain of rigid bodies interconnected by (revolute or prismatic) joints.

- Parameterization:

- Unambiguous and minimal characterization of the robot configuration
- n = degrees of freedom (DOF)
- n = robot joints (rotational or translational)

- Configuration on n -DOF robot is described by joint coordinates:

$$q = [q_1, q_2, \dots, q_n]^T \quad q_i \in Q_i = [q_{i,min}, q_{i,max}]$$

- The configuration space \mathcal{C} is the space where the joint variables q are defined:

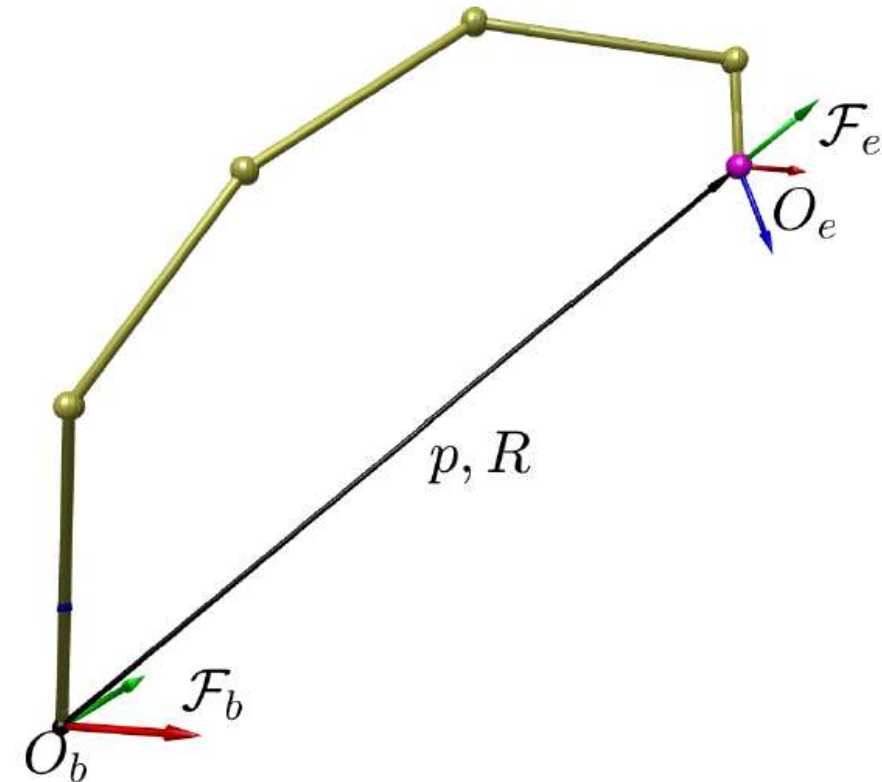
$$\mathcal{C} : Q_1 \times Q_2 \times \dots \times Q_n$$

Robot pose

- End-effector position (operational point) in base coordinate frame:

$$\mathbf{T}_e = \begin{bmatrix} \mathbf{R}(q) & p(q) \\ \mathbf{0} & 1 \end{bmatrix}$$

