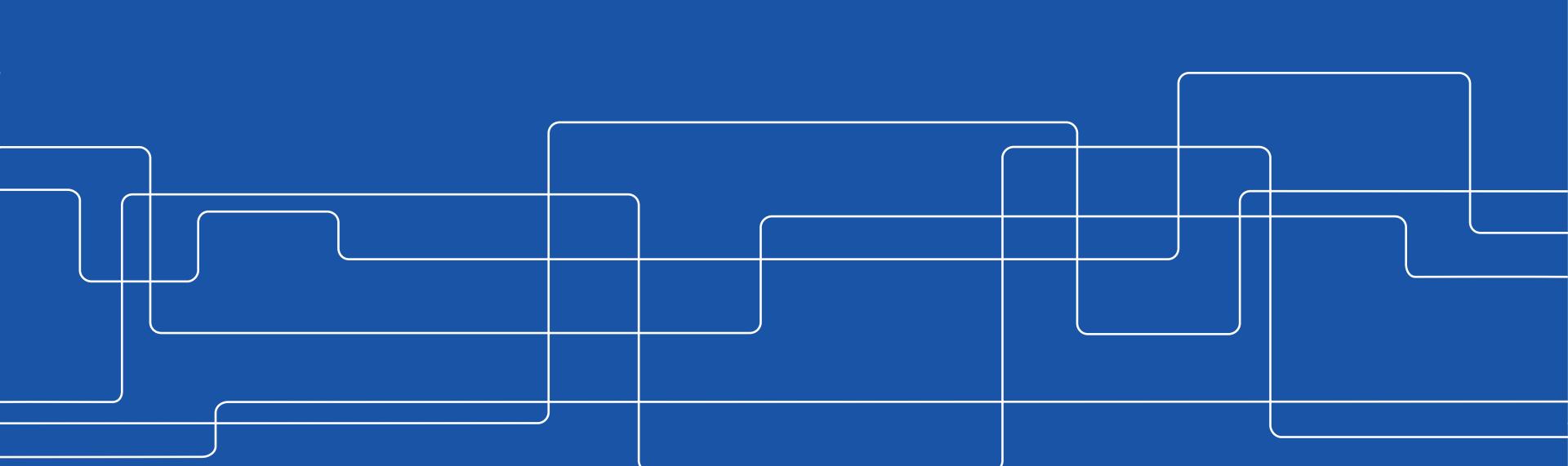




# Introduction to Robotics

DD2410 - Introduction to Robotics

Lecture 4 - Differential Kinematics & Dynamics





## Schedule - Lectures

Aug 30 - 1. Intro, Course fundamentals, Topics, What is a Robot, History, Applications.

Aug 31 - 3 ROS Introduction

Aug 31 - 2 Manipulators, Kinematics

**Sep 07 - 4. Differential kinematics, dynamics**

Sep 09 - 5. Actuators, sensors I (force, torque, encoders, ...)

Sep 12 - 6. Grasping, Motion, Control

Sep 14 - 7. Planning (RRT, A\*, ...)

Sep 19 - 8. Behavior Trees and Task Switching

Sep 21 - 9. Mobility and sensing II (distance, vision, radio, GPS, ...)

Sep 26 - 10. Localisation (where are we?)

Sep 28 - 11. Mapping (how to build the map to localise/navigate w.r.t.?)

Oct 03 - 12. Navigation (how do I get from A to B?)

Oct 05 - Q/A - Open questions to your teachers.



# Overview

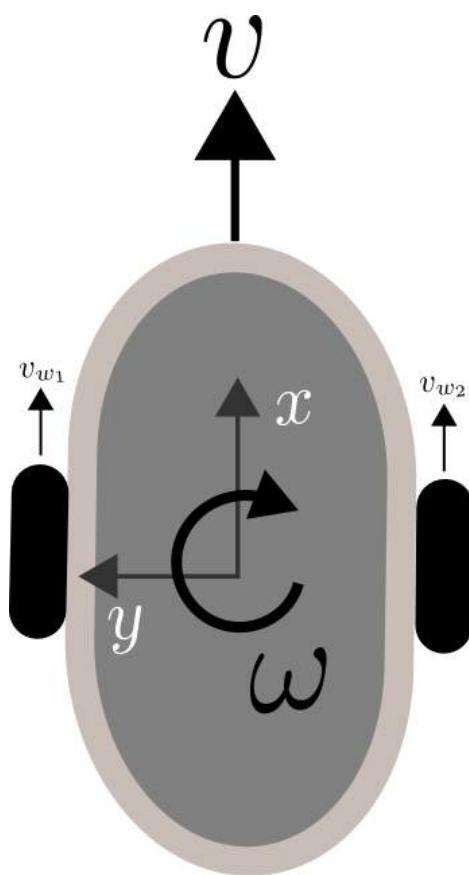
- Differential kinematics
  - Jacobians
  - Singularities
  - Manipulability
  - Calculations
- Dynamics
  - Forces and accelerations
  - algorithms for calculations



## Differential kinematics

- For many operations, we are not interested in the stationary kinematics, but rather the differential kinematics, mainly for the mapping between velocities in configuration space and cartesian space

# Differential kinematics - Vacuum cleaner type



$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ \omega \end{bmatrix}$$

$$v = \frac{v_{w_1} + v_{w_2}}{2}$$

$$\omega = \frac{v_{w_2} - v_{w_1}}{2b}$$

$$v_{w_i} = \frac{2\pi r f \Delta_{\text{enc}}}{\text{ticks per rev}}$$



## Differential kinematics

- The instantaneous transform between velocities in robot configuration space and cartesian space is given by the Jacobian:

$$\dot{X} = J(\Theta) \dot{\Theta}$$

- Where each element  $j_{mn}$  in  $J$  is defined as  $\frac{\partial K(\Theta)_m}{\partial \Theta_n}$



