



THE UNIVERSITY of EDINBURGH
informatics

Advanced Robotics

Dynamics II

Dynamics of Open Chains

(Ref: Ch. 8 of K.M. Lynch & F.C. Park, *Modern Robotics*)

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Recap

- ❑ We will discuss the following three topics:
 - ❑ 1D point mass
 - ❑ A 'general' dynamic robot (=> Dynamics II)
 - ❑ Joint space control
- ❑ For now we assume that the robot is fully actuated and that $\mathbf{v}_q = \dot{\mathbf{q}}$
(ie velocity and configuration space have the same dimension)
- ❑ We also assume motors are equipped with accurate **position** sensors (ie we know $\mathbf{q}$ accurately)

Forward and inverse dynamics

□ As for geometry and kinematics, we are interested in two formulations of the dynamics of a system

□ Forward dynamics: Given $\mathbf{q}$, $\mathbf{v}_q$ and $\boldsymbol{\tau}$, compute joint accelerations $\dot{\mathbf{v}}_q$

Useful for *simulation*

□ Inverse dynamics: Given $\mathbf{q}$, $\mathbf{v}_q$ and $\dot{\mathbf{v}}_q$, compute torque commands $\boldsymbol{\tau}$

Useful for *control*

Two approaches to dynamics

❑ Lagrangian dynamics

- ❑ Intuitive variational formulation (principles you may already be familiar with)
- ❑ Equations get messy quite quickly

❑ Newton-Euler dynamics

- ❑ Not so intuitive, we'll focus mainly on the ideas (detailed derivations in all standard texts)
- ❑ Newton-Euler is practically used in robotics because efficient *recursive* algorithms can be derived from the formulation

Lagrangian dynamics

❑ In classic mechanics, as you would have seen in secondary school, we have the core concept of mechanical energy

❑ Example: free fall of a point mass

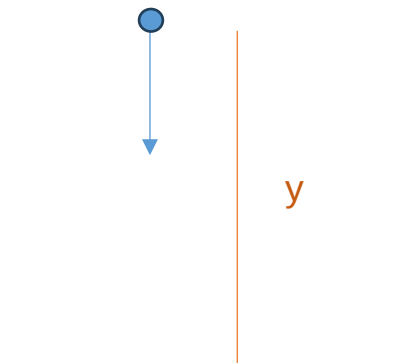
❑ Kinetic Energy T

$$T = \frac{1}{2}m\dot{y}^2$$

❑ Potential Energy U (gravitational)

$$U = mgy$$

❑ If only *conservative* forces are applied, total energy $E = T + U$ is constant



Lagrangian dynamics –intuitions

□ Let's write

$$L(y, \dot{y}) = T - U = \frac{1}{2}m\dot{y}^2 - mgy$$

□ We can verify that

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{y}} \right) = \frac{\partial L}{\partial y}$$

$$\frac{\partial L}{\partial \dot{y}} = m\dot{y}$$

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{y}} \right) = m\ddot{y}$$

$$\frac{\partial L}{\partial y} = -mg$$

□ Which brings us back to the very familiar Newton's law $F = ma$.

□ L is called the Lagrangian, and concisely captures the physics

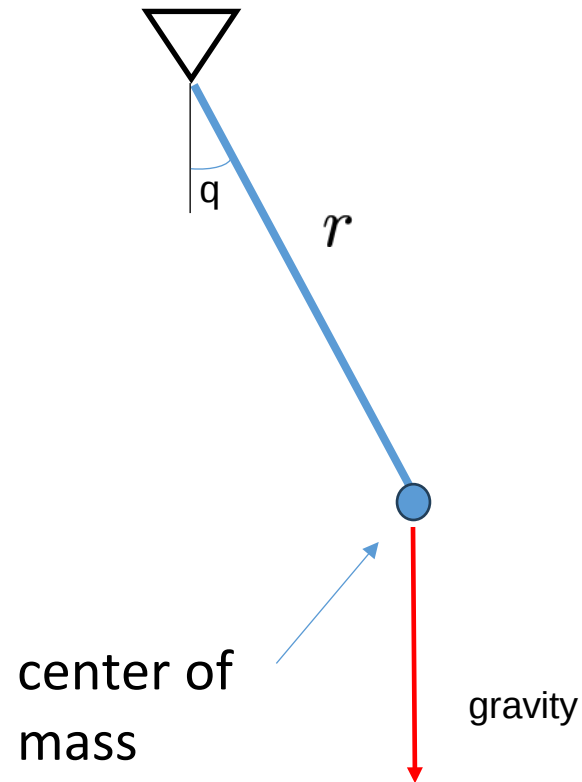
Lagrangian mechanics

- In the general case, it can be proven that this equation equates to the torques

$$\frac{d}{dt} \left(\frac{\partial L(\mathbf{q}, \dot{\mathbf{q}})}{\partial \dot{y}} \right) - \frac{\partial L(\mathbf{q}, \dot{\mathbf{q}})}{\partial y} = \tau$$

- We won't prove this (all variational calculus and mechanics books will cover this - if you were interested), just show it on another example

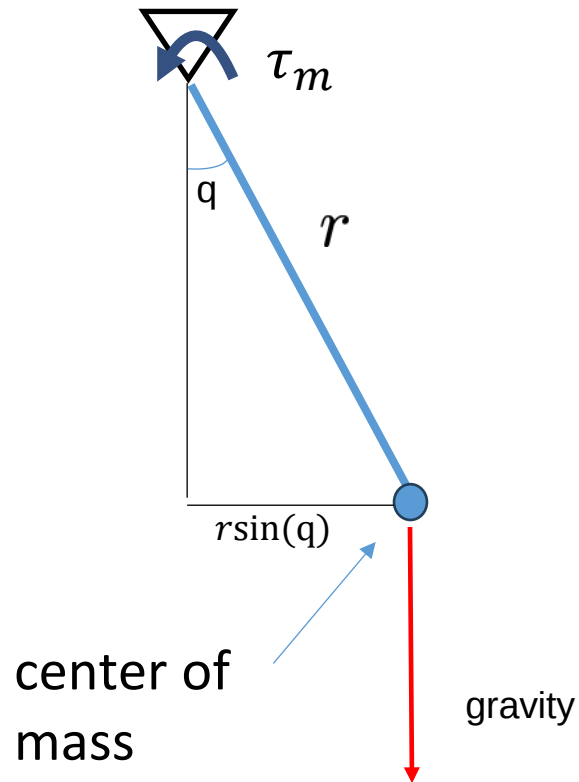
Example: Equation of motion of a 1-DOF robot arm



