



Robotics 1

Inverse differential kinematics Statics and force transformations

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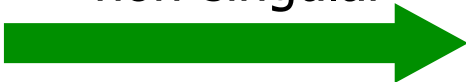
Inversion of differential kinematics

- find the joint velocity vector that realizes a **desired** end-effector “generalized” velocity (linear and angular)

generalized velocity

$$\mathbf{v} = \mathbf{J}(\mathbf{q}) \dot{\mathbf{q}}$$

J square and non-singular


$$\dot{\mathbf{q}} = \mathbf{J}^{-1}(\mathbf{q}) \mathbf{v}$$

- problems
 - **near** a singularity of the Jacobian matrix (high $\dot{\mathbf{q}}$)
 - for **redundant** robots (no standard “inverse” of a rectangular matrix)

in these cases, “more robust” inversion methods are needed



Incremental solution to inverse kinematics problems

- joint velocity inversion can be used also to solve **on-line** and **incrementally** a “sequence” of inverse kinematics problems
- each problem differs by a **small** amount **dr** from previous one

$$r = f_r(q)$$

direct kinematics

$$dr = \frac{\partial f_r(q)}{\partial q} dq = J_r(q) dq$$

differential kinematics

$$r \rightarrow r + dr$$

$$r + dr = f_r(q)$$

first, increment the
desired task variables



$$q = f_r^{-1}(r + dr)$$

then, solve the inverse
kinematics problem

$$dq = J_r^{-1}(q) dr$$

first, solve the inverse
differential kinematics problem



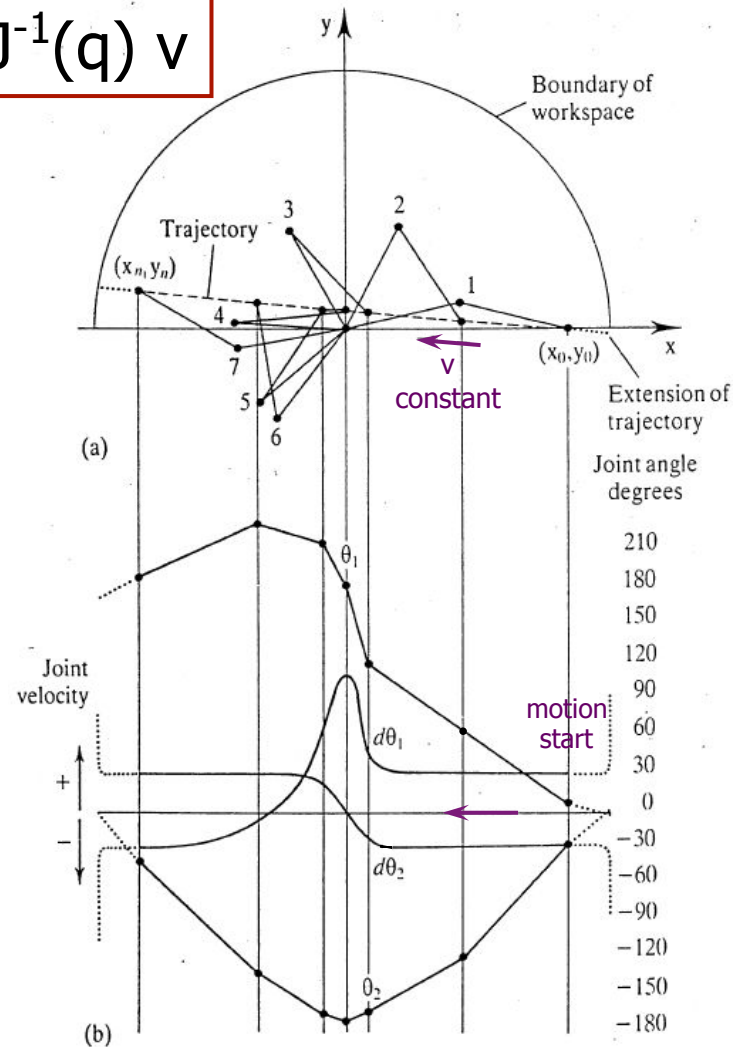
$$q \rightarrow q + dq$$

then, increment the
original joint variables



Behavior near a singularity

$$\dot{q} = J^{-1}(q) v$$



- problems arise only when commanding joint motion by **inversion** of a given Cartesian motion task
- here, a linear Cartesian trajectory for a planar 2R robot
- there is a sudden increase of the displacement/velocity of the **first joint** near $\theta_2 = -\pi$ (end-effector close to the origin), despite the required Cartesian displacement is small

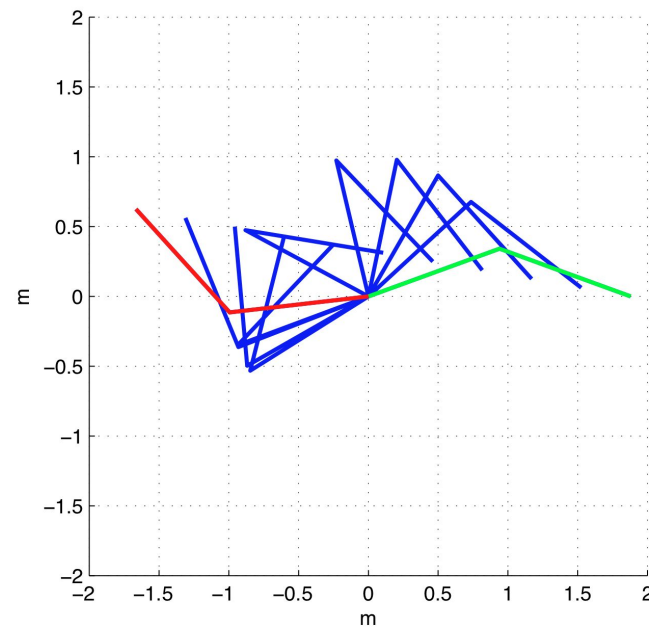
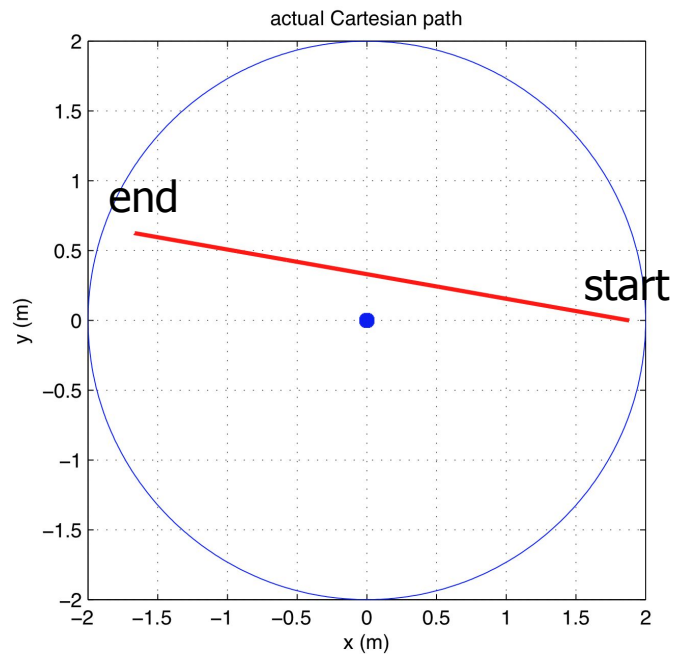


Simulation results

planar 2R robot in straight line Cartesian motion

$$\dot{q} = J^{-1}(q) v$$

regular case



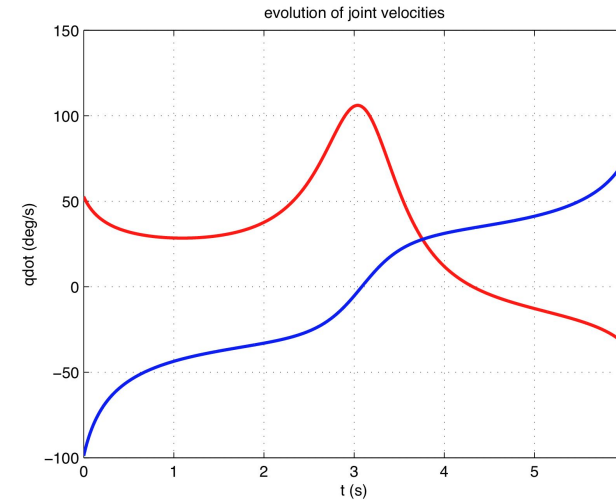
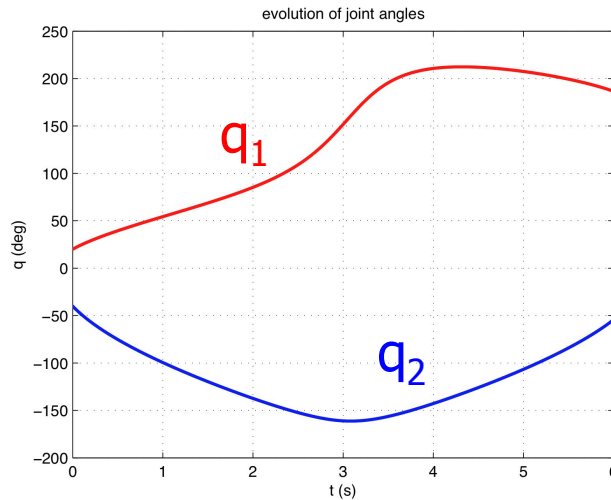
a line from right to left, at $\alpha=170^\circ$ angle with x-axis,
executed at constant speed $v=0.6$ m/s for $T=6$ s