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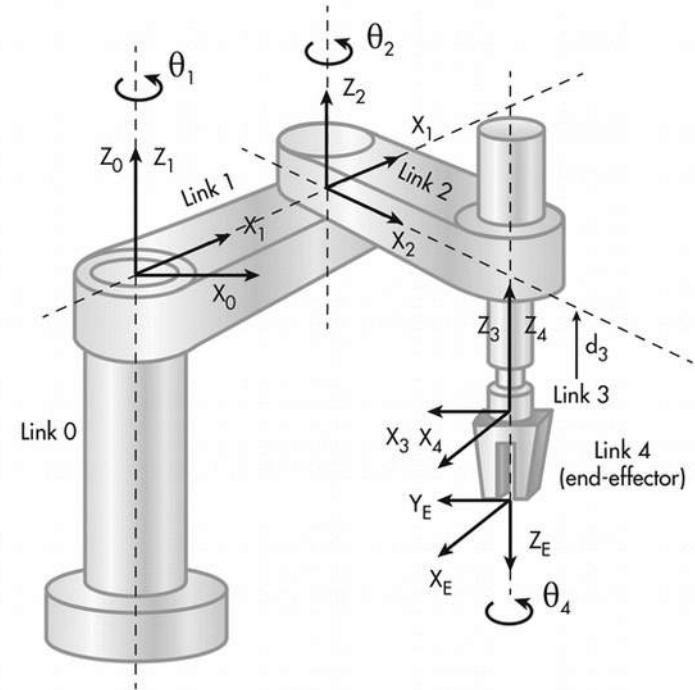
## Forward kinematics

- Transform  ${}^0T_E$  from end effector to base frame is dependant on configuration  $\Theta$
- The function that generates the end effector pose  $X$  given  $\Theta$ , is called **forward kinematics, K**

$$X = K(\Theta), \quad r = {}^0T_E p_E$$

where  $p$  is the position of the endpoint in the last frame

- Commonly, we define  $K(\Theta)$  to output the pose vector  $X = [x \ y \ z \ \alpha \ \beta \ \gamma]^T$ , where  $\alpha \ \beta \ \gamma$  are the *Euler Angles*



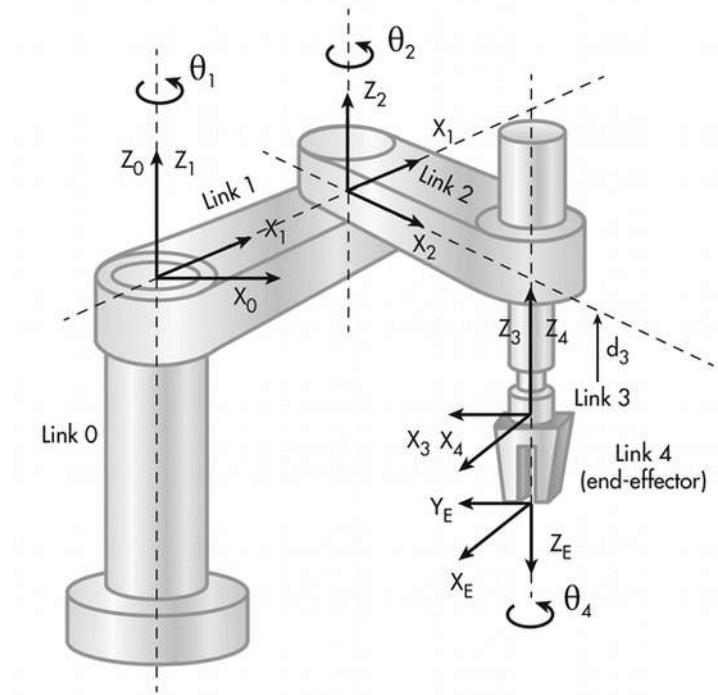
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$${}^0T_E = \begin{pmatrix} R & t \\ 0 & 1 \end{pmatrix}$$

See R-MPC 2.4



## Differential kinematics

- The instantaneous transform between velocities in robot configuration space and cartesian space is given by the Jacobian:

$$\dot{X} = J(\Theta) \dot{\Theta}$$

- Where each element  $j_{mn}$  in  $J$  is defined as  $\frac{\partial K(\Theta)_m}{\partial \theta_n}$
- Thus, each column in  $J$  can be seen as the vector  $\Delta X_i$ , or the motion in  $X$  caused by motion in the joint  $\theta_i$ .



## Differential kinematics

- The closed form of a typical manipulator Jacobian is not printable

# Differential kinematics

- The closed form of a typical manipulator Jacobian is not printable

The Puma 560 can be seen in Figures 1 and 2.

The forward kinematics  $K_f$  can be formulated as:

$$\mathbf{X} = K_f(\boldsymbol{\Theta}) \quad (1)$$

where

$$\mathbf{X} = \begin{bmatrix} x \\ y \\ z \\ p \\ t \\ a \end{bmatrix}, \quad \boldsymbol{\Theta} = \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \\ \theta_4 \\ \theta_5 \\ \theta_6 \end{bmatrix} \quad (2)$$

we have:

$$\begin{aligned} x &= \cos(\theta_1) * [a_2\cos(\theta_2) + a_3\cos(\theta_2 + \theta_3) - d_4\sin(\theta_2 + \theta_3)] - d_3\sin(\theta_1) \\ y &= \sin(\theta_1) * [a_2\cos(\theta_2) + a_3\cos(\theta_2 + \theta_3) - d_4\sin(\theta_2 + \theta_3)] + d_3\cos(\theta_1) \\ z &= -a_3\sin(\theta_2 + \theta_3) + a_2\sin(\theta_2) - d_4\cos(\theta_2 + \theta_3) \\ p &= \tan^{-1}\left(\frac{s_1(c_23c_4s_5+s_23c_5)-c_1s_4s_5}{c_1(c_23c_4s_5+s_23c_5)+s_1s_4s_5}\right) \\ t &= \tan^{-1}\left(\frac{-s_1(c_23c_4s_5+s_23c_5)+c_1s_4s_5}{\sin(\tan^{-1}(-s_1(c_23c_4s_5+s_23c_5)+c_1s_4s_5)-c_1(c_23c_4s_5+s_23c_5)-s_1s_4s_5)(s_23c_4s_5-c_23c_5)}\right) \\ a &= \tan^{-1}\left(\frac{-(s_{23}(s_4c_6-c_4s_5s_6)+c_{23}s_5s_6)}{s_{23}(s_4s_6-c_4c_5s_6)-c_{23}s_5c_6}\right) \end{aligned} \quad (3)$$