mass of pendulum: m

moment of inertia
(about pivot point): I_p

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Newton's second law :

$$I_p \ddot{\mathbf{q}} = \sum_i \tau_i$$

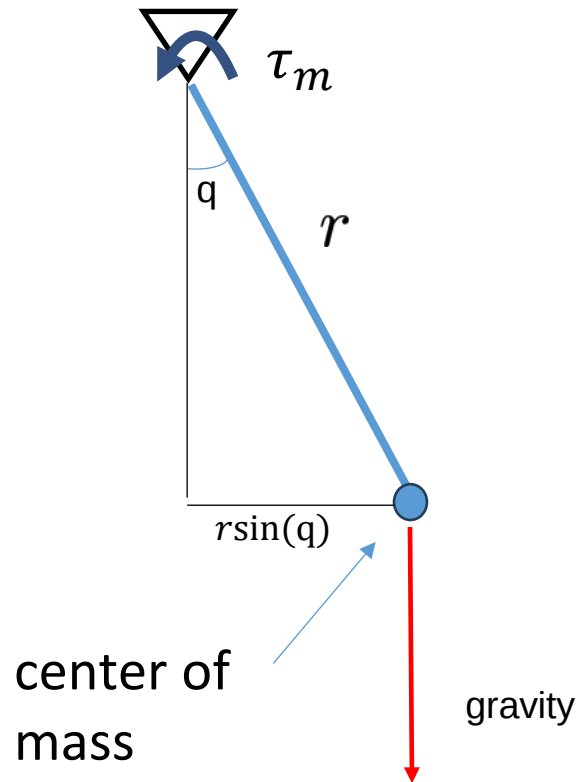
$$\tau_g = -rmg \sin(q)$$

Gravity torque

$$\tau_m$$

Motor torque

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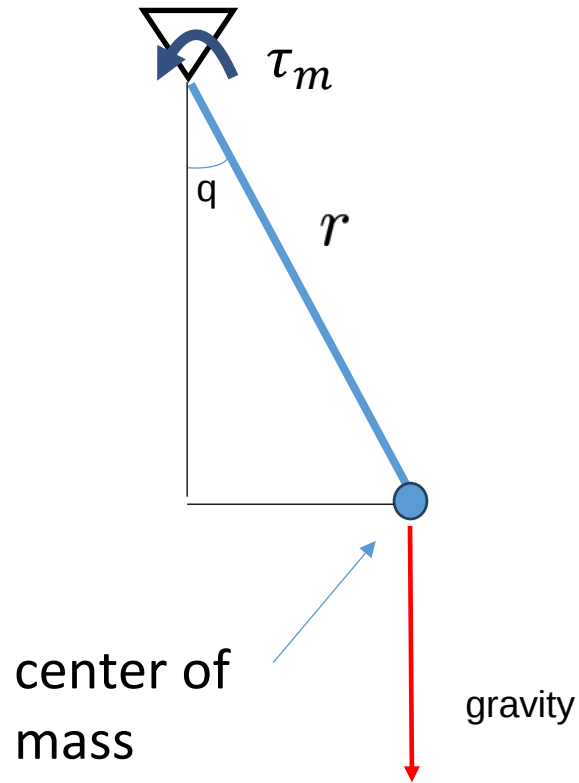
$$\tau_g = -rmg \sin(q)$$

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Motor torque

Lagrangian computation



$$U(q) = mgh, \quad \text{where} \quad h = r(1 - \cos(q))$$

$$T(q) = \frac{1}{2} I_p \dot{q}^2$$

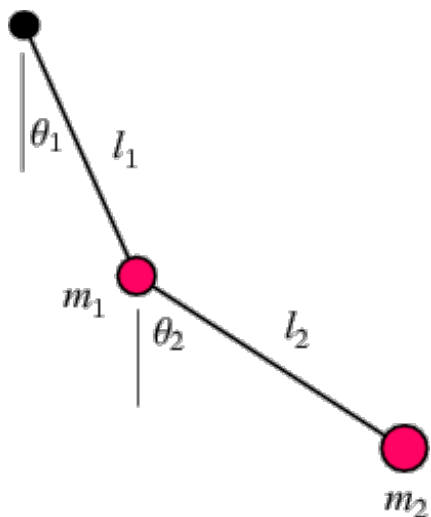
$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}} \right) = I_p \ddot{q} \quad \frac{\partial L}{\partial q} = \frac{\partial}{\partial q} (mgr(1 - \cos(q))) = -mgr \sin(q)$$

$$\tau_m = \frac{d}{dt} \left(\frac{\partial L(\mathbf{q}, \dot{\mathbf{q}})}{\partial \dot{q}} \right) - \frac{\partial L(\mathbf{q}, \dot{\mathbf{q}})}{\partial q}$$

$$\tau_m = I_p \ddot{\mathbf{q}} + rmgsin(q)$$

Why use the Lagrangian?

- ❑ Newton's law equations consider each rigid body *individually*. The problem is that we will have to also consider constraint forces (e.g. double pendulum equations below – source: wolfram)
- ❑ Lagrangian formulations nicely *incorporates* all of these considerations
- ❑ Things get rapidly complicated when we write out complete equations of motion. The differential equations for the double pendulum system are respectively:



$$(m_1 + m_2) L_1^2 \ddot{\theta}_1 + m_2 L_1 L_2 \cos(\theta_1 - \theta_2) \ddot{\theta}_2 + m_2 L_1 L_2 \sin(\theta_1 - \theta_2) \dot{\theta}_2^2 + g(m_1 + m_2) L_1 \sin(\theta_1) = 0$$

$$m_2 L_1 L_2 \cos(\theta_1 - \theta_2) \ddot{\theta}_1 + m_2 L_2^2 \ddot{\theta}_2 - m_2 L_1 L_2 \sin(\theta_1 - \theta_2) \dot{\theta}_1^2 + g m_2 L_2 \sin(\theta_2) = 0$$