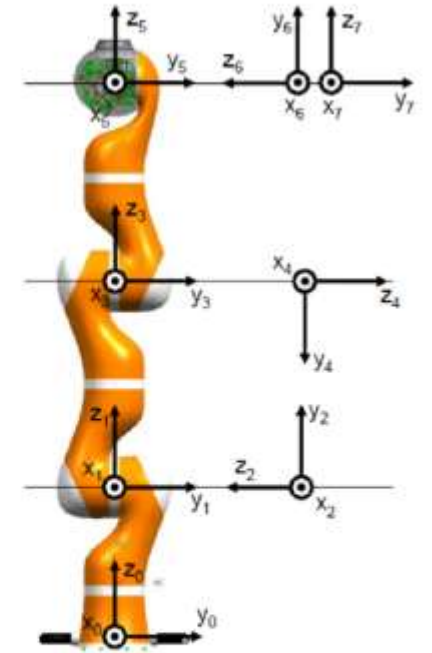
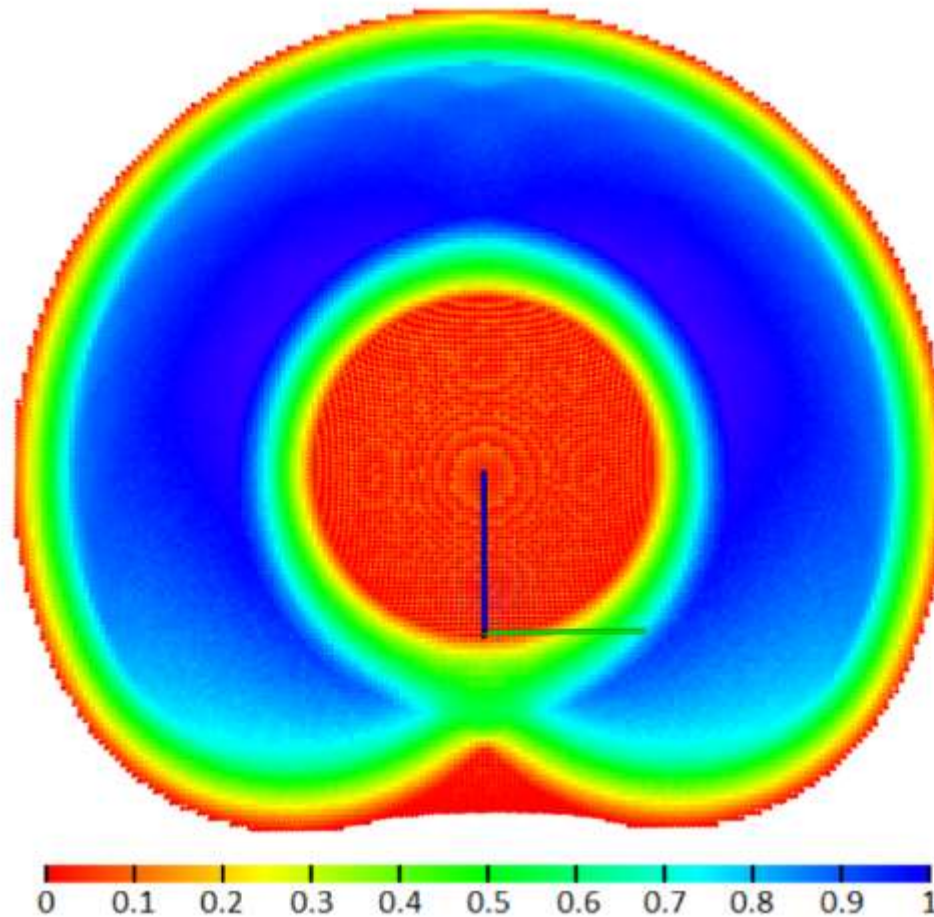
- **The operational space** O is space, where the positions/orientations of the robot end-effector are defined (6-dimensional Cartesian space).
- **Reachable workspace**: is the set of all p where the robot can reach all positions with at least one orientation
- **Dexterous workspace**: is the set of all p where the robot can reach all positions with any feasible orientation



Working envelope of Kuka LWR

Cross-section of the workspace of a KUKA LBR arm in form of a Capability map.

The HSV color scale encodes the dexterity of each region in the space; blue indicates areas with higher dexterity of the manipulator.



Redundant robots

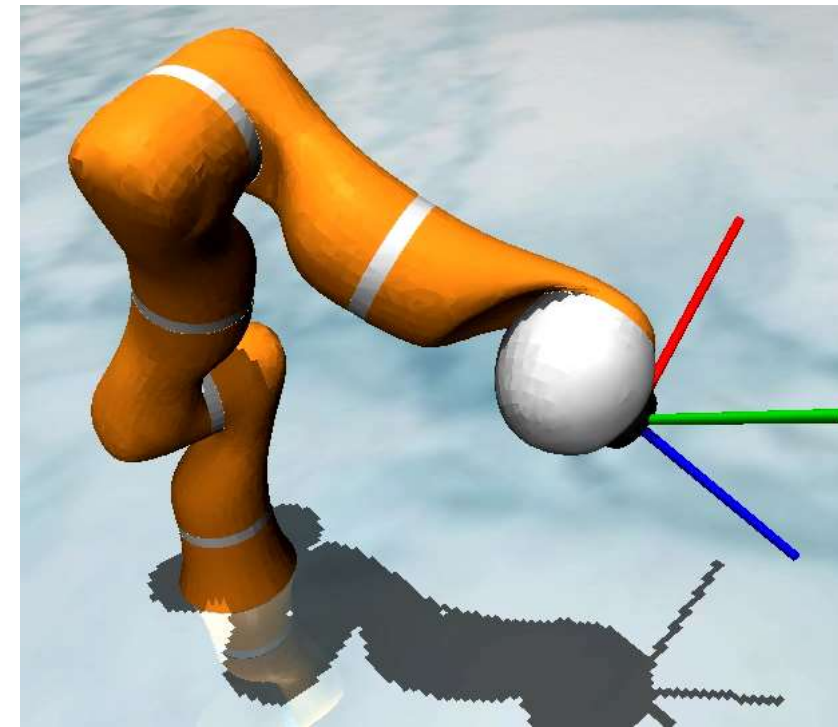
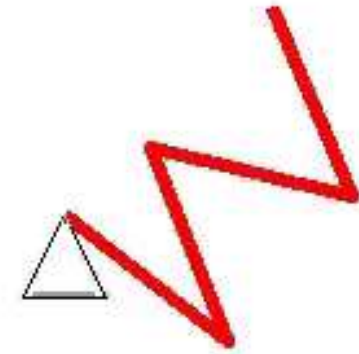
- A robot is redundant if it has **more degrees-of-freedom (DOF) than needed** to accomplish a task.