Where the latter uses shorthand. The full expression is:

$$\begin{aligned} r &= \tan^{-1}\left(\frac{\sin(\theta_1)(\cos(\theta_2+\theta_3)\cos(\theta_4)\sin(\theta_5)+\sin(\theta_2+\theta_3)\cos(\theta_5)-\cos(\theta_1)\sin(\theta_4)\sin(\theta_5))}{\cos(\theta_1)(\cos(\theta_2+\theta_3)\cos(\theta_4)\sin(\theta_5)+\sin(\theta_2+\theta_3)\cos(\theta_5))+\sin(\theta_2+\theta_3)\cos(\theta_5)}\right) \\ t &= \tan^{-1}\left(\frac{-\sin(\theta_1)(\cos(\theta_2+\theta_3)\cos(\theta_4)\sin(\theta_5)+\sin(\theta_2+\theta_3)\cos(\theta_5)+\cos(\theta_1)\sin(\theta_4)\sin(\theta_5))+\cos(\theta_1)\sin(\theta_4)\sin(\theta_5)}{\sin\left(\tan^{-1}\left(\frac{-\sin(\theta_1)(\cos(\theta_2+\theta_3)\cos(\theta_4)\sin(\theta_5)+\sin(\theta_2+\theta_3)\cos(\theta_5)+\cos(\theta_1)\sin(\theta_4)\sin(\theta_5)}{\cos(\theta_1)(\cos(\theta_2+\theta_3)\cos(\theta_4)\sin(\theta_5)+\sin(\theta_2+\theta_3)\cos(\theta_5)+\sin(\theta_1)\sin(\theta_4)\sin(\theta_5)}\right)\right)(\sin(\theta_2+\theta_3)\cos(\theta_4)\sin(\theta_5)-\cos(\theta_2+\theta_3)\cos(\theta_5))}\right) \\ a &= \tan^{-1}\left(\frac{-(\sin(\theta_2+\theta_3)\sin(\theta_4)\cos(\theta_5)-\cos(\theta_4)\cos(\theta_5)\cos(\theta_6)-\cos(\theta_2+\theta_3)\sin(\theta_5)\cos(\theta_6))}{\sin(\theta_2+\theta_3)(\sin(\theta_4)\sin(\theta_6)-\cos(\theta_4)\cos(\theta_5)\cos(\theta_6)-\cos(\theta_2+\theta_3)\sin(\theta_5)\cos(\theta_6)})\right) \end{aligned} \quad (4)$$

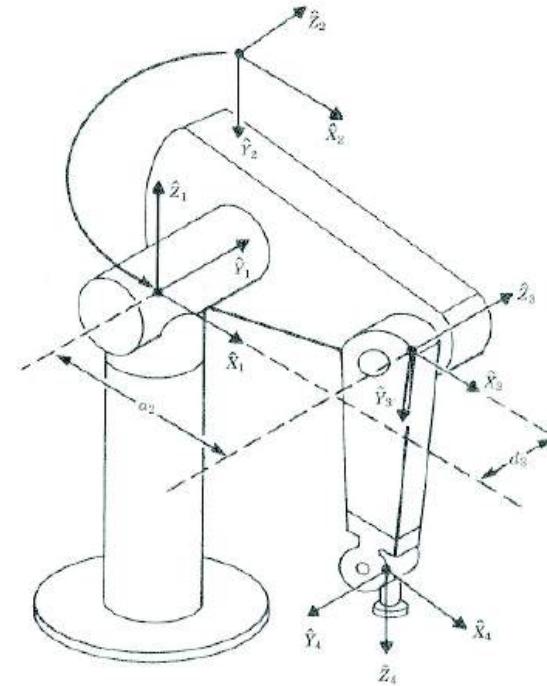


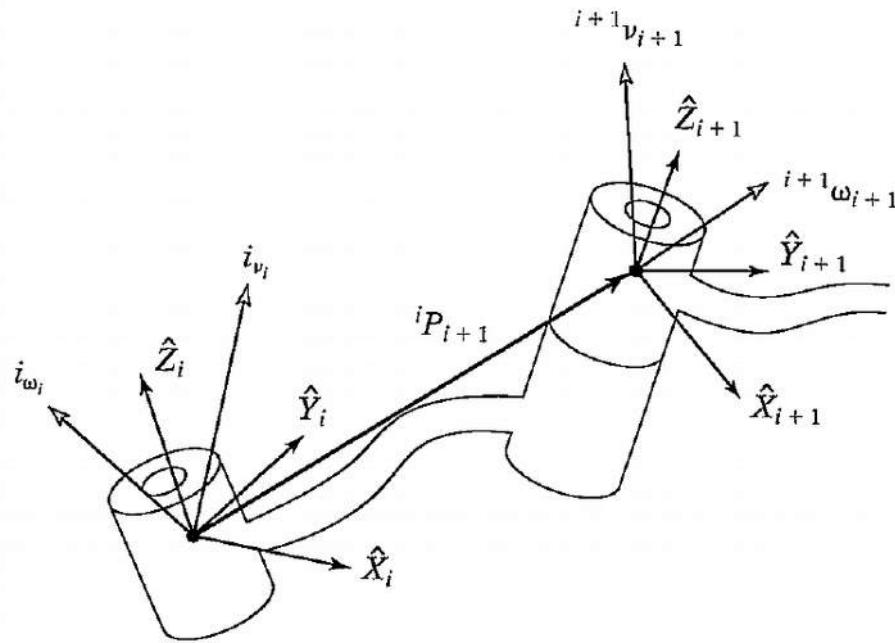
Figure 1: The puma 560



## Differential kinematics (J.J. Craig chapter 5)

- The closed form of a typical manipulator Jacobian is often not printable, but can be derived by sequential application of frame transforms
- The motion of frame  $i+1$ , is a function of the motion of frame  $i$  and the motion of the joint between them.