Indeed the behaviour is ... chaotic



Canonical form for articulated rigid bodies

□ We can regroup terms as follows

$$\underset{\substack{\text{Inertia Matrix}}}{\mathbf{M}(\mathbf{q})} \ddot{\mathbf{q}} + \underset{\substack{\text{Coriolis Matrix}}}{\mathbf{B}(\mathbf{q})} [\dot{\mathbf{q}}\dot{\mathbf{q}}] + \underset{\substack{\text{Centrifugal Matrix}}}{\mathbf{C}(\mathbf{q})} [\dot{\mathbf{q}}^2] + \underset{\substack{\text{Gravity Vector}}}{\mathbf{G}(\mathbf{q})} = \underset{\substack{\text{External Forces}}}{\boldsymbol{\tau}}$$

$$[\dot{\mathbf{q}}\dot{\mathbf{q}}] = \begin{bmatrix} \dot{q}_1\dot{q}_2 & \dot{q}_1\dot{q}_3 & \dots & \dot{q}_{n-1}\dot{q}_n \end{bmatrix}^T$$

$$[\dot{\mathbf{q}}^2] = \begin{bmatrix} \dot{q}_1^2 & \dot{q}_2^2 & \dots & \dot{q}_n^2 \end{bmatrix}^T$$

M,B,C,G are only configuration dependent

What are Coriolis forces?



Other representations in the literature

$$\mathbf{M}(\mathbf{q}) \ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau}$$

$\mathbf{C}$ is a vector with Coriolis plus centrifugal terms

$$\mathbf{M}(\mathbf{q}) \ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) \dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau}$$

$\mathbf{C}$ is a matrix with Coriolis plus centrifugal terms

$$\mathbf{M}(\mathbf{q}) \ddot{\mathbf{q}} + \mathbf{h}(\mathbf{q}, \dot{\mathbf{q}}) = \boldsymbol{\tau}$$

More about the inertia matrix $M(q)$:

- ❑ M defines the Kinetic energy of our robot:

$$T(\mathbf{q}, \dot{\mathbf{q}}) = \frac{1}{2} \dot{\mathbf{q}}^T \mathbf{M}(\mathbf{q}) \dot{\mathbf{q}}$$

- ❑ Because $T \geq 0, \forall(\mathbf{q}, \dot{\mathbf{q}})$, $\mathbf{M}$ is ?

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$$T(\mathbf{q}, \dot{\mathbf{q}}) = \frac{1}{2} \dot{\mathbf{q}}^T \mathbf{M}(\mathbf{q}) \dot{\mathbf{q}}$$

- ❑ Because $T \geq 0, \forall(\mathbf{q}, \dot{\mathbf{q}})$, $\mathbf{M}$ is positive-definite

- ❑ As such it is invertible:

$$\mathbf{M}(\mathbf{q}) \ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau} \quad \Longrightarrow \quad \ddot{\mathbf{q}} = \mathbf{M}(\mathbf{q})^{-1} (-\mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) - \mathbf{G}(\mathbf{q}) + \boldsymbol{\tau})$$

This gives us the equations for forward and inverse dynamics

$$\mathbf{M}(\mathbf{q}) \ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau}$$

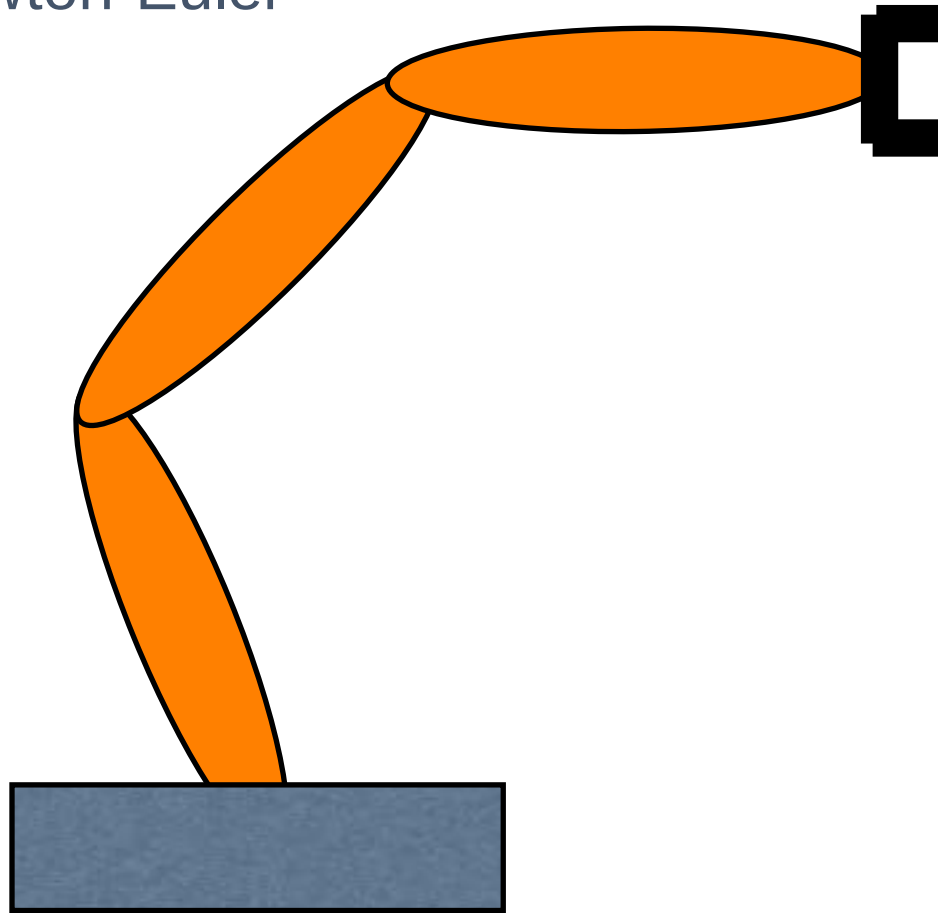
Inverse Dynamics equation = control

$$\ddot{\mathbf{q}} = \mathbf{M}(\mathbf{q})^{-1} (-\mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) - \mathbf{G}(\mathbf{q}) + \boldsymbol{\tau})$$