- Two types of redundancy can be identified:
 - Serial robots that have a joint-space dimension greater than their operational-space dimension are termed **intrinsically redundant** :

$$r_i = \dim(\mathcal{C}) - \dim(\mathcal{O})$$

- A robot is termed **functionally redundant** if the task does not use all operational-space dimensions, $\mathcal{T} \subset \mathcal{O}$:

$$r_f = \dim(\mathcal{O}) - \dim(\mathcal{T})$$



Task space

- Task space $\mathcal{T} \subseteq \mathcal{O}$ is the space where the operation of robot is required.
- DOFs needed for some common tasks:
 - $m = 2$
 - pointing in space
 - positioning in plane
 - $m = 3$
 - orientation in space
 - positioning and orientation in plane
 - $m = 5$
 - positioning and pointing in space
 - $m = 6$
 - positioning and orientation in space

Kinematics of redundant robot

- **Forward kinematics** of redundant robots is given in the form

$$\dot{\mathbf{x}}_{(m \times 1)} = \mathbf{J}_{(m \times n)} \dot{\mathbf{q}}_{(n \times 1)} \quad m < n$$

- **Inverse kinematics** problem is now to solve this set of equations. The system is under constraint and can be solved by choosing some additional constraints. The solution is of the form:

$$\dot{\mathbf{q}} = \mathbf{J}^\dagger \dot{\mathbf{x}}$$

- Where \mathbf{J}^\dagger is some **generalized** inverse of \mathbf{J} , e.g. any matrix satisfying

$$\mathbf{J}\mathbf{J}^\dagger\mathbf{J} = \mathbf{J}$$

- **Note:** Generalized inverse always exists.

Least-norm solution

- Consider: $y = Ax$
- Where $x \in \mathbb{R}^n$, $y \in \mathbb{R}^m$, $A \in \mathbb{R}^{m \times n}$, and $m < n$ (We have more variables than equations).

- Optimization problem:

$$\min \|y\| \quad \text{subject to: } Ay = x$$

- Assume A has full row rank, $\mathcal{R}(A) = m$. Then the solution has form:

$$\{x \mid Ax = y\} = \{x_p + z \mid z \in \mathcal{N}(A)\}$$

- One particular solution is

$$x_p = A^T (A A^T)^{-1} y$$

Pseudo-inverse

- Most commonly used pseudo-inverses are **Moore-Penrose pseudo-inverse** (minimal joint velocities)

$$\mathbf{J}^\dagger = \mathbf{J}^T (\mathbf{J}\mathbf{J}^T)^{-1}$$

- or **weighted pseudo-inverse**, where \mathbf{W} is a weighting matrix:

$$\mathbf{J}^\dagger = \mathbf{W}^{-1} \mathbf{J}^T (\mathbf{J}\mathbf{W}^{-1} \mathbf{J}^T)^{-1}$$

- Special case when $\mathbf{W} = \mathbf{H} \rightarrow$ **dynamic consistent** pseudoinverse.
- When the robot approaches a **singular configuration** this solution **becomes inefficient**, very high joint velocities are required even for small task space velocities in the directions which become unfeasible in the singularity.

Damped least-squares inverse

- From the mathematical point of view, the Jacobian \mathbf{J} becomes near singular configuration **ill-conditioned** (some singular values of \mathbf{J} become very small).
- Solution is the damped least-squares inverse (DLS)
$$\mathbf{J}^* = \mathbf{W}^{-1} \mathbf{J}^T (\mathbf{J} \mathbf{W}^{-1} \mathbf{J}^T + \lambda^2 \mathbf{I})^{-1}$$
- where λ is the damping factor. The additional damping term $\lambda^2 \mathbf{I}$ decreases the task space accuracy in favor of **feasible joint velocities**.
- \mathbf{J}^* **does not fulfill** all Moore-Penrose conditions, e.g. $\mathbf{J} \mathbf{J}^* \mathbf{J} \neq \mathbf{J}$. Hence, the DLS inverse **should not be used** in the calculation of the null-space projectors.

General inverse kinematics solution

- The question is now, how to incorporate any constraints in the general solution given in the form

$$\dot{q} = \mathbf{J}^\dagger \dot{x} + (\mathbf{I}_n - \mathbf{J}^\dagger \mathbf{J}) \dot{\varphi}$$

- where \mathbf{I}_n is identity matrix and $\dot{\varphi}$ an arbitrary vector.
- The term $\mathbf{J}^\dagger \dot{x}$ represents the **particular** solution which satisfies the main task and any "rigid" constraints depending on the selected generalized inverse.
- To find a suitable generalized inverse \mathbf{J}^\dagger we specify some performance criterion. By finding the optimum of this criterion we get the desired generalized inverse.

General inverse kinematics solution . . .

$$\dot{q} = \mathbf{J}^\dagger \dot{x} + (\mathbf{I}_n - \mathbf{J}^\dagger \mathbf{J}) \dot{\varphi}$$

- The term $(\mathbf{I}_n - \mathbf{J}^\dagger \mathbf{J}) \dot{\varphi}$ is the **homogenous** solution and serves to purely reconfigure the robot arm without affecting the task.
- The **homogenous** solution is typically used to achieve some additional goals, i.e. different joint velocities \dot{q} can be obtained, which result in the same end-effector velocity \dot{x} . Typically, it is used for
 - obstacle avoidance
 - some kind of optimization
 - singularity avoidance
 - joint limits avoidance
 - pose control
 - . . .