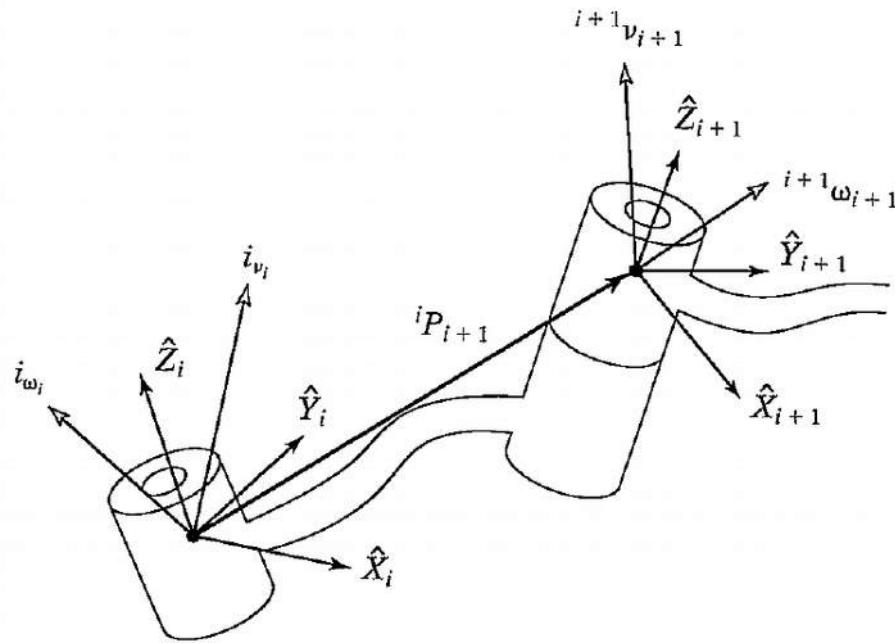
## Differential kinematics (R-MPC chapter 3): Rotational joints



$${}^{i+1}\omega_{i+1} = {}^{i+1}_i R {}^i\omega_i + \dot{\theta}_{i+1} {}^{i+1}\hat{Z}_{i+1}$$

$${}^{i+1}v_{i+1} = {}^{i+1}_i R ({}^i v_i + {}^i\omega_i \times {}^i P_{i+1})$$

## Differential kinematics (R-MPC chapter 3) - Prismatic joints



$${}^{i+1}\omega_{i+1} = {}^{i+1}_i R {}^i\omega_i,$$

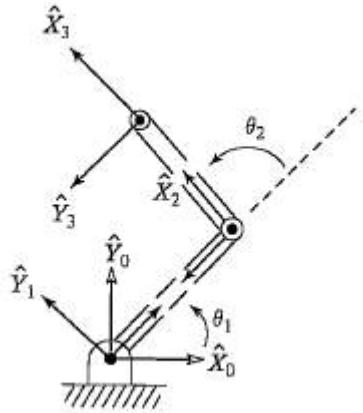
$${}^{i+1}v_{i+1} = {}^{i+1}_i R ({}^i v_i + {}^i\omega_i \times {}^i P_{i+1}) + \dot{d}_{i+1} {}^{i+1}\hat{Z}_{i+1}$$



## Differential kinematics (J.J. Craig chapter 5)

- Consequetive application of link transforms gives us velocities in end effector frame
- Note: resulting velocities are multilinear in joint velocities!
- Multiplying by rotation transform  ${}^B R_E$  gives us velocities in base frame
- Thus we can derive  $J(\Theta)$

## Example: Planar robot



$${}^1\omega_1 = \begin{bmatrix} 0 \\ 0 \\ \dot{\theta}_1 \end{bmatrix},$$

$${}^1v_1 = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix},$$

$${}^2\omega_2 = \begin{bmatrix} 0 \\ 0 \\ \dot{\theta}_1 + \dot{\theta}_2 \end{bmatrix},$$

$${}^2v_2 = \begin{bmatrix} c_2 & s_2 & 0 \\ -s_2 & c_2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ l_1\dot{\theta}_1 \\ 0 \end{bmatrix} = \begin{bmatrix} l_1s_2\dot{\theta}_1 \\ l_1c_2\dot{\theta}_1 \\ 0 \end{bmatrix},$$

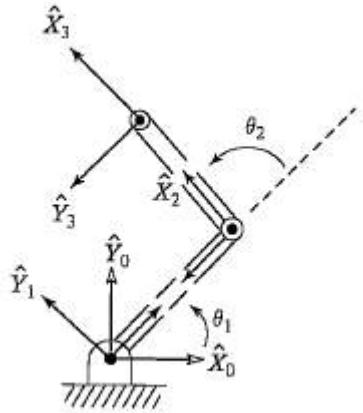
$${}^3\omega_3 = {}^2\omega_2,$$

$${}^3v_3 = \begin{bmatrix} l_1s_2\dot{\theta}_1 \\ l_1c_2\dot{\theta}_1 + l_2(\dot{\theta}_1 + \dot{\theta}_2) \\ 0 \end{bmatrix}.$$

$${}^0R = {}^0R \quad {}^1R \quad {}^2R = \begin{bmatrix} c_{12} & -s_{12} & 0 \\ s_{12} & c_{12} & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$${}^0v_3 = \begin{bmatrix} -l_1s_1\dot{\theta}_1 - l_2s_{12}(\dot{\theta}_1 + \dot{\theta}_2) \\ l_1c_1\dot{\theta}_1 + l_2c_{12}(\dot{\theta}_1 + \dot{\theta}_2) \\ 0 \end{bmatrix}.$$

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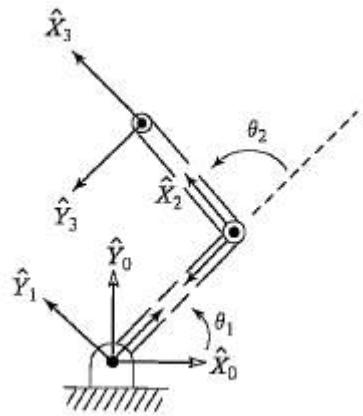
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$${}^0J(\Theta) = \begin{bmatrix} -l_1s_1 - l_2s_{12} & -l_2s_{12} \\ l_1c_1 + l_2c_{12} & l_2c_{12} \end{bmatrix}$$