Forward Dynamics equation = simulation

A recursive algorithm for articulated robots: intuitions

- Based on Newton-Euler

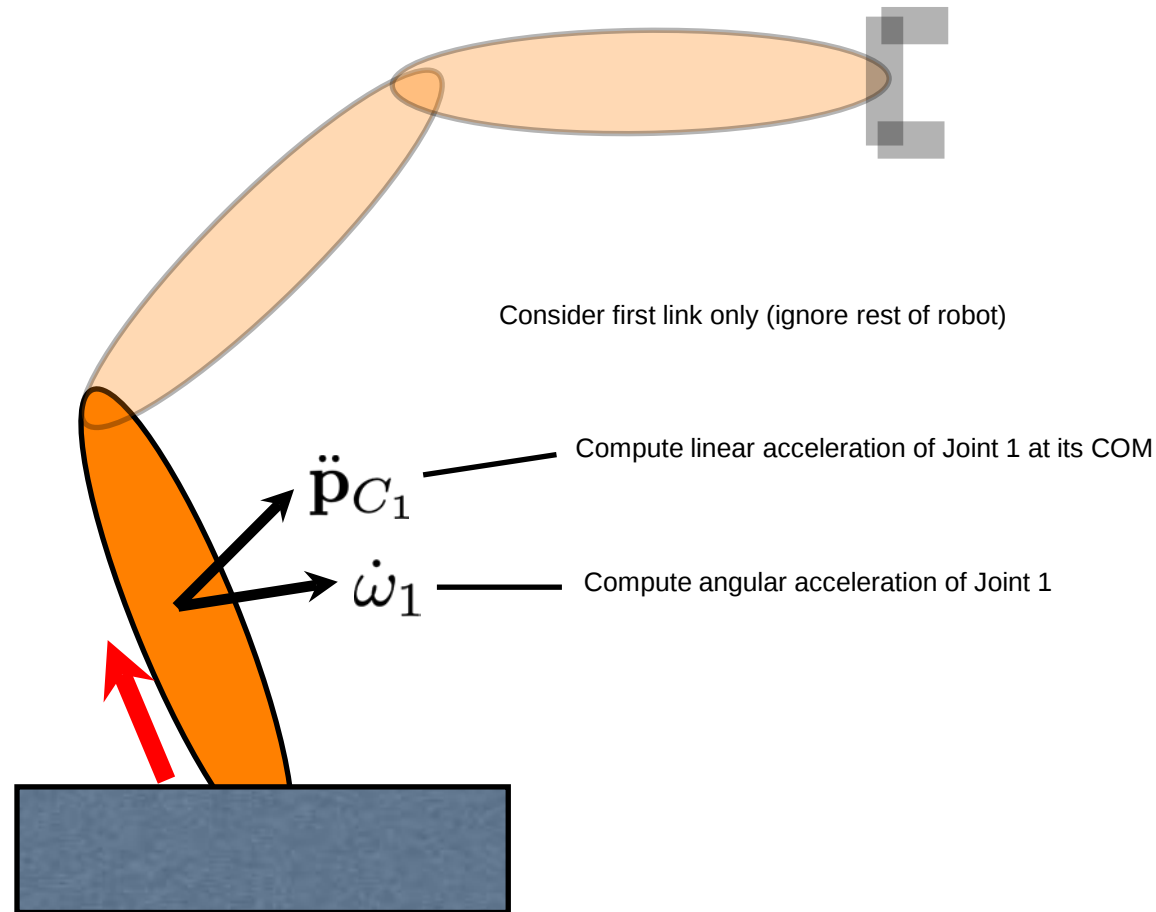


First given $\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}$
compute all link
accelerations, starting
from the base

A recursive algorithm for articulated robots: intuitions

❑ Based on Newton-Euler

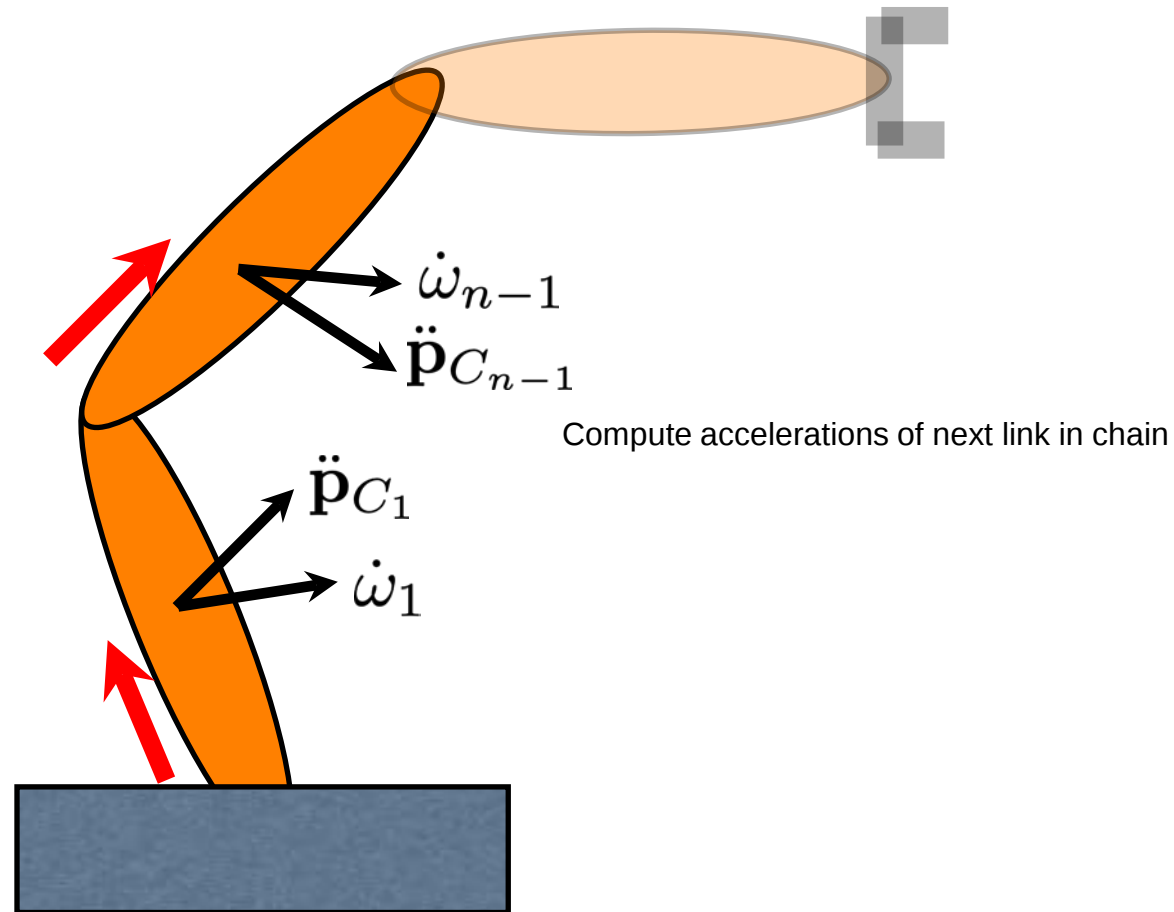
$\mathbf{p}_{C_1}$ denotes the coordinates of the first link C_1 , more on this next week