Some performance measures

- Manipulability

- A commonly used measure is the manipulability measure defined as

$$w = \sqrt{\sigma_1 \sigma_2 \cdots \sigma_m} = \sqrt{\det(\mathbf{J}\mathbf{J}^T)}$$

- Condition number

- The condition number ρ is the ratio between the maximal and the minimal singular value of \mathbf{J} .

$$\rho = \sqrt{\frac{\sigma_{\max}}{\sigma_{\min}}} \quad \nabla \rho = \frac{1}{2\rho} \frac{\sigma_{\min} \nabla \sigma_{\max} - \sigma_{\max} \nabla \sigma_{\min}}{\sigma_{\min}^2}$$

- Gravity torques norm

- Considering only the gravity, the performance measure p representing the weighted norm of joint torques can be expressed as

$$p = \mathbf{g}(\mathbf{q})^T \mathbf{W} \mathbf{g}(\mathbf{q}) \quad \nabla p(\mathbf{q}) = 2 \left(\frac{\partial \mathbf{g}(\mathbf{q})}{\partial \mathbf{q}} \right)^T \mathbf{W} \mathbf{g}(\mathbf{q})$$

Control at the kinematic level

- For velocity control the following **kinematic controller** can be used:

$$\dot{q}_c = \mathbf{J}^+ \dot{x}_c + \mathbf{N} \dot{\varphi} \quad \mathbf{N} = (\mathbf{I} - \mathbf{J}^+ \mathbf{J})$$

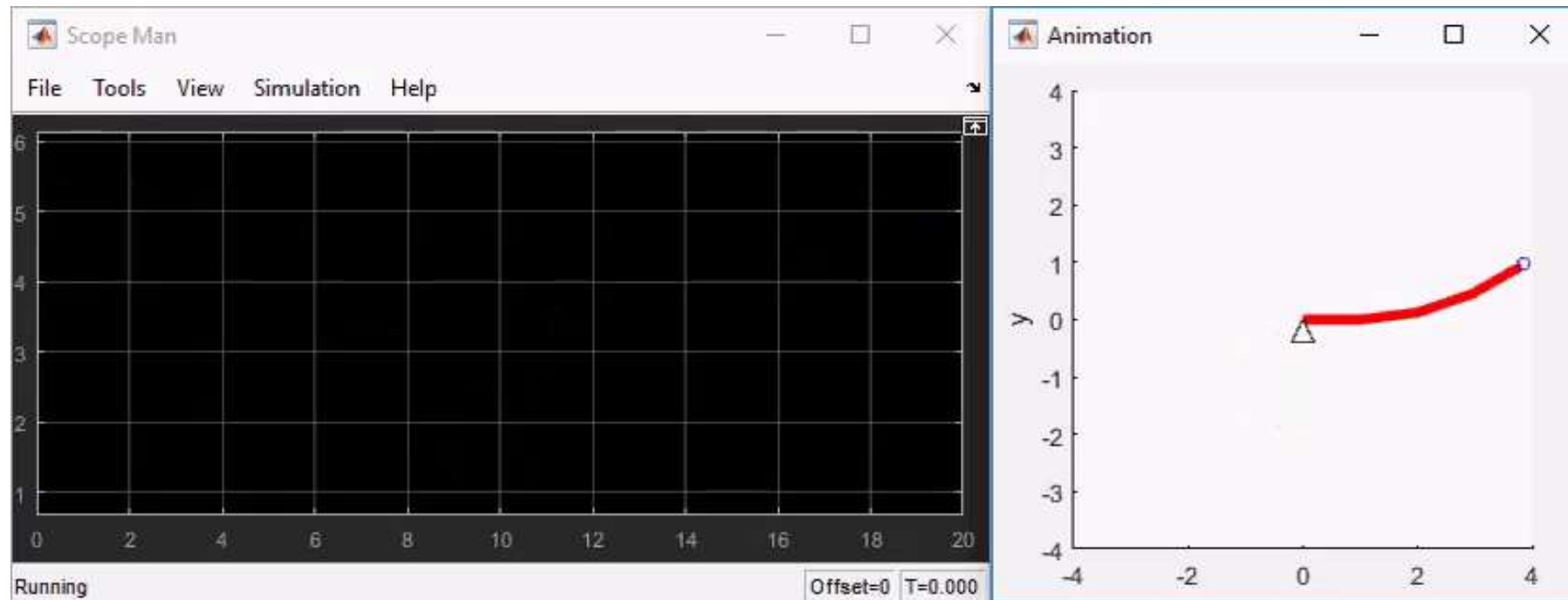
- Primary task: end-effector position \dot{x}_c :

$$\dot{x}_c = \dot{x}_E + \mathbf{K}_p (\mathbf{x}_E - \mathbf{x})$$

- Secondary tasks: we use joint velocities $\dot{\varphi}$ (self motion).