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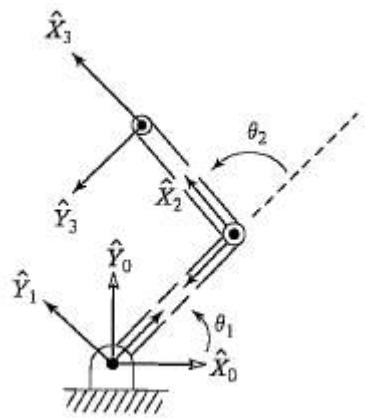
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$$\dot{X} = J(\Theta)\dot{\Theta}$$

$$\dot{X} = \begin{bmatrix} -l_1 s_1 - l_2 s_{12} & -l_2 s_{12} \\ l_1 c_1 + l_2 c_{12} & l_2 c_{12} \end{bmatrix} \begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \end{bmatrix},$$



## Jacobians (R-MPC 3.1.3)

$$\dot{\mathbf{X}} = \mathbf{J}(\Theta) \dot{\Theta}$$

- Column  $j_i$  in  $\mathbf{J}$  is the contribution of the  $i$ :th joint to the velocity of the end effector.
- Each column in  $\mathbf{J}$  can be computed individually.



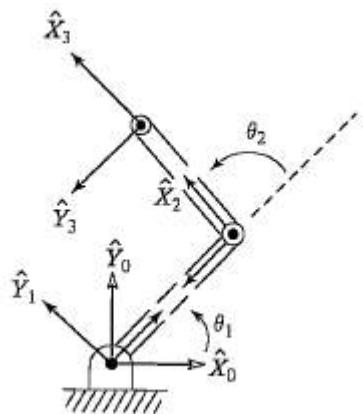
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Assignment 2: Second part

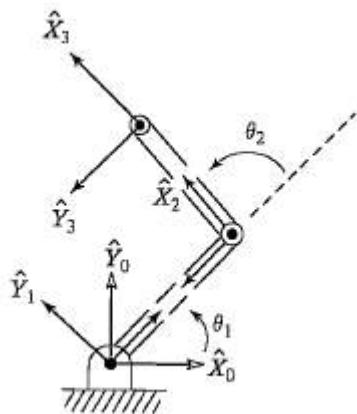
## Example: Planar robot



$${}^0 J(\Theta) = \begin{bmatrix} -l_1 s_1 - l_2 s_{12} & -l_2 s_{12} \\ l_1 c_1 + l_2 c_{12} & l_2 c_{12} \end{bmatrix}$$

What happens if both angles are 0?

## Example: Planar robot

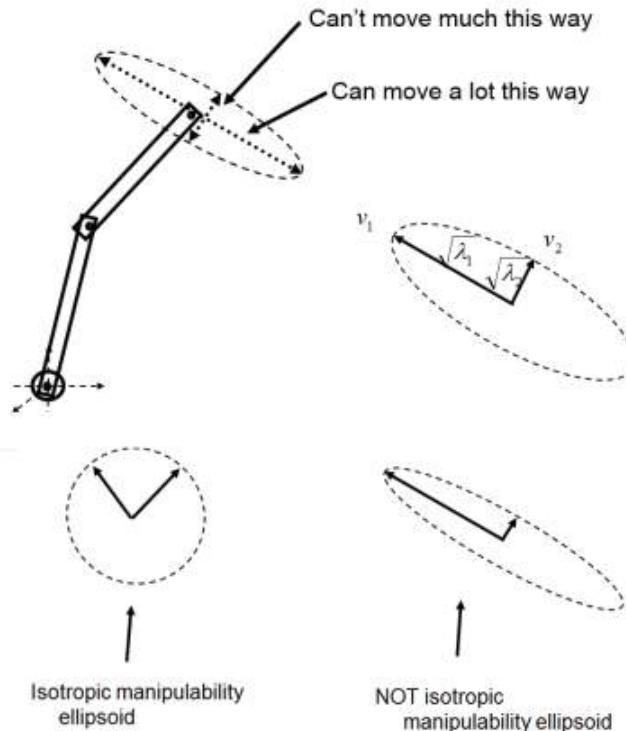


$$\dot{\mathbf{x}} = \begin{bmatrix} -l_1 s_1 - l_2 s_{12} & -l_2 s_{12} \\ l_1 c_1 + l_2 c_{12} & l_2 c_{12} \end{bmatrix} \begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \end{bmatrix}$$

What happens if both angles are 0?

When the Jacobian loses rank, we get a kinematic singularity - we lose the ability to generate motion in some direction!

- We can generalize this into a concept of manipulability w

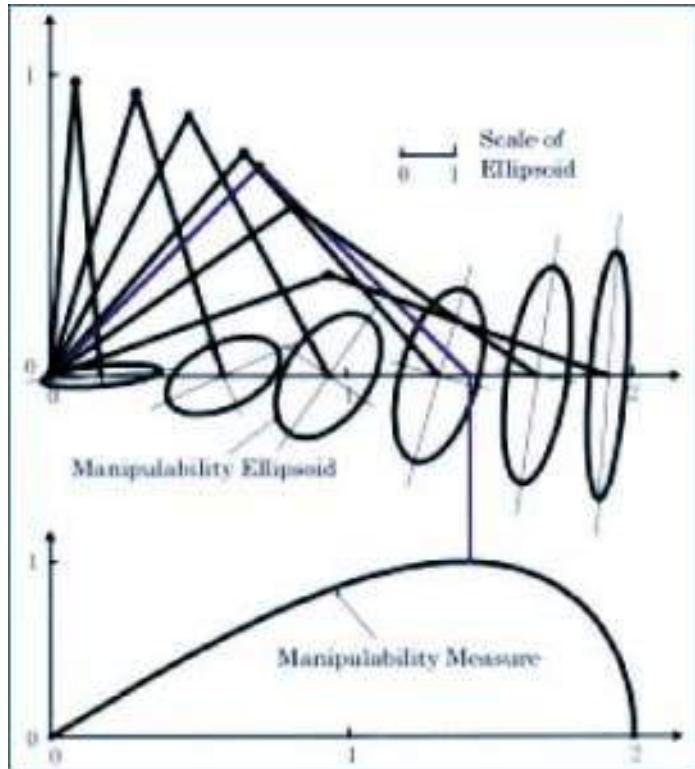


$$w = \sqrt{\det(JJ^T)}$$

w is proportional to the volume of the *manipulability ellipsoid*.