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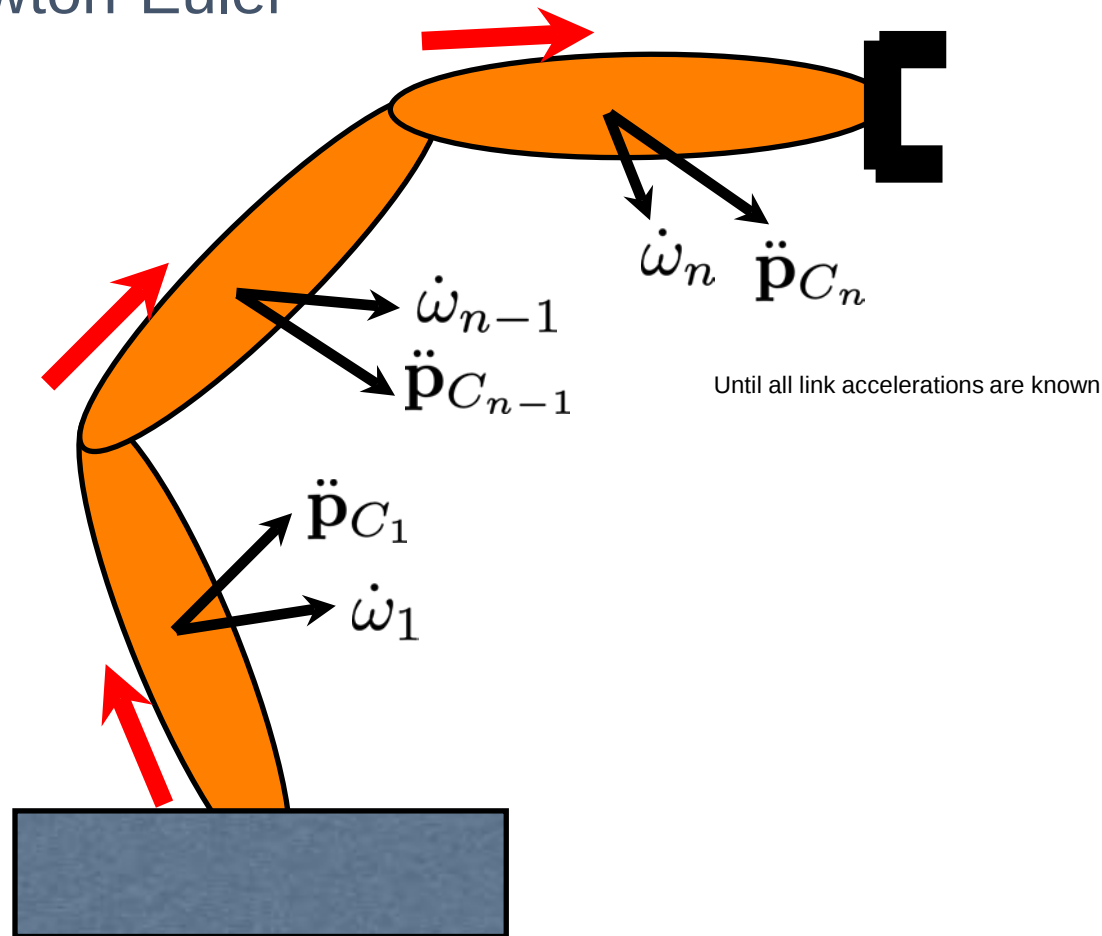
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