Example: Kinematic control - Planar 4R

- Task space: PTP motion, kinematic control using Moore-Penrose pseudoinverse.
- Null-space: Optimization of robot pose by maximizing manipulability (avoiding singular configurations)



Control at the dynamic level

- Using the acceleration formulation we can use the following **dynamic controller**

$$\boldsymbol{\tau} = \mathbf{H}(\bar{\mathbf{J}}(\ddot{\mathbf{x}}_c - \dot{\mathbf{J}}\dot{\mathbf{q}}) + \bar{\mathbf{N}}(\boldsymbol{\phi} + \dot{\mathbf{J}}\dot{\mathbf{x}}) + \mathbf{h} + \mathbf{g})$$

- Primary task:** end-effector acceleration (position)

$$\ddot{\mathbf{x}}_c = \ddot{\mathbf{x}}_d + \mathbf{K}_v\dot{\mathbf{e}} + \mathbf{K}_p\mathbf{e}$$

- Secondary tasks:** we use joint velocities $\dot{\boldsymbol{\phi}}$ (self motion)

$$\boldsymbol{\phi} = \ddot{\boldsymbol{\phi}} + \mathbf{K}_n\bar{\mathbf{N}}(\dot{\boldsymbol{\phi}} - \dot{\mathbf{q}})$$

- By selecting proper controller parameters the following dynamic properties can be achieved

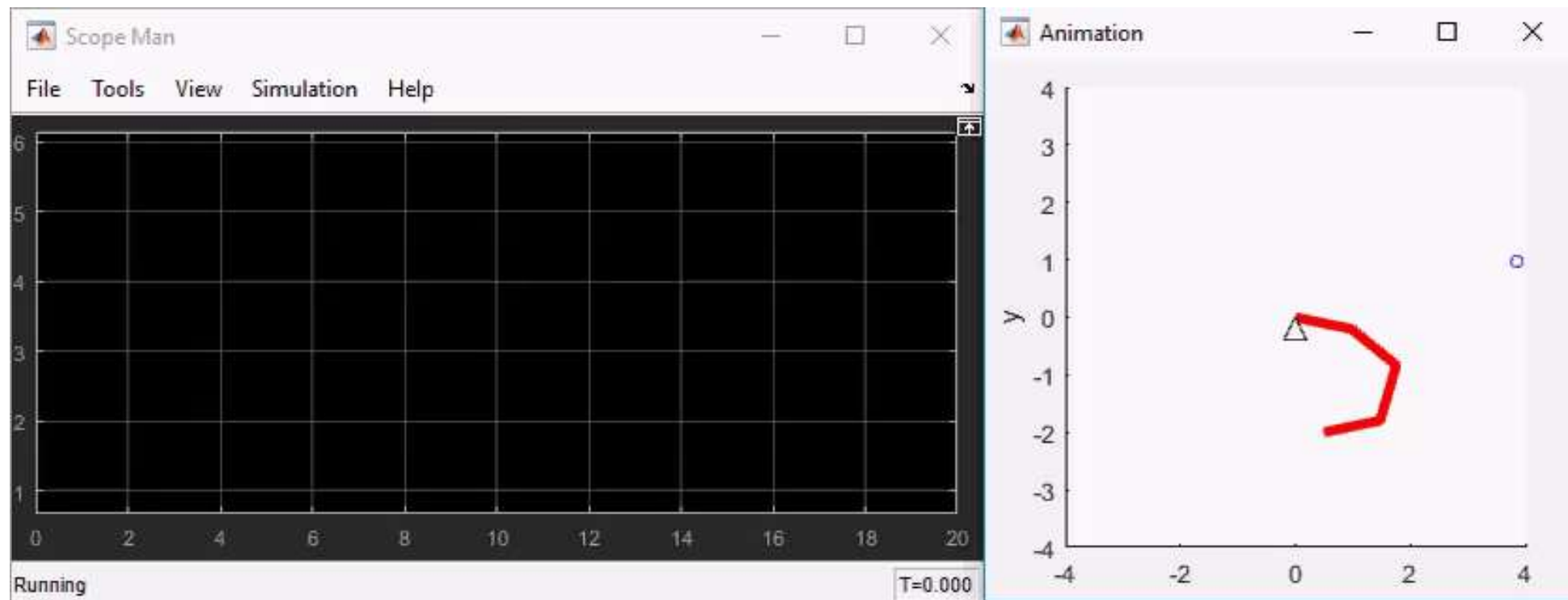
$$\Lambda\ddot{\mathbf{e}} + \Lambda\mathbf{K}_v\dot{\mathbf{e}} + \Lambda\mathbf{K}_p\mathbf{e} = -\mathbf{F}$$

$$\mathbf{H}_n\ddot{\mathbf{e}}_n + \mathbf{H}_n\mathbf{K}_n\dot{\mathbf{e}}_n = -\bar{\mathbf{N}}^T\boldsymbol{\tau}_F$$

- Effective inertia matrix in \mathbf{N} : $\mathbf{H}_n = \bar{\mathbf{N}}^T\mathbf{H}\bar{\mathbf{N}} = \mathbf{H} - \mathbf{J}^T\Lambda\mathbf{J}$