# Manipulability

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w is proportional to the volume of the *manipulability ellipsoid*.

## Manipulability example

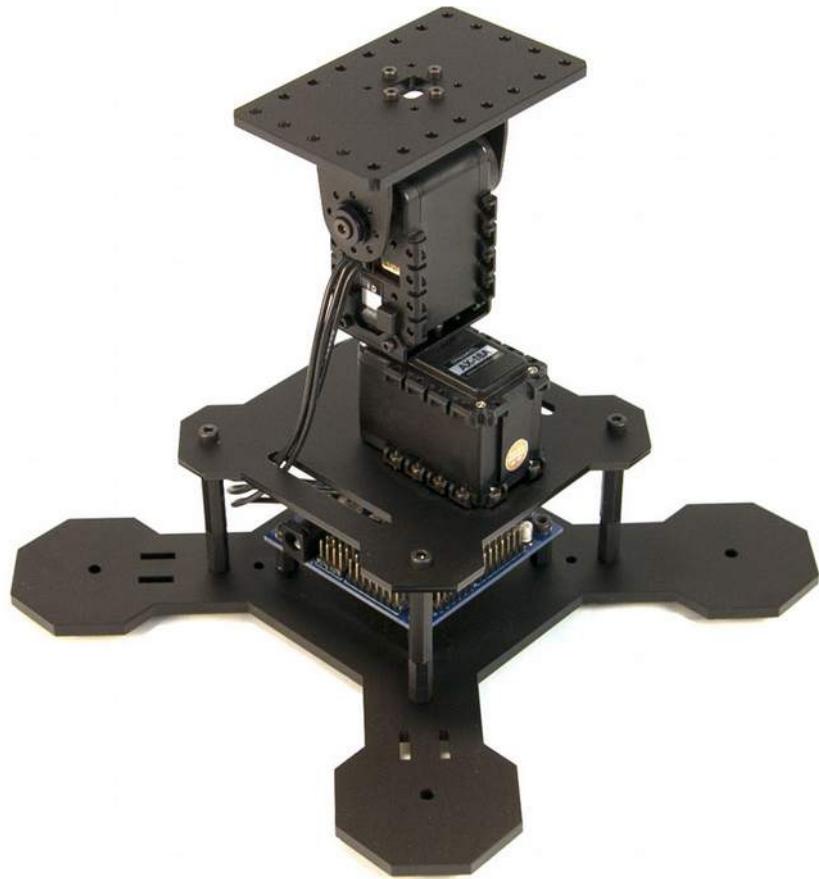


Image: Trossen Robotics



## Jacobians

- The inverse Jacobian is trivial to calculate, as long as the Jacobian matrix is invertible.
- If  $J$  is not invertible, we can often use pseudo-inverse instead.



## Jacobians for numerical inverse kinematics

- We want to find the inverse kinematics

$$\Theta = K^{-1}(X)$$



## Jacobians for numerical inverse kinematics

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$$\Theta = K^{-1}(X)$$

- We start with an approximation

$$\hat{\Theta} = \Theta + \epsilon_{\Theta}$$



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$$X + \epsilon_X = K(\Theta + \epsilon_{\Theta})$$



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- We want to find the inverse kinematics

$$\Theta = K^{-1}(X)$$

- We start with an approximation

$$\hat{\Theta} = \Theta + \epsilon_{\Theta}$$

$$X + \epsilon_X = K(\Theta + \epsilon_{\Theta})$$

$$K(\Theta) + \epsilon_X = K(\Theta + \epsilon_{\Theta})$$



## Jacobians for numerical inverse kinematics

- We want to find the inverse kinematics

$$\Theta = K^{-1}(X)$$

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$$\hat{\Theta} = \Theta + \epsilon_{\Theta}$$

$$X + \epsilon_X = K(\Theta + \epsilon_{\Theta})$$

$$K(\Theta) + \epsilon_X = K(\Theta + \epsilon_{\Theta})$$

- With linear approximation, we get

$$\epsilon_X \approx J(\Theta) \epsilon_{\Theta}$$



## Jacobians for numerical inverse kinematics

- We want to find the inverse kinematics

$$\Theta = K^{-1}(X)$$

- We start with an approximation

$$\hat{\Theta} = \Theta + \epsilon_{\Theta}$$

$$X + \epsilon_X = K(\Theta + \epsilon_{\Theta})$$

$$K(\Theta) + \epsilon_X = K(\Theta + \epsilon_{\Theta})$$

- With linear approximation, we get (assuming invertible J)

$$\epsilon_X \approx J(\Theta) \epsilon_{\Theta}$$

$$\epsilon_{\Theta} \approx J^{-1}(\Theta) \epsilon_X$$



## Jacobians for numerical inverse kinematics

- **Algorithm for finding inverse kinematics**

Given target  $\mathbf{X}$  and initial approximation  $\hat{\Theta}$



## Jacobians for numerical inverse kinematics

- **Algorithm for finding inverse kinematics**

Given target  $\mathbf{X}$  and initial approximation  $\hat{\Theta}$

repeat

until  $\epsilon_X \leq tolerance$



## Jacobians for numerical inverse kinematics

- **Algorithm for finding inverse kinematics**

Given target  $\mathbf{X}$  and initial approximation  $\hat{\Theta}$

repeat

$$\hat{\mathbf{X}} := K(\hat{\Theta})$$

until  $\epsilon_X \leq tolerance$



## Jacobians for numerical inverse kinematics

- **Algorithm for finding inverse kinematics**

Given target  $\mathbf{X}$  and initial approximation  $\hat{\Theta}$

repeat

$$\hat{\mathbf{X}} := K(\hat{\Theta})$$

$$\epsilon_X := \hat{\mathbf{X}} - \mathbf{X}$$

until  $\epsilon_X \leq tolerance$



## Jacobians for numerical inverse kinematics

- **Algorithm for finding inverse kinematics**

Given target  $\mathbf{X}$  and initial approximation  $\widehat{\boldsymbol{\Theta}}$