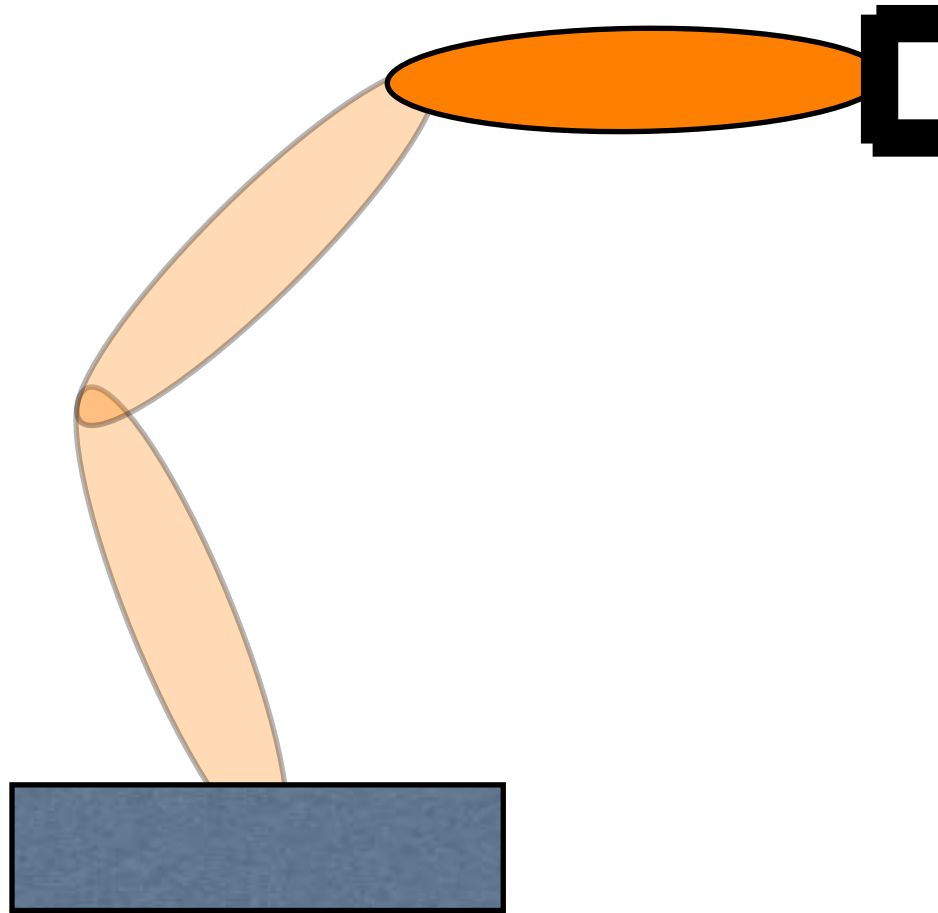
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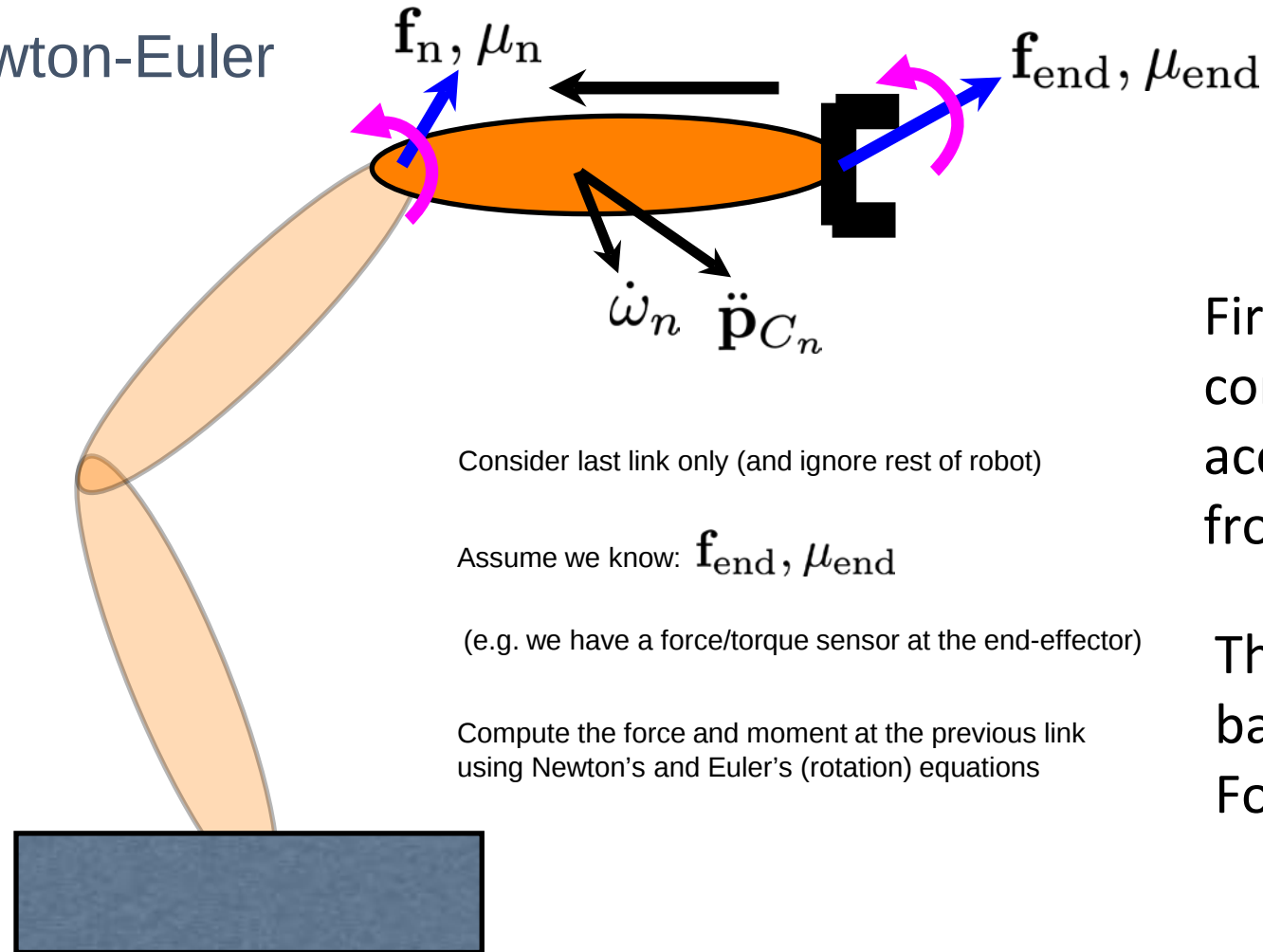