Example: Dynamic control - Planar 4R

- Task space: PTP motion, dynamic control using inertia-weighted pseudoinverse.
- Null-space: Optimization of robot pose by maximizing manipulability (avoiding singular configurations)

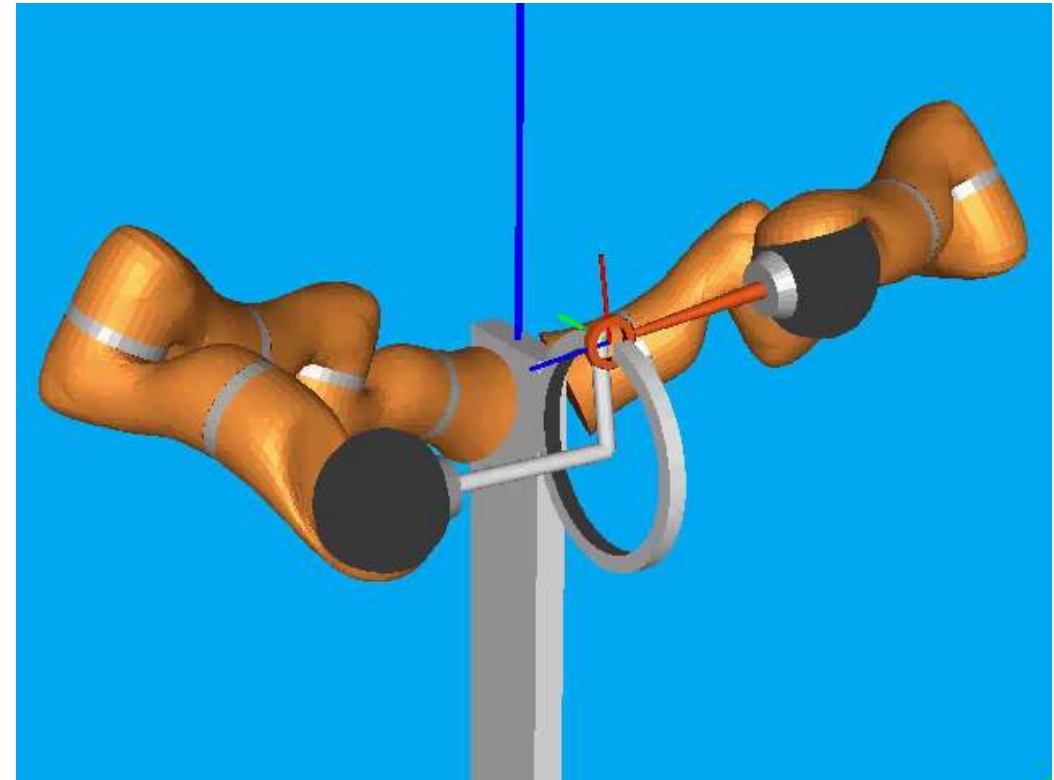


Control of functional redundant robots

- Some task do not require controlled motion in all spatial directions.
- Example: The motion of the ring is free around the hoop, so we can remove rotation around the x -axis from the task control.

$$\dot{\mathbf{x}} = \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{z} \\ \omega_x \\ \omega_y \\ \omega_z \end{bmatrix}$$

- The problem is when the ring is moved along the hoop and the control is not adequate anymore.

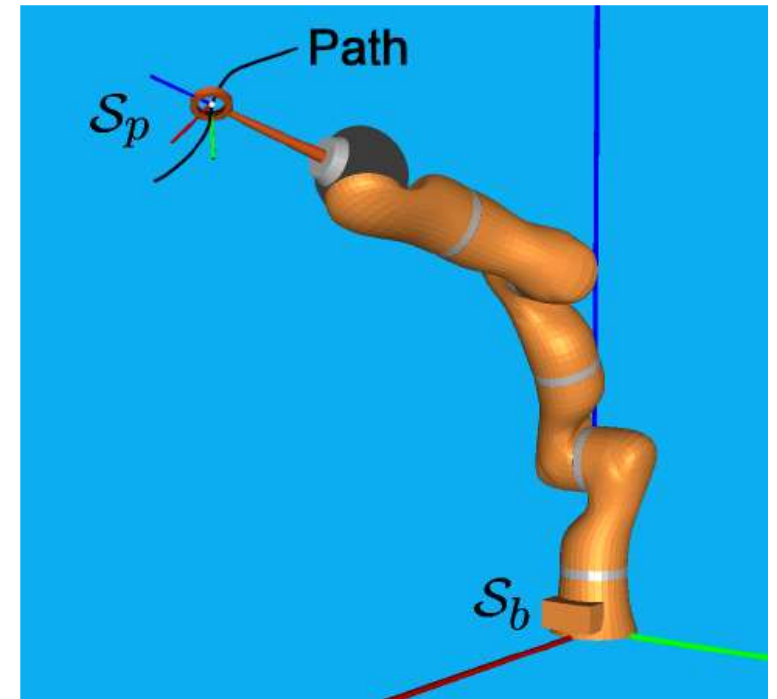


Control of functional redundant robots . . .