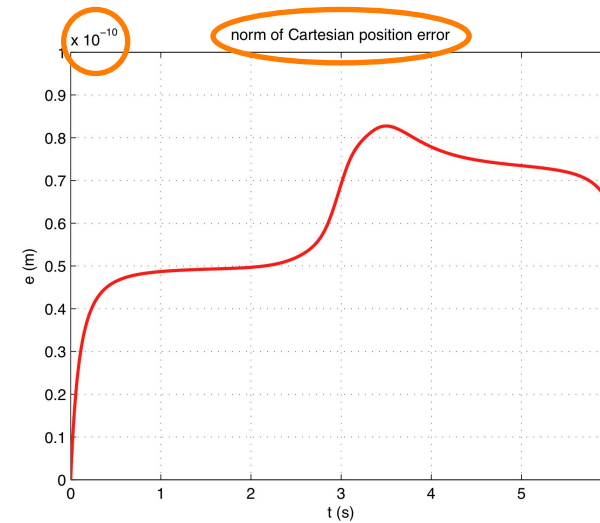
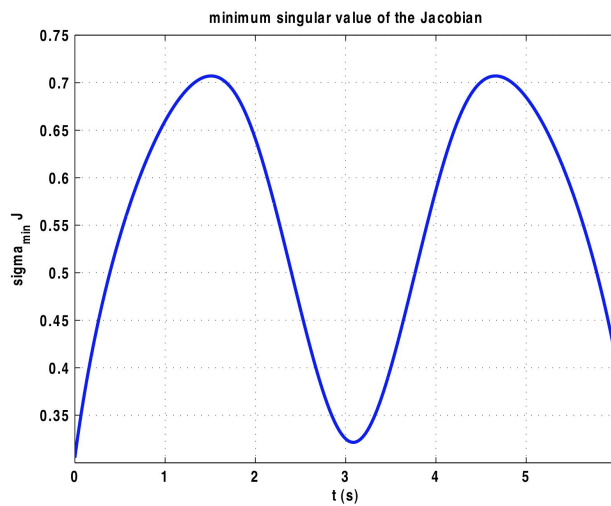
Simulation results

planar 2R robot in straight line Cartesian motion

path at $\alpha=170^\circ$



regular case



error due only to numerical integration (10⁻¹⁰)

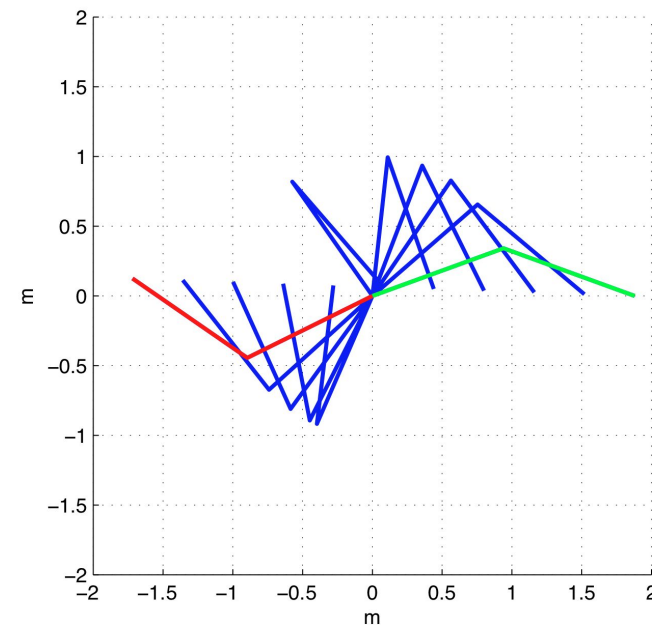
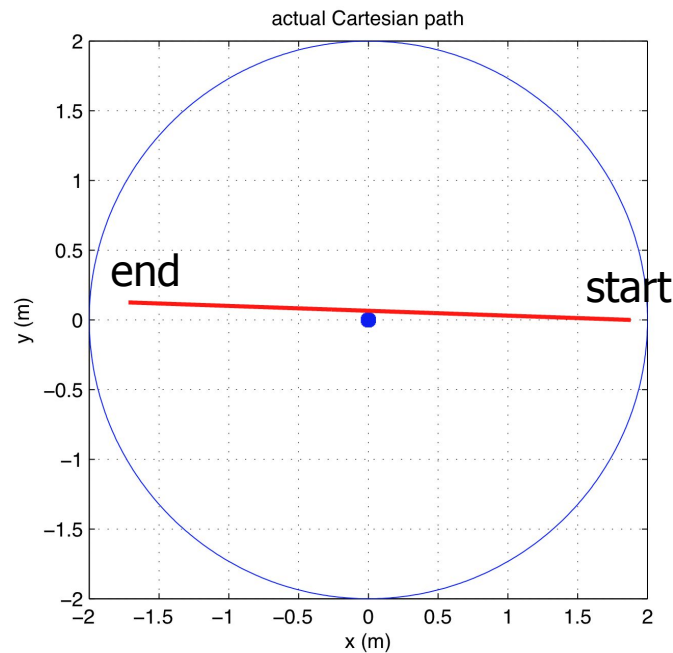


Simulation results

planar 2R robot in straight line Cartesian motion

$$\dot{q} = J^{-1}(q) v$$

close to **singular** case



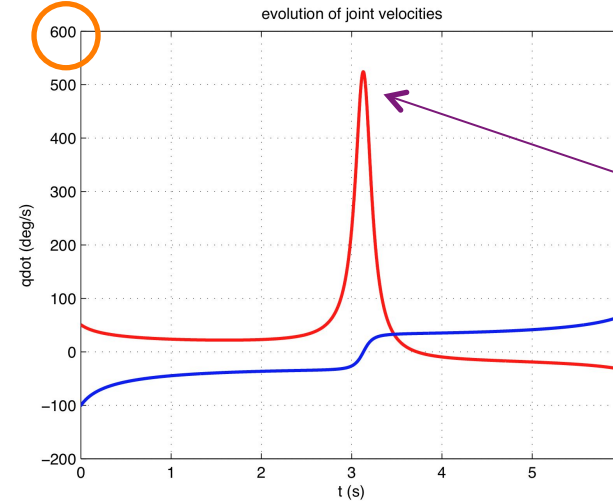
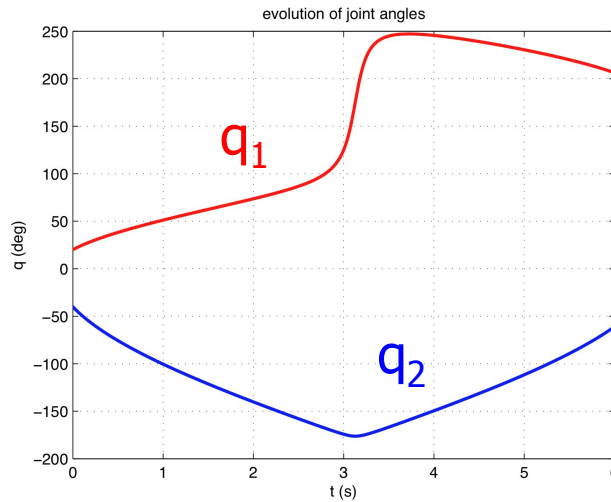
a line from right to left, at $\alpha=178^\circ$ angle with x-axis,
executed at constant speed $v=0.6$ m/s for $T=6$ s



Simulation results

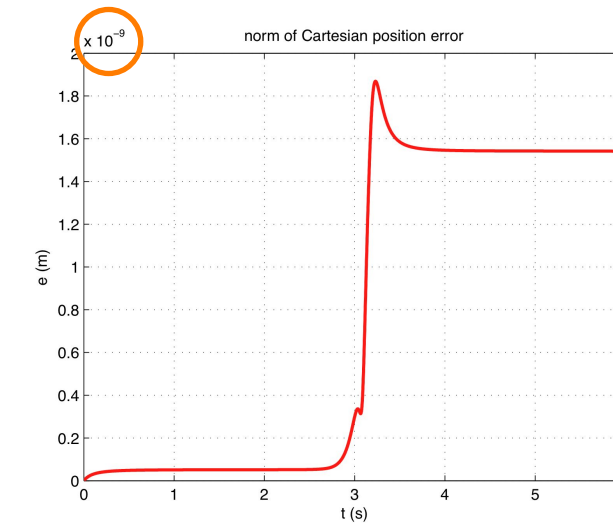
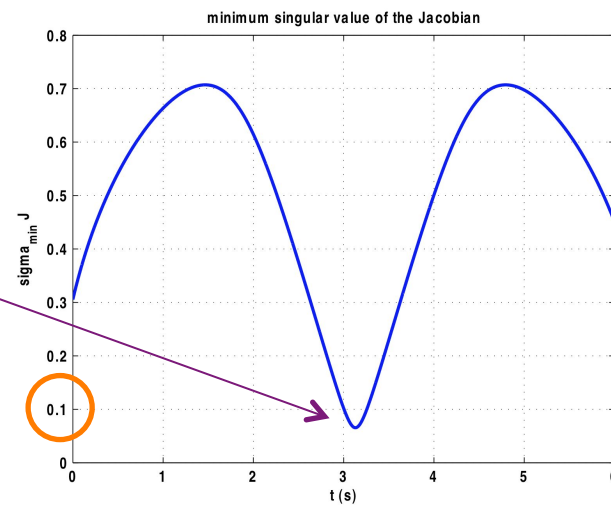
planar 2R robot in straight line Cartesian motion

path at $\alpha=178^\circ$



large peak of \dot{q}_1

close to singular case



still very small, but increased numerical integration error ($2 \cdot 10^{-9}$)