First given $\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}$
compute all link
accelerations, starting
from the base

Then, proceed
backwards to compute
Forces and moments

A recursive algorithm for articulated robots: intuitions

□ Based on Newton-Euler



Consider last link only (and ignore rest of robot)

Assume we know: $\mathbf{f}_{\text{end}}, \mu_{\text{end}}$

(e.g. we have a force/torque sensor at the end-effector)

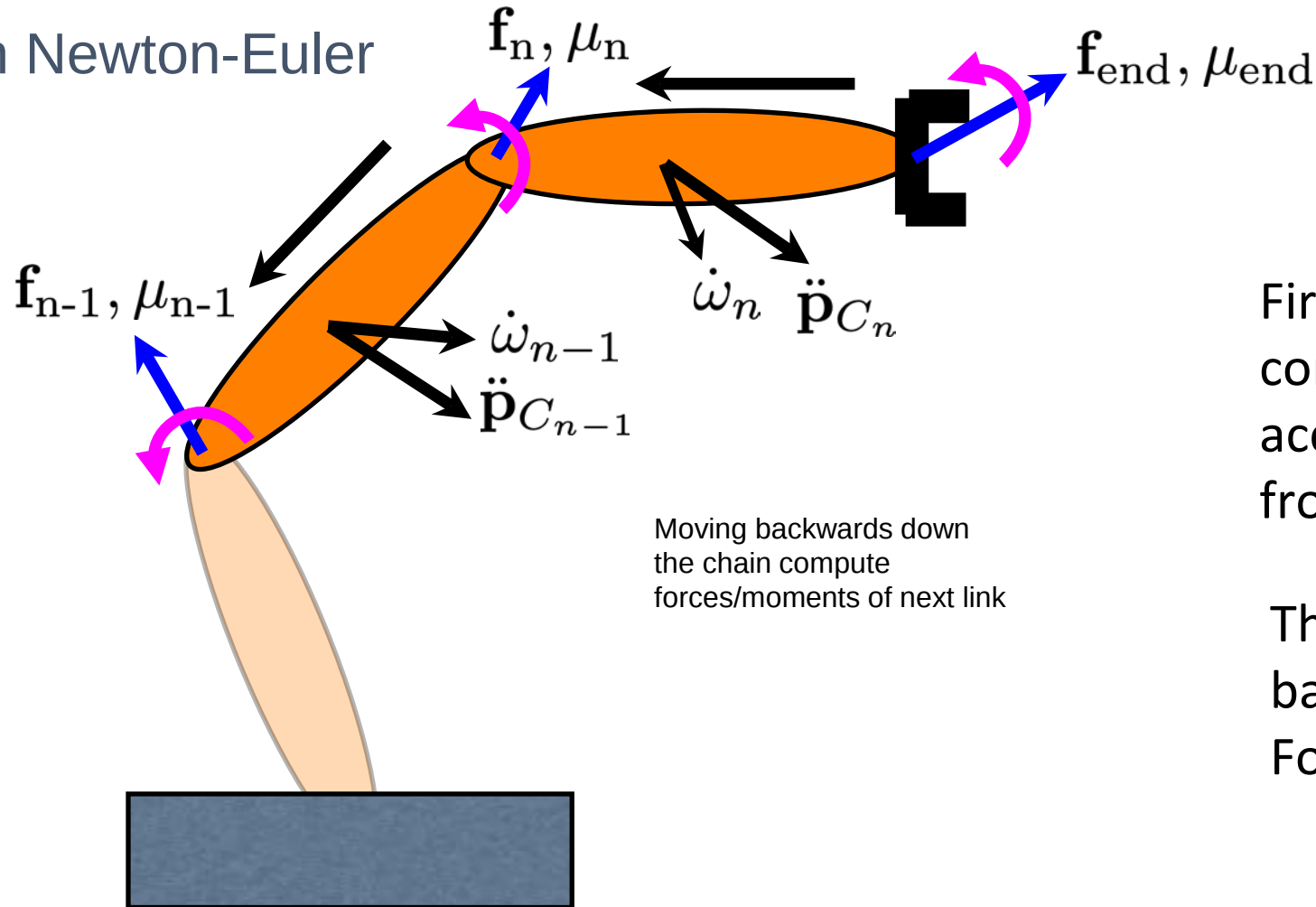
Compute the force and moment at the previous link using Newton's and Euler's (rotation) equations

First given $\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}$
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A recursive algorithm for articulated robots: intuitions

- Based on Newton-Euler

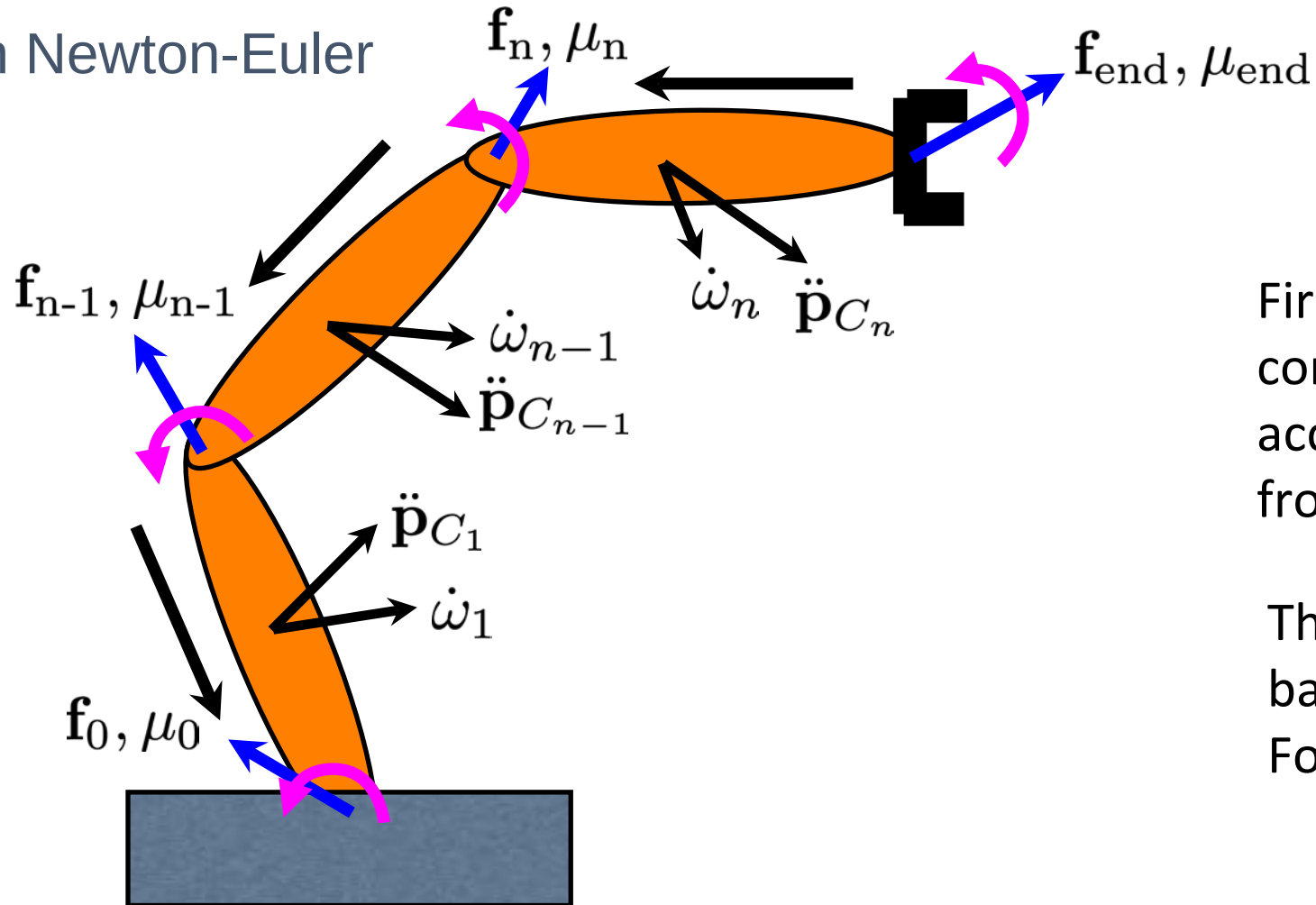


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A recursive algorithm for articulated robots

- ❑ The Recursive Newton-Euler Algorithm (RNEA) gives us an iterative way to compute Inverse Dynamics:

$$\mathbf{M}(\mathbf{q}) \ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau}$$

$$RNEA(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) = \boldsymbol{\tau}$$

- ❑ Without requiring to work out the mass matrix. However we still need $\mathbf{M}$ for forward dynamics

$$\ddot{\mathbf{q}} = \mathbf{M}(\mathbf{q})^{-1} (-\mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) - \mathbf{G}(\mathbf{q}) + \boldsymbol{\tau})$$

Can RNEA be used to compute the mass matrix?