repeat

$$\widehat{\mathbf{X}} := \mathbf{K}(\widehat{\boldsymbol{\Theta}})$$

$$\boldsymbol{\epsilon}_x := \widehat{\mathbf{X}} - \mathbf{X}$$

$$\boldsymbol{\epsilon}_{\boldsymbol{\Theta}} := \mathbf{J}^{-1}(\widehat{\boldsymbol{\Theta}}) \boldsymbol{\epsilon}_x$$

until  $\boldsymbol{\epsilon}_x \leq tolerance$



## Jacobians for numerical inverse kinematics

- **Algorithm for finding inverse kinematics**

Given target  $\mathbf{X}$  and initial approximation  $\hat{\Theta}$

repeat

$$\hat{\mathbf{X}} := \mathbf{K}(\hat{\Theta})$$

$$\epsilon_x := \hat{\mathbf{X}} - \mathbf{X}$$

$$\epsilon_\Theta := \mathbf{J}^{-1}(\hat{\Theta}) \epsilon_x$$

$$\hat{\Theta} := \hat{\Theta} - \epsilon_\Theta$$

until  $\epsilon_x \leq tolerance$



## Jacobians for static forces

- Virtual work must be same independent of coordinates

$$\mathcal{F}^T \delta \chi = \tau^T \delta \Theta$$

- We remember that:

$$\delta \chi = J \delta \Theta$$

- Which gives us:

$$\mathcal{F}^T J = \tau^T$$

$$\tau = J^T \mathcal{F}$$



## Jacobians for static forces

$$\tau = J^T \mathcal{F}$$

- We can now see that for singular configurations, there will be directions where the required torque for a given force goes to zero, or inversely, **the forces generated by a given torque tend to infinity**. This may cause damage to the robot or the environment.



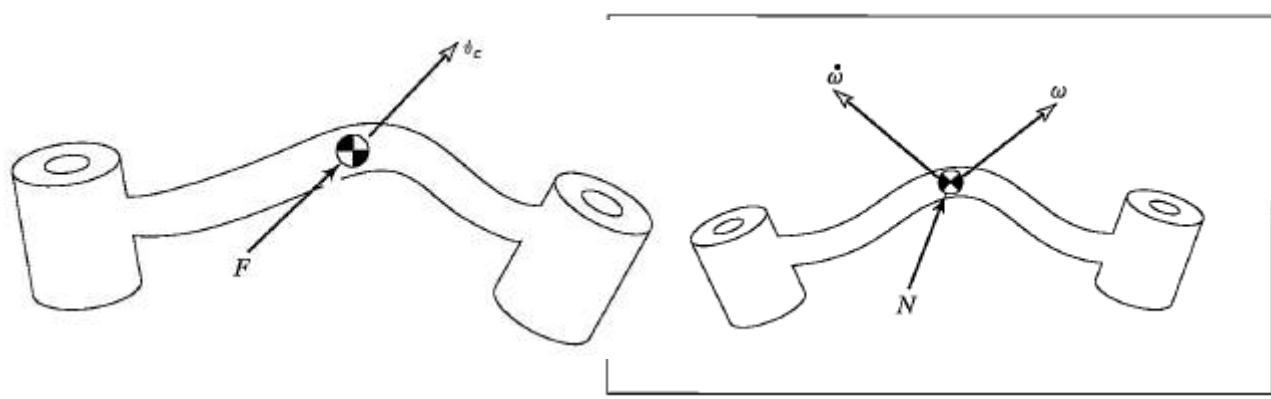
## Jacobians for static forces

$$\tau = J^T \mathcal{F}$$

- We can also calculate inverse kinematics by virtual forces and torques. We apply a "force" correcting the end effector position, calculate the torques this would generate, and move the robot accordingly. This gives us the update step:

$$\epsilon_{\Theta} = J^T(\hat{\Theta}) \epsilon_x$$

- This is useful when inverse of J does not exist, but typically converges slower.



$$F = m\dot{v}_C,$$

$$N = {}^C_I \dot{\omega} + \omega \times {}^C_I \omega,$$

$${}^A I = \begin{bmatrix} I_{xx} & -I_{xy} & -I_{xz} \\ -I_{yx} & I_{yy} & -I_{yz} \\ -I_{zx} & -I_{zy} & I_{zz} \end{bmatrix},$$

$$I_{xx} = \iiint_V (y^2 + z^2) \rho dv,$$

$$I_{yy} = \iiint_V (x^2 + z^2) \rho dv,$$

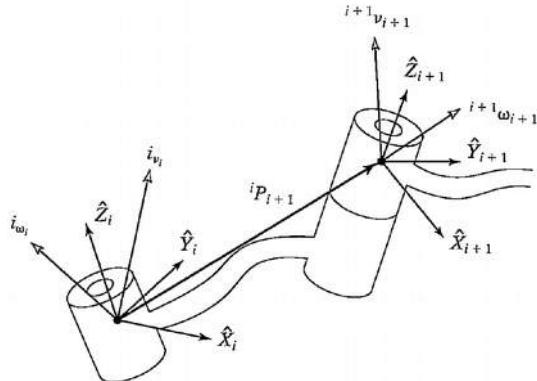
$$I_{zz} = \iiint_V (x^2 + y^2) \rho dv,$$