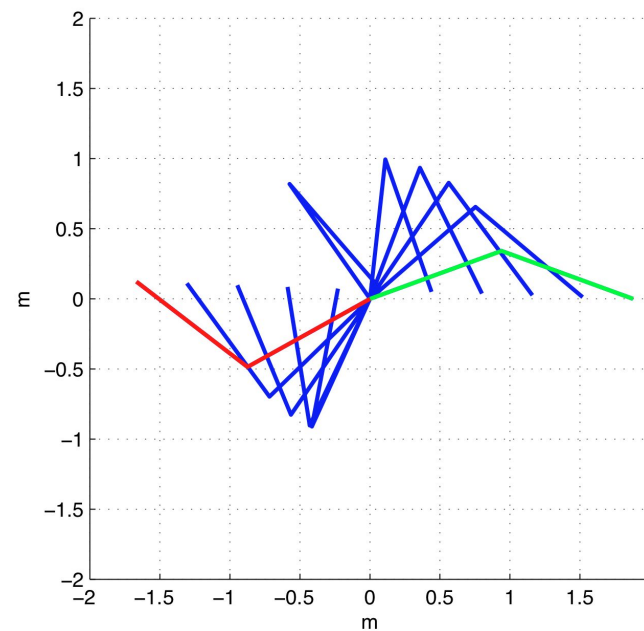
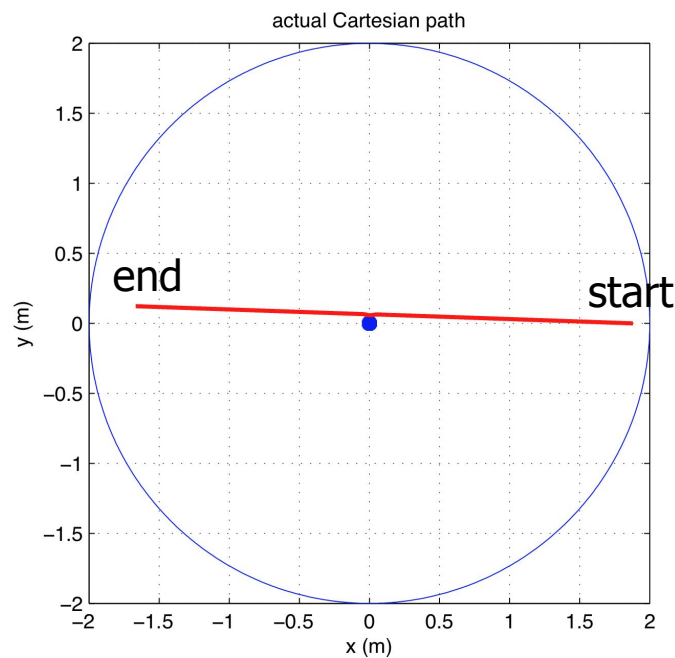


Simulation results

planar 2R robot in straight line Cartesian motion

$$\dot{q} = J^{-1}(q) v$$

close to **singular** case
with joint velocity **saturation** at $V_i=300^\circ/s$



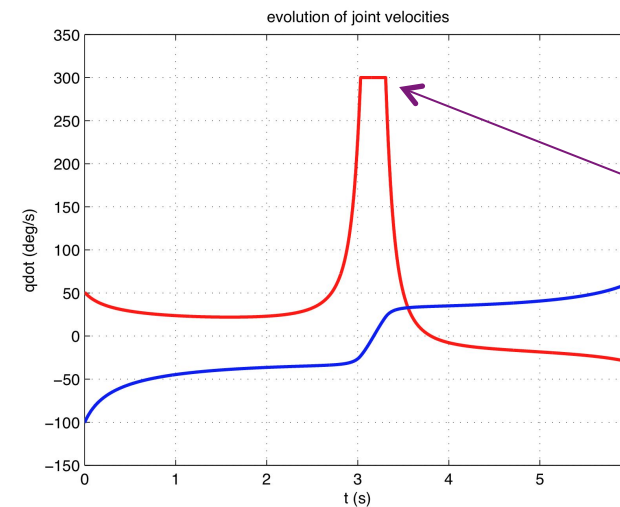
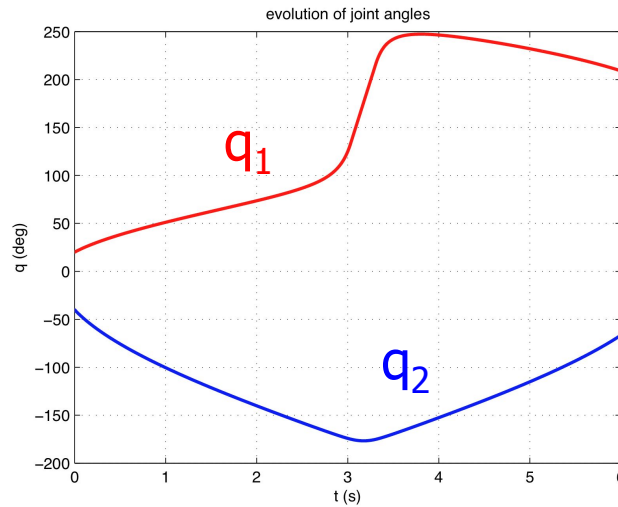
a line from right to left, at $\alpha=178^\circ$ angle with x-axis,
executed at constant speed $v=0.6$ m/s for $T=6$ s



Simulation results

planar 2R robot in straight line Cartesian motion

path at $\alpha=178^\circ$

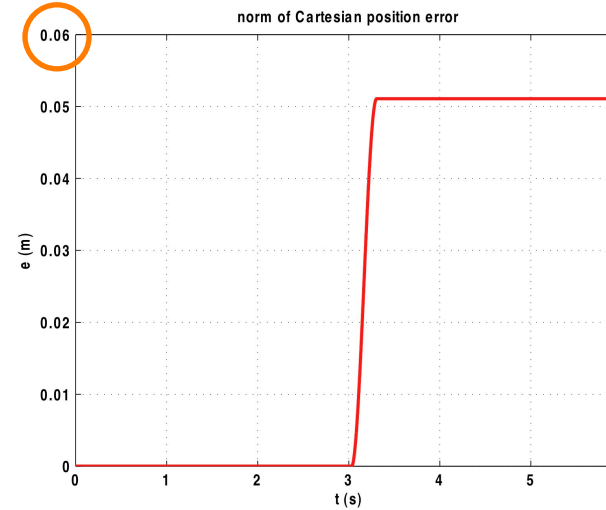
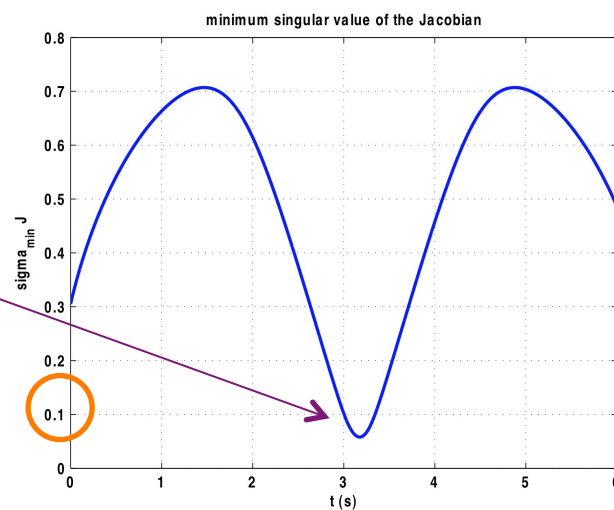


saturated value of \dot{q}_1



actual position error!! (6 cm)

close to singular case





Damped Least Squares method

$$\min_{\dot{q}} H = \frac{\lambda}{2} \|\dot{q}\|^2 + \frac{1}{2} \|J\dot{q} - v\|^2, \quad \lambda \geq 0$$

$$\dot{q} = (\lambda I_n + J^T J)^{-1} J^T v = J^T (\lambda I_m + J J^T)^{-1} v$$

equivalent expressions, but this one is more convenient in redundant robots!

- inversion of differential kinematics as an **optimization problem**
- function H = **weighted** sum of two objectives (minimum error norm on achieved end-effector velocity and minimum norm of joint velocity)
- $\lambda = 0$ when "far enough" from a singularity
- with $\lambda > 0$, there is a (vector) **error** ε ($= v - J\dot{q}$) in executing the desired end-effector velocity v (check that $\varepsilon = \lambda (\lambda I_m + J J^T)^{-1} v$!), but the joint velocities are always **reduced** ("damped")
- J_{DLS} can be used for both $m = n$ and $m < n$ cases

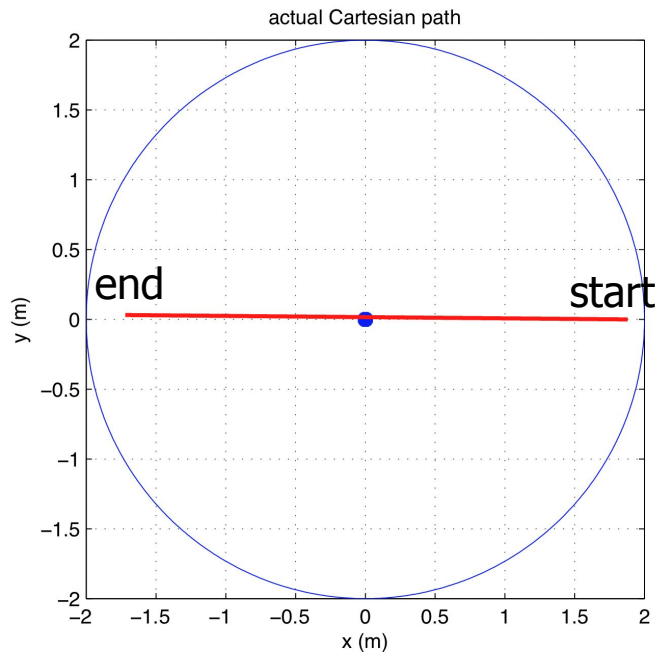


Simulation results

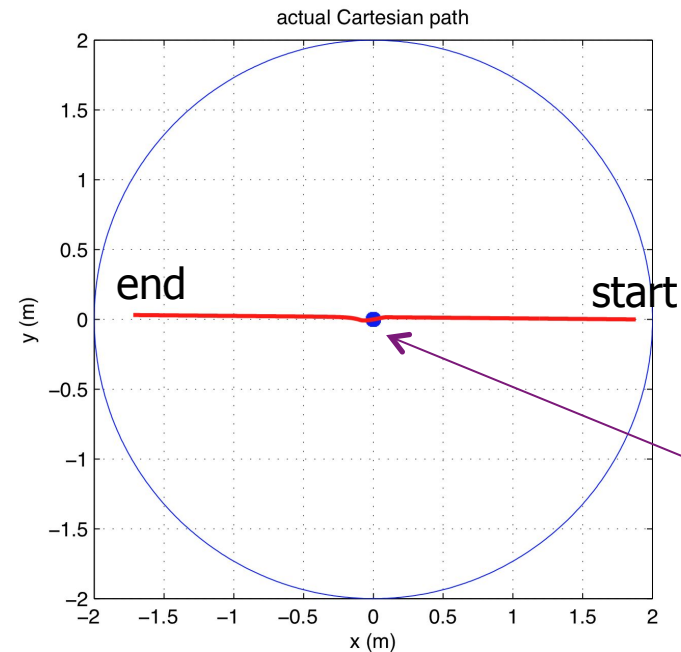
planar 2R robot in straight line Cartesian motion

a comparison of inverse and damped inverse Jacobian methods
even closer to singular case

$$\dot{q} = J^{-1}(q) v$$



$$\dot{q} = J_{DLS}(q) v$$



a line from right to left, at $\alpha=179.5^\circ$ angle with x-axis,
executed at constant speed $v=0.6$ m/s for $T=6$ s