- To exploit the available functional redundancy it is necessary to and a task frame where the **redundant DOFs are rows of the Jacobian matrix**.
- If the task frame is changing along the task path, we have to consider this in the control.
- Mapping between the path frame \mathcal{S}_p and base frame \mathcal{S}_b (only rotation)

$$\tilde{\mathbf{R}}_t = \begin{bmatrix} \mathbf{R}_t & \mathbf{0}_{3 \times 3} \\ \mathbf{0}_{3 \times 3} & \mathbf{R}_t \end{bmatrix}$$

- We map the control into the workspace which is anchored on the path



$$\dot{\mathbf{q}}_c = (\tilde{\mathbf{R}}_t^T \mathbf{J})^\# \left(\tilde{\mathbf{R}}_t^T \begin{bmatrix} \mathbf{K}_p \mathbf{e}_p + \dot{\mathbf{p}}_d \\ \mathbf{K}_o \mathbf{e}_o + \boldsymbol{\omega}_d \end{bmatrix} \right) + (\mathbf{I} - (\tilde{\mathbf{R}}_t^T \mathbf{J})^\# \tilde{\mathbf{R}}_t^T \mathbf{J}) \dot{\mathbf{q}}_n,$$

Control of functional redundant robots . . .

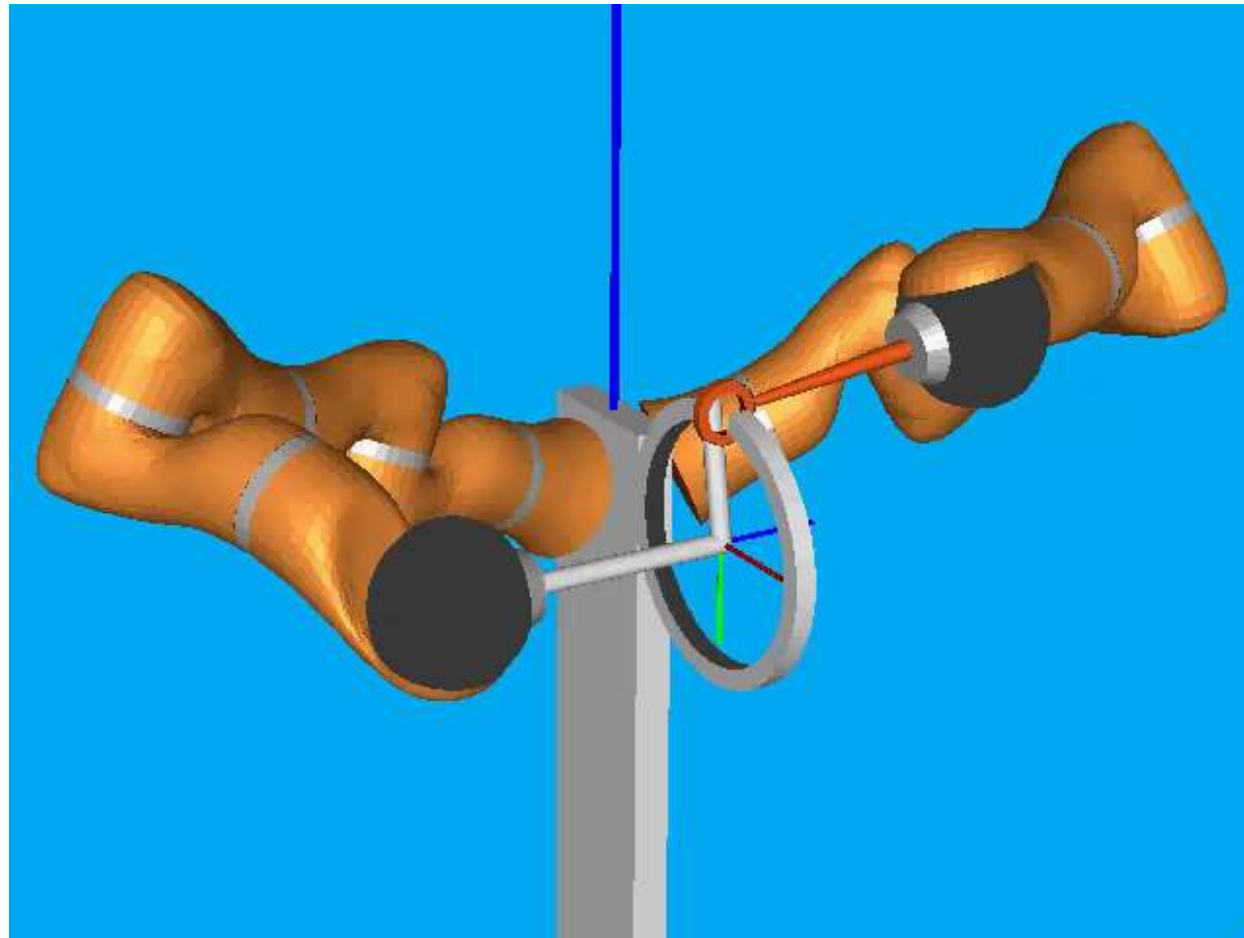
$$\dot{q}_c = \mathbf{J}^\dagger \begin{bmatrix} \mathbf{K}_p e_p + \dot{p}_d \\ \mathbf{K}_o e_o + \dot{\omega}_d \end{bmatrix}$$

- Let assume that for the tasks the linear motion in direction of y-axis and orientation around z-axis is not important.