$$I_{xy} = \iiint_V xy \rho dv,$$

$$I_{xz} = \iiint_V xz \rho dv,$$

$$I_{yz} = \iiint_V yz \rho dv,$$

## Dynamics (R- MPC chapter 7) - Rotational joints



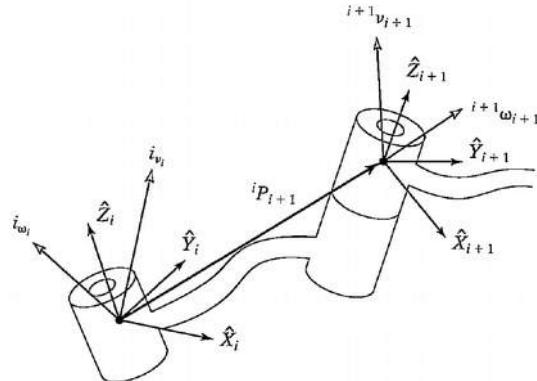
$${}^{i+1}\omega_{i+1} = {}^{i+1}_i R {}^i\omega_i + \dot{\theta}_{i+1} {}^{i+1}\hat{Z}_{i+1}$$

$${}^{i+1}\nu_{i+1} = {}^{i+1}_i R ({}^i\nu_i + {}^i\omega_i \times {}^iP_{i+1})$$

$${}^{i+1}\dot{\omega}_{i+1} = {}^{i+1}_i R {}^i\dot{\omega}_i + {}^{i+1}_i R {}^i\omega_i \times \dot{\theta}_{i+1} {}^{i+1}\hat{Z}_{i+1} + \ddot{\theta}_{i+1} {}^{i+1}\hat{Z}_{i+1}$$

$${}^{i+1}\dot{\nu}_{i+1} = {}^{i+1}_i R [{}^i\dot{\omega}_i \times {}^iP_{i+1} + {}^i\omega_i \times ({}^i\omega_i \times {}^iP_{i+1}) + {}^i\dot{\nu}_i]$$

# Dynamics (R-MPC chapter 7) - Prismatic joints



$${}^{i+1}\omega_{i+1} = {}^{i+1}_i R {}^i\omega_i,$$

$${}^{i+1}v_{i+1} = {}^{i+1}_i R ({}^i v_i + {}^i\omega_i \times {}^i P_{i+1}) + \dot{d}_{i+1} {}^{i+1}\hat{Z}_{i+1}$$

$${}^{i+1}\dot{\omega}_{i+1} = {}^{i+1}_i R {}^i\dot{\omega}_i$$

$$\begin{aligned} {}^{i+1}\ddot{v}_{i+1} = & {}^{i+1}_i R ({}^i\dot{\omega}_i \times {}^i P_{i+1} + {}^i\omega_i \times ({}^i\omega_i \times {}^i P_{i+1}) + {}^i\ddot{v}_i) \\ & + 2 {}^{i+1}\omega_{i+1} \times \dot{d}_{i+1} {}^{i+1}\hat{Z}_{i+1} + \ddot{d}_{i+1} {}^{i+1}\hat{Z}_{i+1} \end{aligned}$$

$${}^i\dot{v}_{C_i} = {}^i\dot{\omega}_i \times {}^i P_{C_i} + {}^i\omega_i \times ({}^i\omega_i + {}^i P_{C_i}) + {}^i\ddot{v}_i$$



# Dynamics

Newton - Euler approach:

- Find the acceleration and velocity of each joint, working outwards
- Find the necessary torque/force to generate that acceleration, adding the external forces and torques, working inwards



# Dynamics

Outward iterations:  $i : 0 \rightarrow 5$

$$\begin{aligned} {}^{i+1}\omega_{i+1} &= {}^i{}_i R {}^i\omega_i + \dot{\theta}_{i+1} {}^{i+1}\hat{Z}_{i+1}, \\ {}^{i+1}\dot{\omega}_{i+1} &= {}^i{}_i R {}^i\dot{\omega}_i + {}^i{}_i R {}^i\omega_i \times \dot{\theta}_{i+1} {}^{i+1}\hat{Z}_{i+1} + \ddot{\theta}_{i+1} {}^{i+1}\hat{Z}_{i+1}, \\ {}^{i+1}\dot{v}_{i+1} &= {}^i{}_i R ({}^i\dot{\omega}_i \times {}^i P_{i+1} + {}^i\omega_i \times ({}^i\omega_i \times {}^i P_{i+1}) + {}^i\dot{v}_i), \\ {}^{i+1}\dot{v}_{C_{i+1}} &= {}^{i+1}\dot{\omega}_{i+1} \times {}^{i+1}P_{C_{i+1}} \\ &\quad + {}^{i+1}\omega_{i+1} \times ({}^{i+1}\omega_{i+1} \times {}^{i+1}P_{C_{i+1}}) + {}^{i+1}\dot{v}_{i+1}, \\ {}^{i+1}F_{i+1} &= m_{i+1} {}^{i+1}\dot{v}_{C_{i+1}}, \\ {}^{i+1}N_{i+1} &= {}^{C_{i+1}}I_{i+1} {}^{i+1}\dot{\omega}_{i+1} + {}^{i+1}\omega_{i+1} \times {}^{C_{i+1}}I_{i+1} {}^{i+1}\omega_{i+1}. \end{aligned}$$

Inward iterations:  $i : 6 \rightarrow 1$

$$\begin{aligned} {}^i f_i &= {}^i{}_i R {}^{i+1}f_{i+1} + {}^i F_i, \\ {}^i n_i &= {}^i N_i + {}^i{}_i R {}^{i+1}n_{i+1} + {}^i P_{C_i} \times {}^i F_i \\ &\quad + {}^i P_{i+1} \times {}^i{}_i R {}^{i+1}f_{i+1}, \\ \tau_i &= {}^i n_i^T {}^i \hat{Z}_i. \end{aligned}$$



## Dynamics (R-MPC chapter 7)

The resulting dynamic equations can be written on the form (state-space equation):

$$\tau = M(\Theta)\ddot{\Theta} + V(\Theta, \dot{\Theta}) + G(\Theta) + J^T f$$



## Dynamics (DLR)

$$\tau = M(\Theta) \ddot{\Theta} + V(\Theta, \dot{\Theta}) + G(\Theta) + J^T f$$



## Dynamics (DLR)





## Dynamics (DLR)





## State of the art - industrial manipulation





# Dynamics (DLR)



End-Effector Airbags for Accelerating  
Human-Robot Collaboration