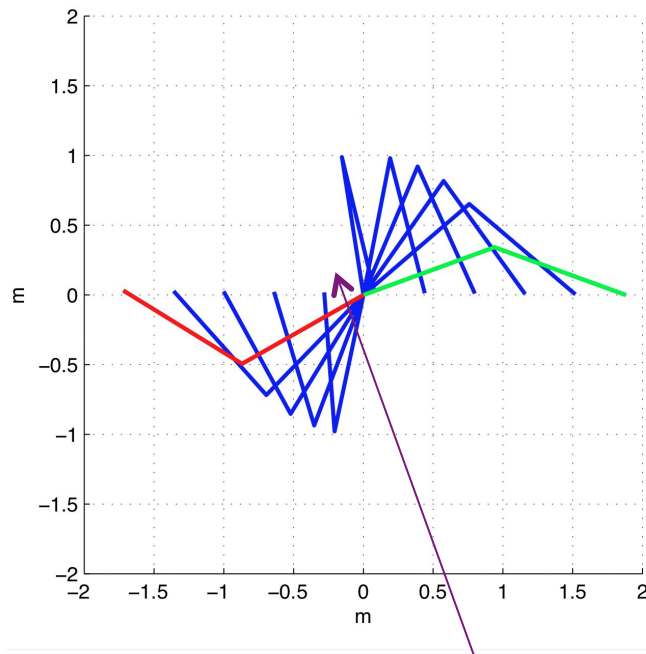
Simulation results

planar 2R robot in straight line Cartesian motion

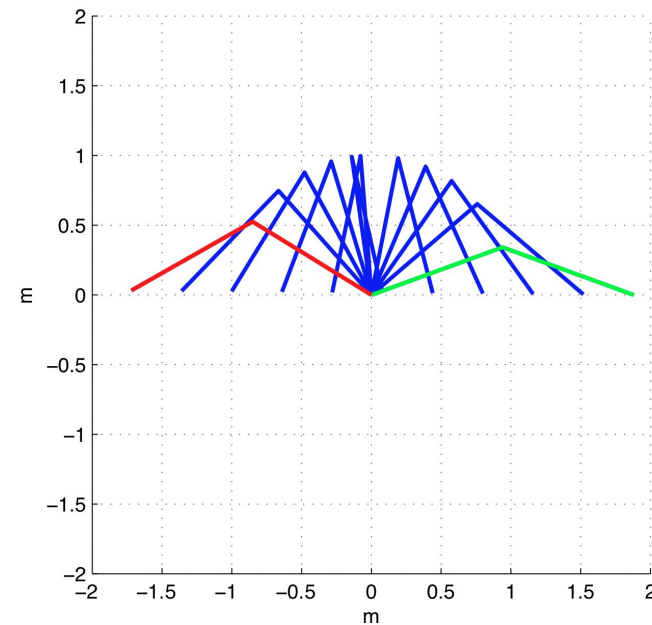
$$\dot{q} = J^{-1}(q) v$$

path at
 $\alpha = 179.5^\circ$

$$\dot{q} = J_{DLS}(q) v$$



here, a **very fast** reconfiguration of first joint ...



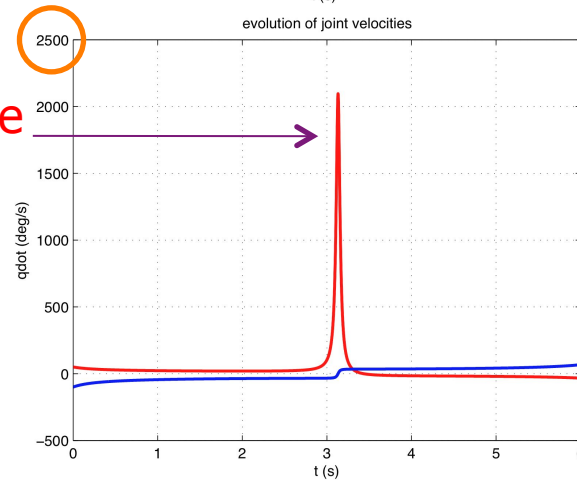
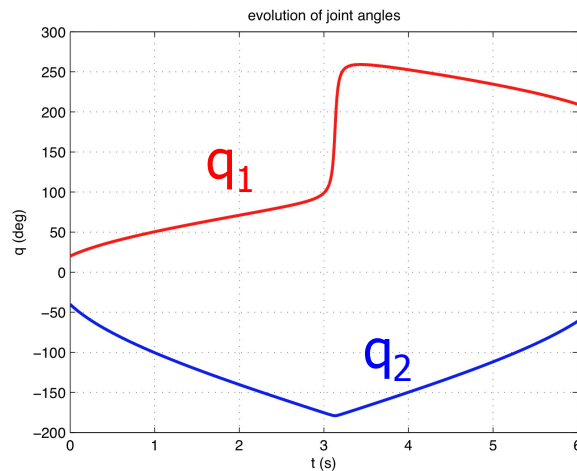
a completely **different inverse solution**, around/after crossing the region close to the folded singularity



Simulation results

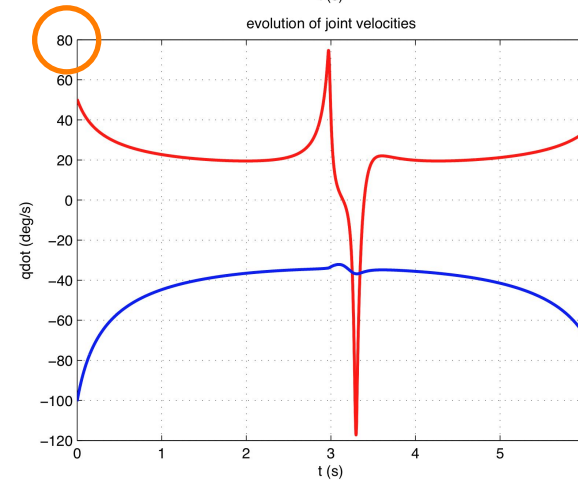
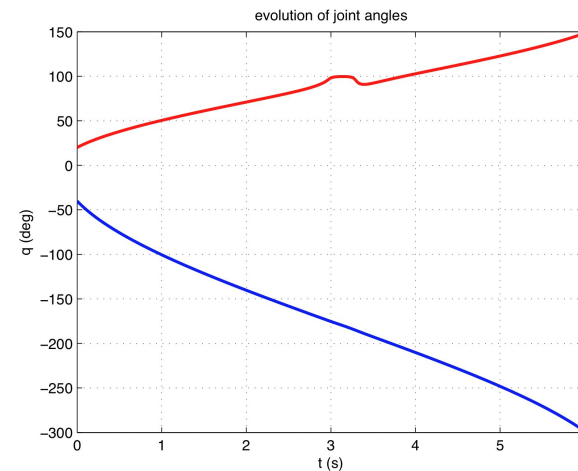
planar 2R robot in straight line Cartesian motion

$$\dot{q} = J^{-1}(q) v$$



extremely large peak velocity of first joint!!

$$\dot{q} = J_{DLS}(q) v$$



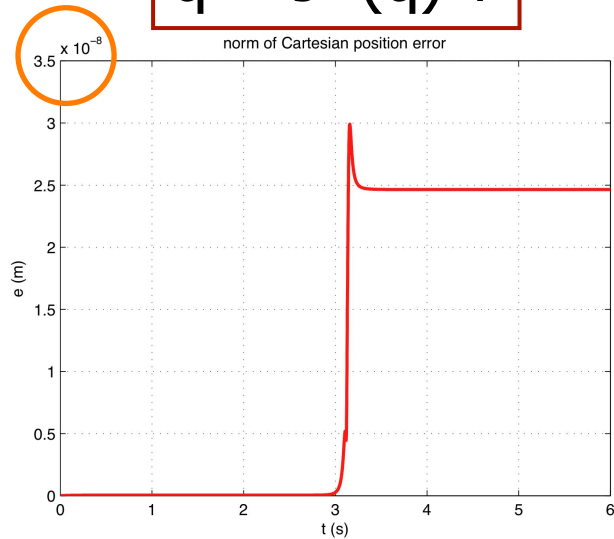
smooth joint motion with limited joint velocities!