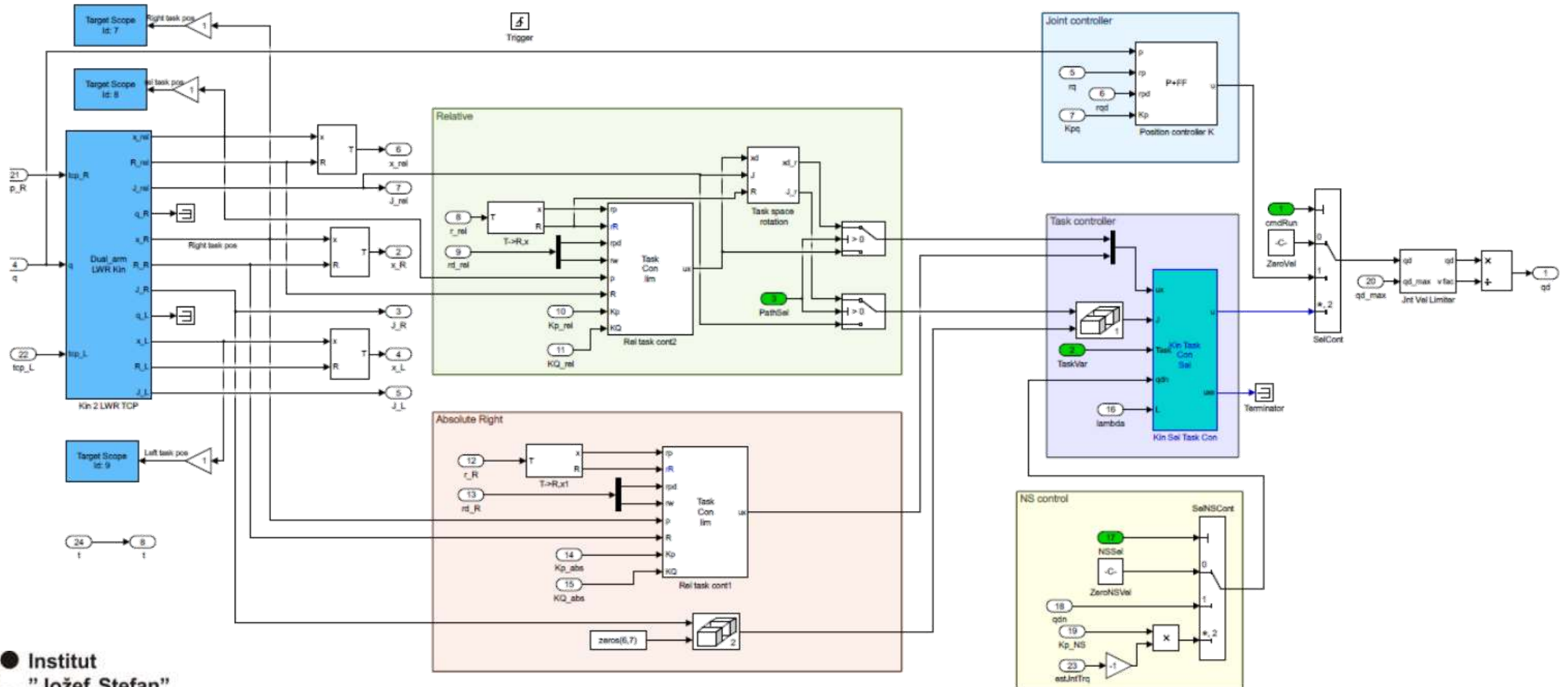
$$\dot{q}_c = \begin{bmatrix} J_{11} & \cdots & J_{1n} \\ J_{21} & \cdots & J_{2n} \\ J_{31} & \cdots & J_{3n} \\ J_{41} & \cdots & J_{4n} \\ J_{51} & \cdots & J_{5n} \\ J_{61} & \cdots & J_{6n} \end{bmatrix}^\# \left[\begin{array}{c} \mathbf{K}_p \begin{bmatrix} e_{p,x} \\ e_{p,y} \\ e_{p,x} \end{bmatrix} + \begin{bmatrix} \dot{p}_{d,x} \\ \dot{p}_{d,y} \\ \dot{p}_{d,z} \end{bmatrix} \\ \mathbf{K}_o \begin{bmatrix} e_{o,x} \\ e_{o,y} \\ e_{o,z} \end{bmatrix} + \begin{bmatrix} \dot{\omega}_{d,x} \\ \dot{\omega}_{d,y} \\ \dot{\omega}_{d,z} \end{bmatrix} \end{array} \right]$$

Control in the path space

- Free DOF is the rotation around path (rotation axis is in the direction of y-axis of path space - y-axis of the \mathcal{S}_p connected to the path).



Kuka dual-arm controller



LWR hoop - Obstacle avoidance