- ❑ A trick is to compute M column by column by setting:

$$\mathbf{g} = 0 \quad \dot{\mathbf{q}} = 0 \quad \ddot{q}_i = 1 \quad \ddot{q}_j = 0, \forall j \neq i$$

- ❑ In such a case RNEA return the i-th column of M
- ❑ As a result, computing inverse dynamics is faster than forward dynamics:
 - $O(n^2)$ for computing *direct dynamics*,
 - $O(n)$ for computing *inverse dynamics*.

How do we control this ? (transition slide)

General Robot System Dynamics

- **State** $x = (q, \dot{q}) \in \mathbb{R}^{2n}$
 - joint positions $q \in \mathbb{R}^n$
 - joint velocities $\dot{q} \in \mathbb{R}^n$
- **Controls** $u \in \mathbb{R}^n$ are the *torques* generated in each motor.
- The **system dynamics** are:

$$M(q) \ddot{q} + C(q, \dot{q}) \dot{q} + G(q) = u$$

$M(q) \in \mathbb{R}^{n \times n}$ is positive definite inertia matrix
(can be inverted $\rightarrow$ forward simulation of dynamics)

$C(q, \dot{q}) \in \mathbb{R}^n$ are the centripetal and coriolis forces

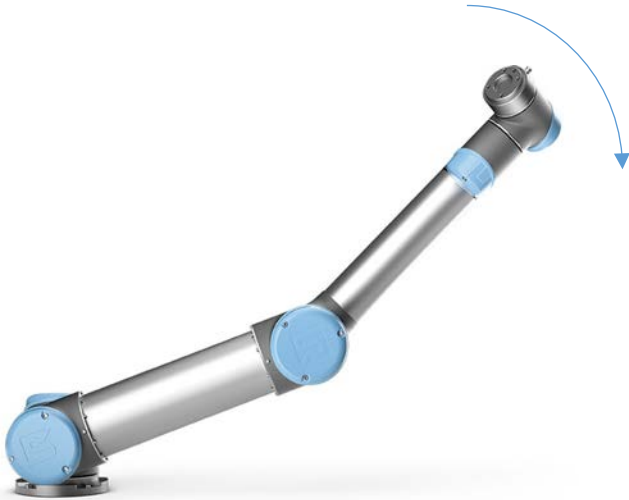
$G(q) \in \mathbb{R}^n$ are the gravitational forces

u are the joint torques

Computing $M(q)$ and $F(q, \dot{q})$

□ More compact form as:

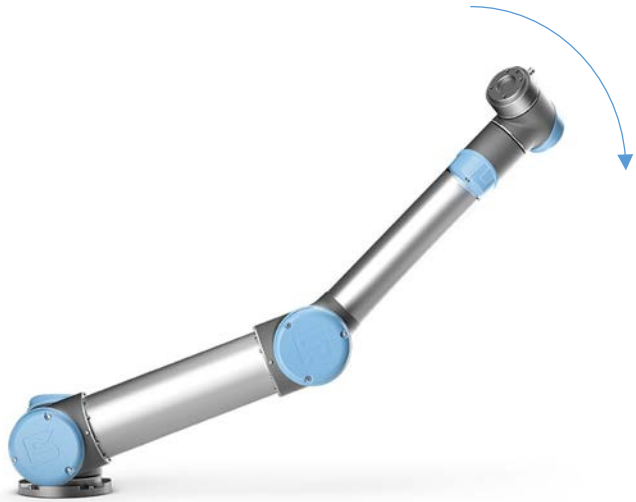
$$M(q) \ddot{q} + F(q, \dot{q}) = u$$



There exist efficient algorithms to compute M and F .
This is implemented in Pinocchio

Implications for (multi-body) dynamics

- If we know the desired $\ddot{q}^*$ for each joint, the eqn.
 $M(q) \ddot{q}^* + F(q, \dot{q}) = u^*$ gives the desired torques.



Joint Space control

- Where could we get the desired $\ddot{q}^*$ from?

Assume we have a nice smooth **reference trajectory** $q_{0:T}^{\text{ref}}$ (generated with some motion profile or alike), we can at each t read off the desired acceleration as

Open loop

$$\ddot{q}_t^{\text{ref}} := \frac{1}{\tau} [(q_{t+1} - q_t)/\tau - (q_t - q_{t-1})/\tau] = (q_{t-1} + q_{t+1} - 2q_t)/\tau^2$$

What if we directly use desired reference acceleration?

Tiny errors in acceleration will **accumulate** greatly over time and this makes this an **unstable** approach!

Joint Space control

Choose a desired acceleration $\ddot{q}_t^*$ that implies a *PD-like behavior around the reference trajectory!*

$$\ddot{q}_t^* = \ddot{q}_t^{\text{ref}} + K_p(q_t^{\text{ref}} - q_t) + K_d(\dot{q}_t^{\text{ref}} - \dot{q}_t)$$



$$M(q) \ddot{q}^* + F(q, \dot{q}) = u^*$$

This is a standard and convenient way of tracking a reference trajectory when the **robot dynamics are known**: *all the joints will behave exactly like a 1D point mass around the reference trajectory!*

Summary

- We covered the following three topics:

- 1D point mass
- A 'general' dynamic robot (→ Dynamics II)

- Joint space control method

$$\ddot{q}_t^* = \ddot{q}_t^{\text{ref}} + K_p(q_t^{\text{ref}} - q_t) + K_d(\dot{q}_t^{\text{ref}} - \dot{q}_t)$$



$$M(q) \ddot{q}^* + F(q, \dot{q}) = u^*$$