Simulation results

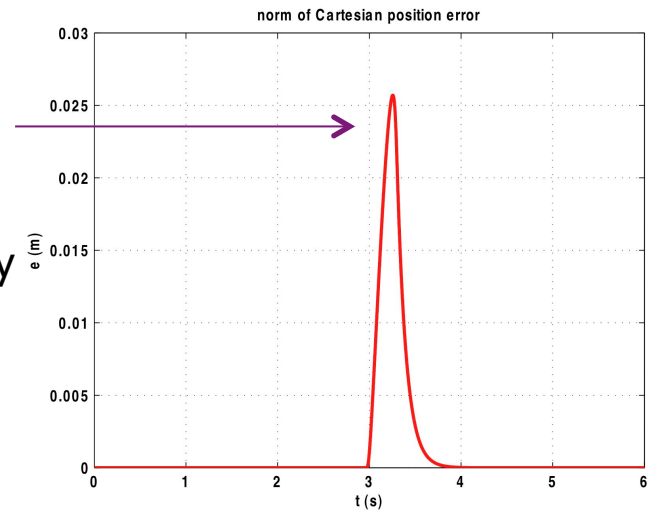
planar 2R robot in straight line Cartesian motion

$$\dot{q} = J^{-1}(q) v$$



increased numerical integration error ($3 \cdot 10^{-8}$)

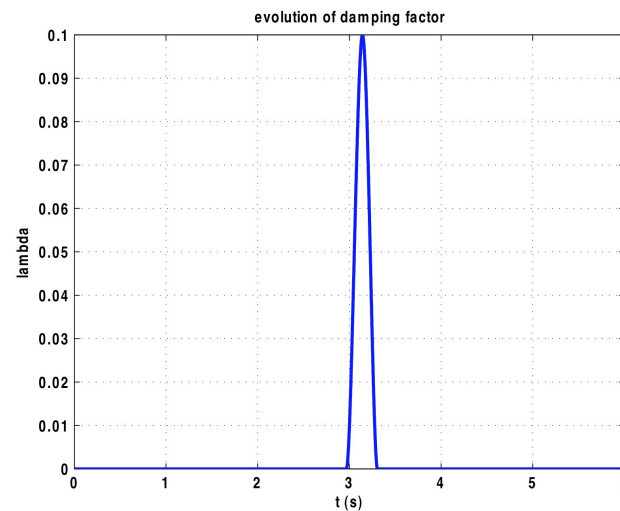
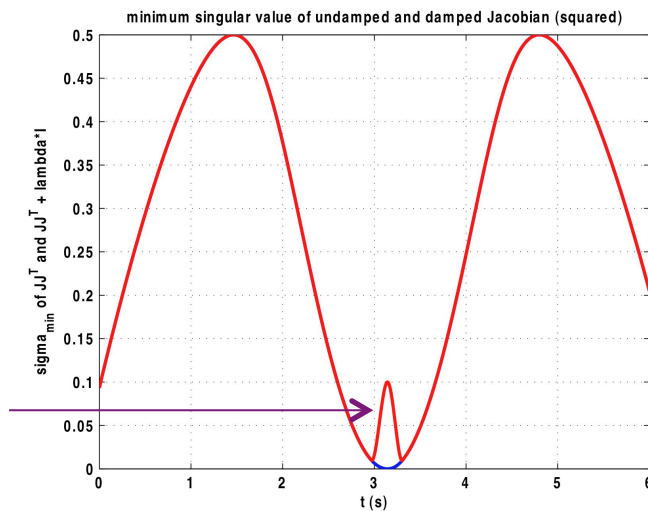
$$\dot{q} = J_{DLS}(q) v$$



error (25 mm) when crossing the singularity, later recovered by feedback action ($v \Rightarrow v+Ke$)

minimum singular value of JJ^T and $\lambda I + JJ^T$

they differ only when damping factor is non-zero



damping factor λ is chosen non-zero only close to singularity!



Use of the pseudo-inverse

a constrained optimization (minimum norm) problem

$$\min_{\dot{q}} H = \frac{1}{2} \|\dot{q}\|^2 \quad \text{such that} \quad J\dot{q} - v = 0$$

solution

$$\dot{q} = J^\# v$$

pseudo-inverse of J

- if $v \in \mathcal{R}(J)$, the constraint is satisfied (v is feasible)
- else $J\dot{q} = v^\perp$ where v^\perp minimizes the error $\|J\dot{q} - v\|$

orthogonal projection of v on $\mathcal{R}(J)$



Properties of the pseudo-inverse

it is the **unique** matrix that satisfies the **four** relationships

- $JJ^\#J = J \quad J^\#JJ^\# = J^\#$

$$(J^\#J)^T = J^\#J \quad (JJ^\#)^T = JJ^\#$$

- if rank $\rho = m = n$: $J^\# = J^{-1}$

- if $\rho = m < n$: $J^\# = J^T (JJ^T)^{-1}$

it **always** exists and is computed in general numerically using the SVD = Singular Value Decomposition of J
(e.g., with the MATLAB function **pinv**)



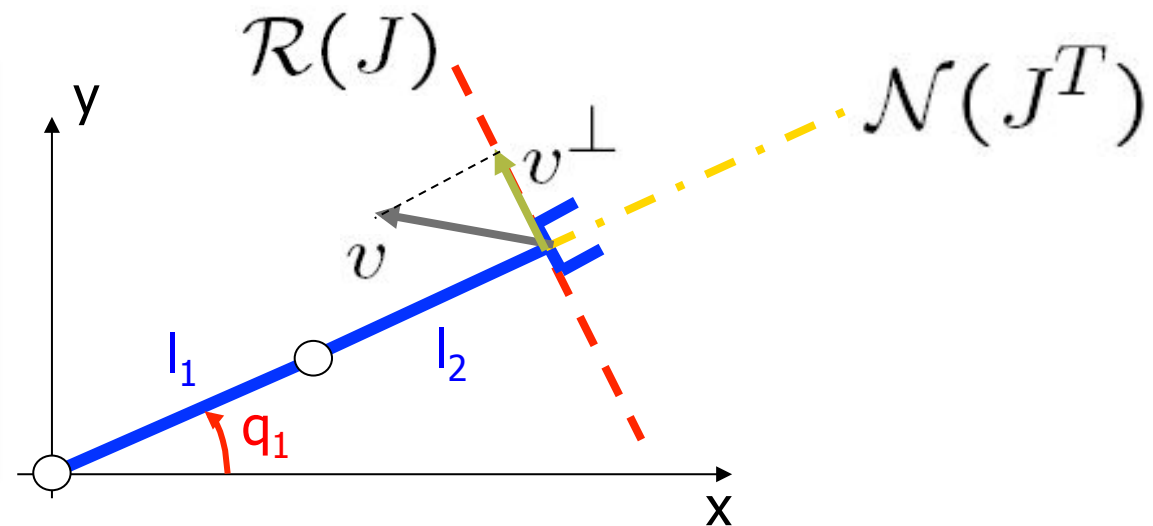
Numerical example

Jacobian of 2R arm with $l_1 = l_2 = 1$ and $q_2 = 0$ (rank $\rho = 1$)

$$J = \begin{bmatrix} -2s_1 & -s_1 \\ 2c_1 & c_1 \end{bmatrix} \quad J^\# = \frac{1}{5} \begin{bmatrix} -2s_1 & 2c_1 \\ -s_1 & c_1 \end{bmatrix}$$

$$\dot{q} = J^\# v$$

is the minimum norm joint velocity vector that realizes v^\perp





General solution for $m < n$

all solutions (an infinite number) of the inverse differential kinematics problem can be written as

$$\dot{q} = J^\# v + (I - J^\# J) \xi$$

← any joint velocity...

“projection” matrix in the kernel of J

this is **also** the solution to a slightly **modified** constrained optimization problem (biased toward the joint velocity ξ , chosen to avoid obstacles, joint limits, etc.)

$$\min H = \frac{1}{2} \|\dot{q} - \xi\|^2 \quad \text{such that} \quad J\dot{q} - v = 0$$

verification of which **actual** task velocity is going to be obtained

$$v_{actual} = J\dot{q} = J(J^\# v + (I - J^\# J)\xi) = JJ^\# v + \cancel{(J - JJ^\# J)\xi} = JJ^\# (Jw) = (JJ^\# J)w = Jw = v$$

↑
if $v \in \mathfrak{R}(J) \Rightarrow v = Jw$, for some w



Higher-order differential inversion

- inversion of motion from task to joint space can be performed also at a **higher** differential level

- **acceleration**-level: given q, \dot{q}

$$\ddot{q} = J_r^{-1}(q) (\ddot{r} - \dot{J}_r(q)\dot{q})$$

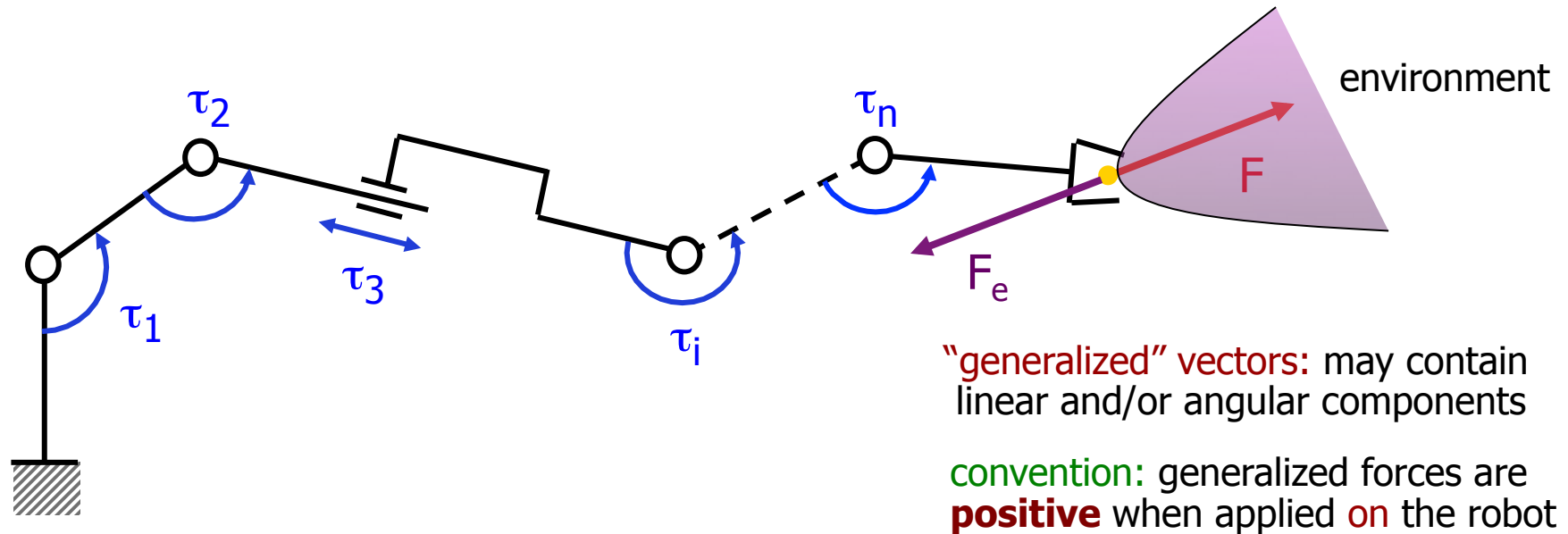
- **jerk**-level: given q, \dot{q}, \ddot{q}

$$\ddot{q} = J_r^{-1}(q) (\ddot{r} - \dot{J}_r(q)\ddot{q} - 2\ddot{J}_r(q)\dot{q})$$

- the (inverse) of the Jacobian is always the **leading** term
- **smoother** joint motions are expected (at least, due to the existence of higher-order time derivatives $\ddot{r}, \ddot{\ddot{r}}, \dots$)



Generalized forces and torques



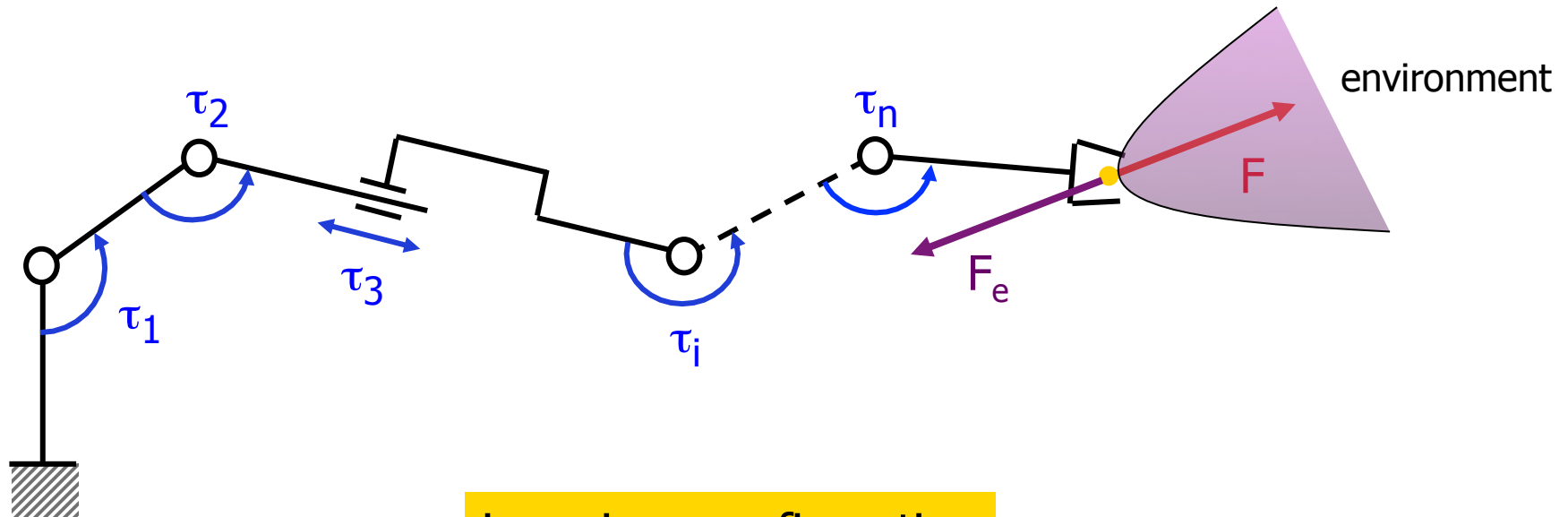
“generalized” vectors: may contain linear and/or angular components

convention: generalized forces are **positive** when applied **on** the robot

- τ = forces/torques exerted **by the motors** at the robot joints
- F = **equivalent** forces/torques exerted at the robot end-effector
- F_e = forces/torques exerted **by the environment** at the end-effector
- principle of action and reaction: $F_e = -F$
*reaction from environment is **equal and opposite** to the robot action on it*



Transformation of forces – Statics



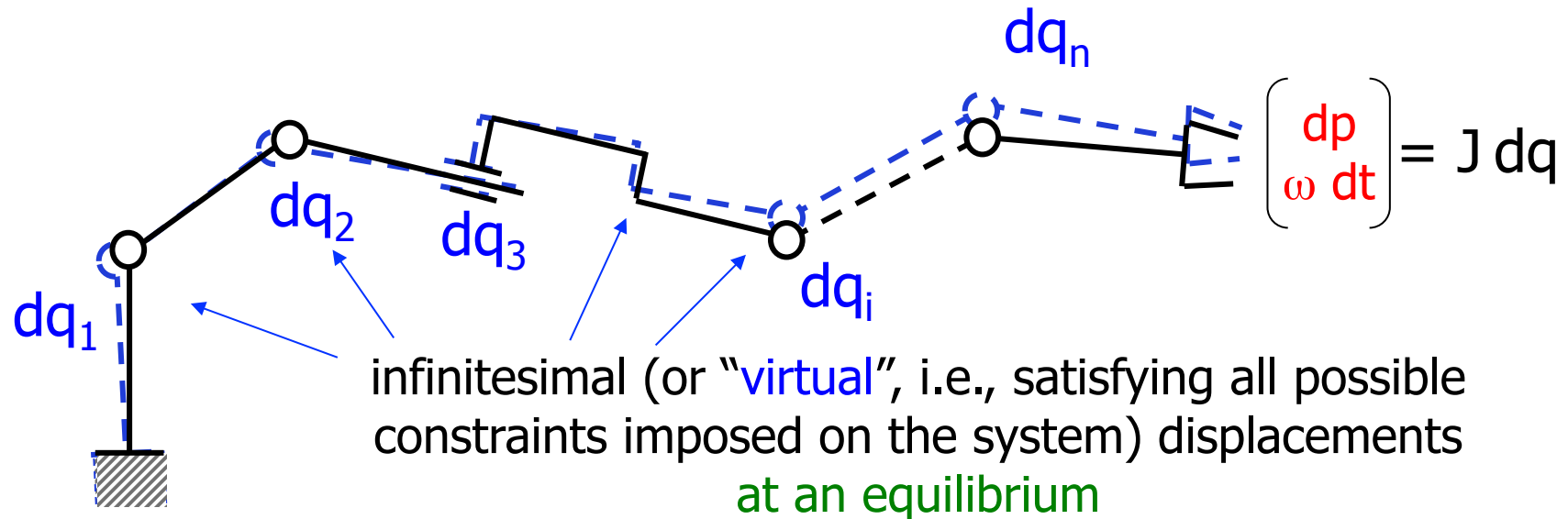
in a given configuration

- what is the transformation between F at robot end-effector and τ at joints?
in **static equilibrium** conditions (i.e., **no motion**):
- what F will be exerted on environment by a τ applied at the robot joints?
- what τ at the joints will balance a $F_e (= -F)$ exerted by the environment?

all equivalent formulations



Virtual displacements and works

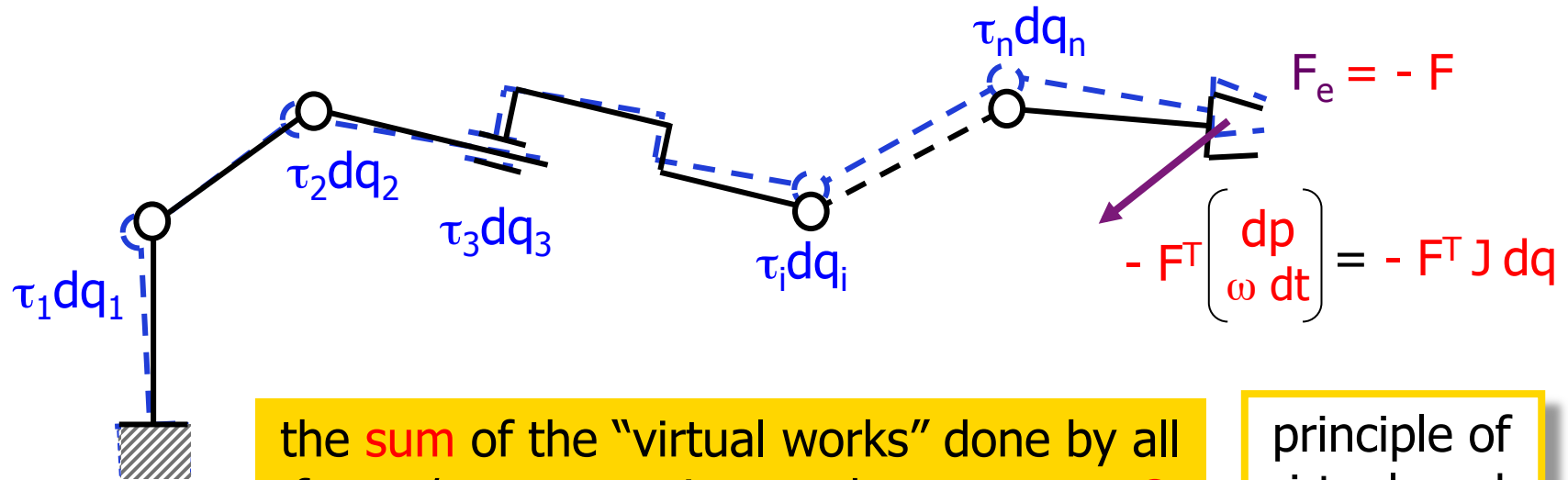


- without kinetic energy variation (zero acceleration)
- without dissipative effects (zero velocity)

the "virtual work" is the work done by all forces/torques acting **on** the system for a given virtual displacement



Principle of virtual work



the **sum** of the “virtual works” done by all forces/torques acting **on** the system = **0**

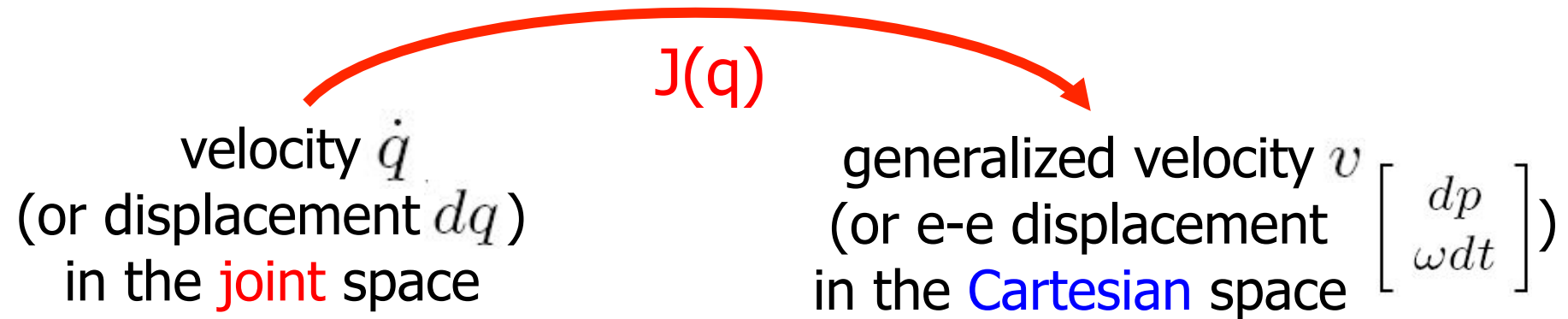
principle of virtual work

$$\tau^T dq - F^T \begin{bmatrix} dp \\ \omega dt \end{bmatrix} = \tau^T dq - F^T J dq = 0 \quad \boxed{\forall dq}$$

➔ $\tau = J^T(q)F$



Duality between velocity and force



the singular configurations
for the **velocity map** are the **same**
as those for the **force map**

$$\rho(J) = \rho(J^T)$$

Dual subspaces of velocity and force

summary of definitions



$$\begin{aligned}\mathcal{R}(J) &= \{v \in \mathbb{R}^m : \exists \dot{q} \in \mathbb{R}^n, J\dot{q} = v\} \\ \mathcal{N}(J^T) &= \{v \in \mathbb{R}^m : \nexists \dot{q} \in \mathbb{R}^n, J\dot{q} = v\} \\ &= \{F \in \mathbb{R}^m : J^T F = 0\} \\ \mathcal{R}(J) + \mathcal{N}(J^T) &= \mathbb{R}^m\end{aligned}$$

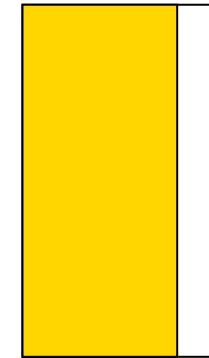
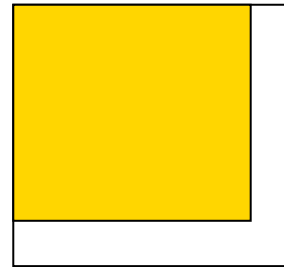
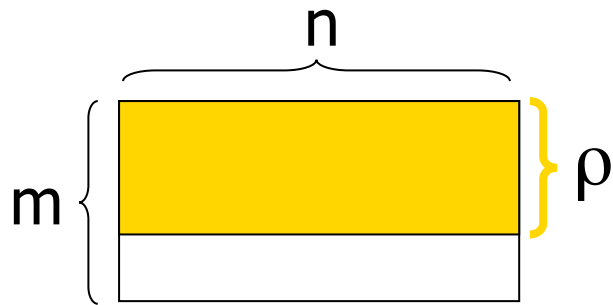
$$\begin{aligned}\mathcal{R}(J^T) &= \{\tau \in \mathbb{R}^n : \exists F \in \mathbb{R}^m, J^T F = \tau\} \\ \mathcal{N}(J) &= \{\tau \in \mathbb{R}^n : \nexists F \in \mathbb{R}^m, J^T F = \tau\} \\ &= \{\dot{q} \in \mathbb{R}^n : J\dot{q} = 0\} \\ \mathcal{R}(J^T) + \mathcal{N}(J) &= \mathbb{R}^n\end{aligned}$$

Velocity and force singularities

list of possible cases



$$\rho = \text{rank}(J) = \text{rank}(J^T) \leq \min(m, n)$$



1. $\rho < m$

$$\exists \dot{q} \neq 0 : J\dot{q} = 0$$

$$\exists F \neq 0 : J^T F = 0$$

2. $\rho = m$

$$\exists \dot{q} \neq 0 : J\dot{q} = 0$$

$$\mathcal{N}(J^T) = \{0\}$$

1. $\det J = 0$

$$\exists \dot{q} \neq 0 : J\dot{q} = 0$$

$$\exists F \neq 0 : J^T F = 0$$

2. $\det J \neq 0$

$$\mathcal{N}(J) = \{0\}$$

$$\mathcal{N}(J^T) = \{0\}$$

1. $\rho < n$

$$\exists \dot{q} \neq 0 : J\dot{q} = 0$$

$$\exists F \neq 0 : J^T F = 0$$

2. $\rho = n$

$$\exists F \neq 0 : J^T F = 0$$

$$\mathcal{N}(J) = \{0\}$$



Example of singularity analysis

planar 2R arm with generic link lengths l_1 and l_2

$$J(q) = \begin{bmatrix} -l_1 s_1 - l_2 s_{12} & -l_2 s_{12} \\ l_1 c_1 + l_2 c_{12} & l_2 c_{12} \end{bmatrix} \quad \det J(q) = l_1 l_2 s_2$$

singularity at $q_2 = 0$ (arm straight) \Rightarrow

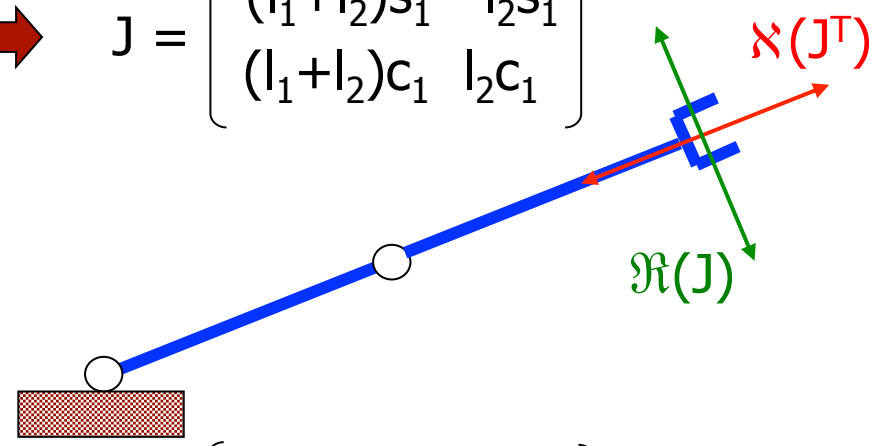
$$J = \begin{bmatrix} -(l_1 + l_2) s_1 & -l_2 s_1 \\ (l_1 + l_2) c_1 & l_2 c_1 \end{bmatrix}$$

$$\mathfrak{R}(J) = \alpha \begin{bmatrix} -s_1 \\ c_1 \end{bmatrix}$$

$$\mathfrak{N}(J^T) = \alpha \begin{bmatrix} c_1 \\ s_1 \end{bmatrix}$$

$$\mathfrak{R}(J^T) = \beta \begin{bmatrix} l_1 + l_2 \\ l_2 \end{bmatrix}$$

$$\mathfrak{N}(J) = \beta \begin{bmatrix} l_2 \\ -(l_1 + l_2) \end{bmatrix}$$



singularity at $q_2 = \pi$ (arm folded) \Rightarrow

$$J = \begin{bmatrix} (l_2 - l_1) s_1 & l_2 s_1 \\ -(l_2 - l_1) c_1 & -l_2 c_1 \end{bmatrix}$$

$\mathfrak{R}(J)$ and $\mathfrak{N}(J^T)$ as above

$$\mathfrak{R}(J^T) = \beta \begin{bmatrix} l_2 - l_1 \\ l_2 \end{bmatrix} \text{ (for } l_1 = l_2, \beta \begin{bmatrix} 0 \\ 1 \end{bmatrix} \text{)}$$

$$\mathfrak{N}(J) = \beta \begin{bmatrix} l_2 \\ -(l_2 - l_1) \end{bmatrix} \text{ (for } l_1 = l_2, \beta \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{)}$$



Velocity manipulability

- in a given configuration, we wish to evaluate how “effective” is the mechanical **transformation** between joint velocities and end-effector velocities
 - “how easily” can the end-effector be moved in the various directions of the task space
 - equivalently, “how far” is the robot from a singular condition
- we consider all end-effector velocities that can be obtained by choosing joint velocity vectors of **unit norm**

$$\dot{q}^T \dot{q} = 1 \quad \rightarrow \quad v^T J^{\#T} J^{\#} v = 1$$

task **velocity** manipulability **ellipsoid**

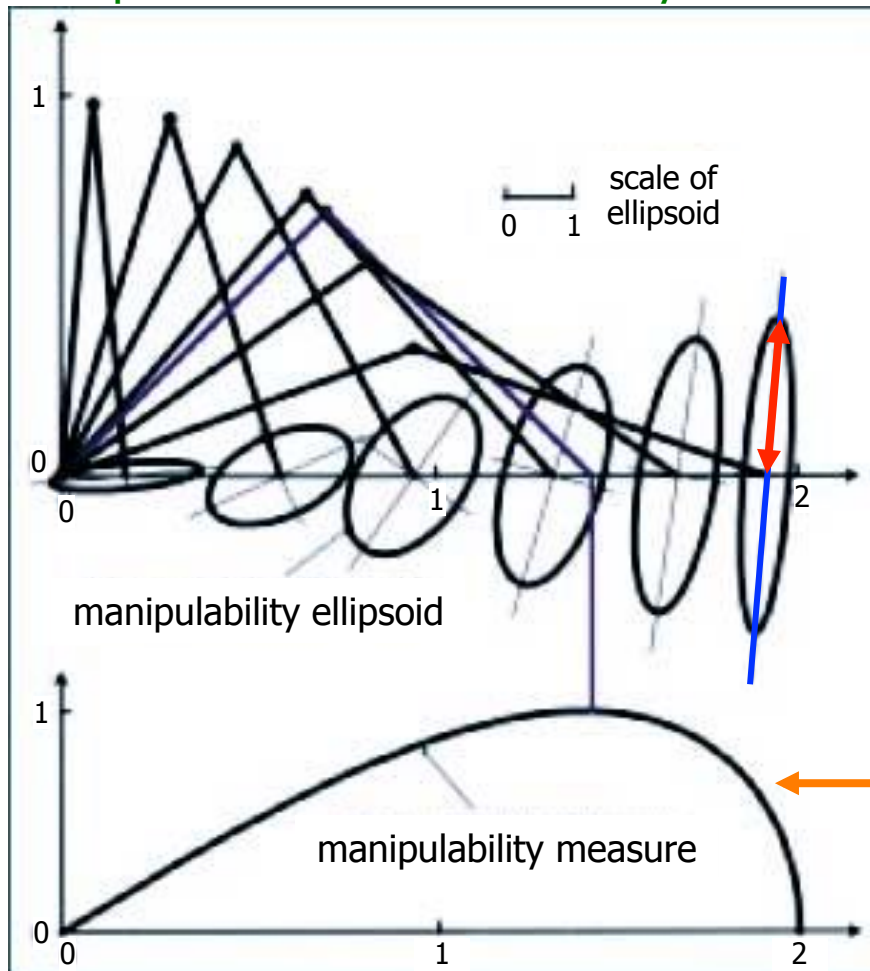
$(JJ^T)^{-1}$ if $\rho = m$

note: the “core” matrix of the ellipsoid equation $v^T A^{-1} v = 1$ is the matrix A !

Manipulability ellipsoid in velocity



planar 2R arm with unitary links



length of principal (semi-)axes:
singular values of J (in its SVD)

$$\sigma_i\{J\} = \sqrt{\lambda_i\{JJ^T\}} \geq 0$$

in a singularity, the ellipsoid
loses a dimension
(for $m=2$, it becomes a segment)

direction of principal axes:
(orthogonal) eigenvectors
associated to λ_i

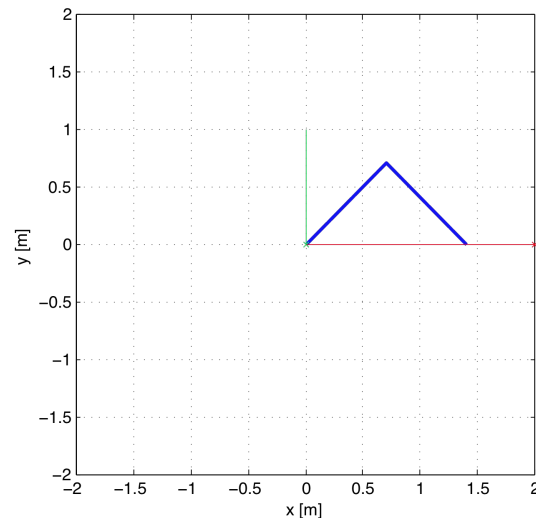
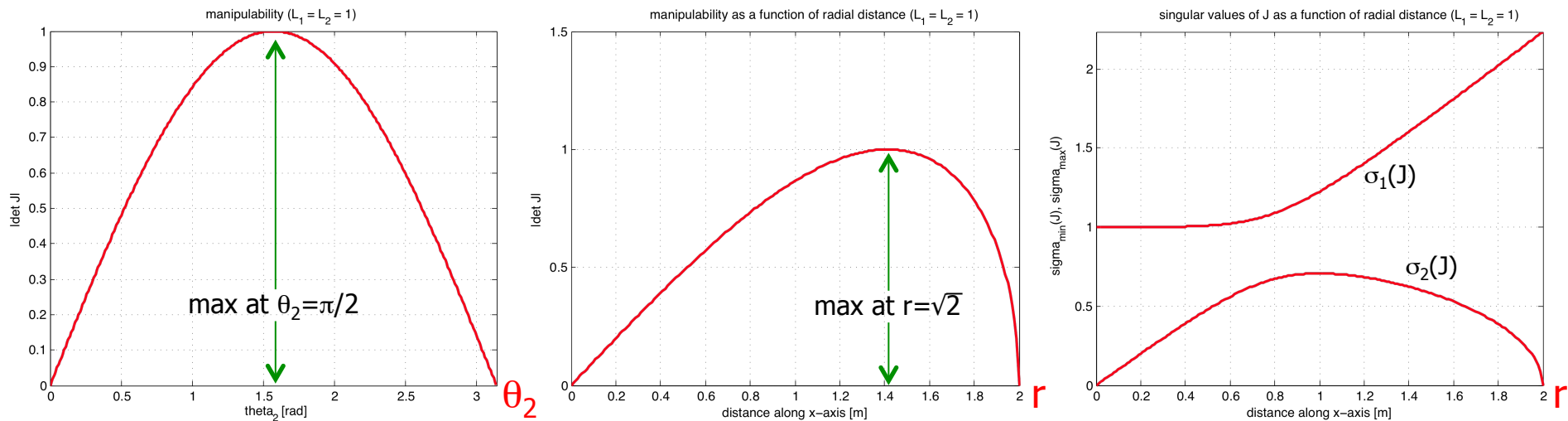
$$w = \sqrt{\det JJ^T} = \prod_{i=1}^m \sigma_i \geq 0$$

proportional to the **volume** of the
ellipsoid (for $m=2$, to its area)



Manipulability measure

planar 2R arm with unitary links: Jacobian J is square $\Rightarrow \sqrt{\det(JJ^T)} = \sqrt{\det J \cdot \det J^T} = |\det J| = \prod_{i=1}^2 \sigma_i$



best posture for manipulation
(similar to a human arm!)

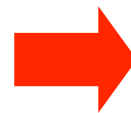
full isotropy is never obtained
in this case, since it always $\sigma_1 \neq \sigma_2$



Force manipulability

- in a given configuration, evaluate how “effective” is the **transformation** between joint torques and end-effector forces
 - “how easily” can the end-effector apply generalized forces (or balance applied ones) in the various directions of the task space
 - in singular configurations, **there are directions** in the task space where external forces/torques are balanced by the robot without the need of **any** joint torque
- we consider all end-effector forces that can be applied (or balanced) by choosing joint torque vectors of **unit norm**

$$\tau^T \tau = 1$$



$$F^T J J^T F = 1$$

same directions of the principal axes of the velocity ellipsoid, but with semi-axes of **inverse** lengths



task **force**
manipulability **ellipsoid**

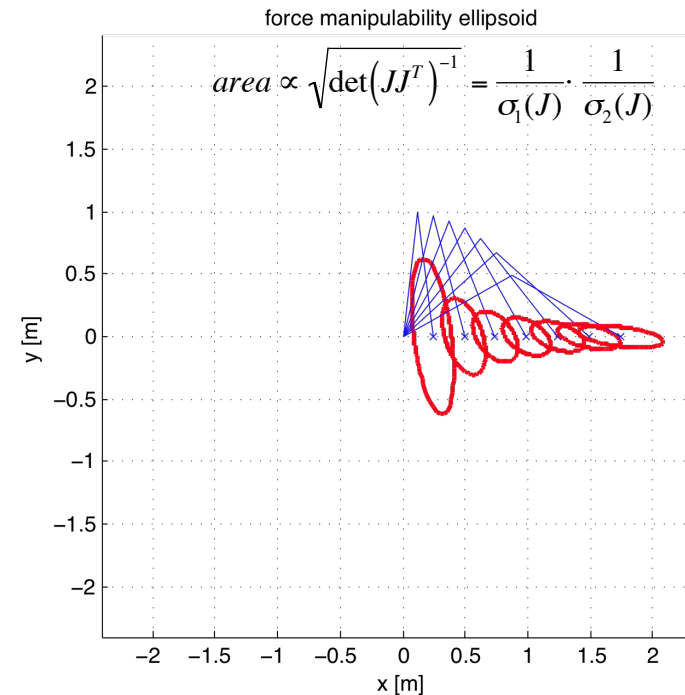
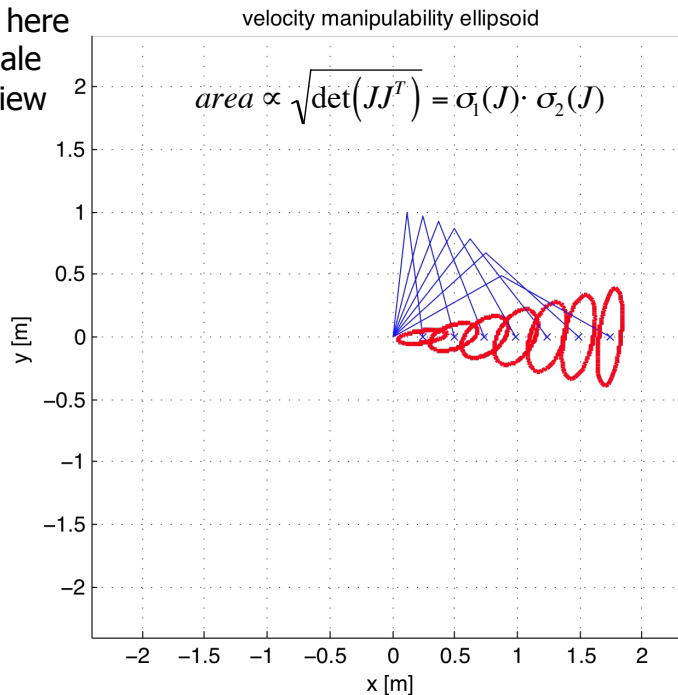
Velocity and force manipulability

dual comparison of actuation vs. control



note:
velocity and force
ellipsoids have here
a different scale
for a better view

planar 2R arm with unitary links



Cartesian **actuation** task (high joint-to-task transformation ratio):
preferred velocity (or force) directions are those where the ellipsoid *stretches*



Cartesian **control** task (low transformation ratio = high resolution):
preferred velocity (or force) directions are those where the ellipsoid *shrinks*



Velocity and force transformations

- the same reasoning made for relating **end-effector to joint** forces/torques (static equilibrium + principle of virtual work) is used also for relating forces and torques applied **at different places of a rigid body and/or** expressed **in different reference frames**

relation among generalized velocities

$$\begin{bmatrix} v_A \\ \omega_A \end{bmatrix} = \begin{bmatrix} {}^A R_B & -{}^A R_B S({}^B r_{BA}) \\ 0 & {}^A R_B \end{bmatrix} \begin{bmatrix} v_B \\ \omega_B \end{bmatrix} = J_{BA} \begin{bmatrix} v_B \\ \omega_B \end{bmatrix}$$

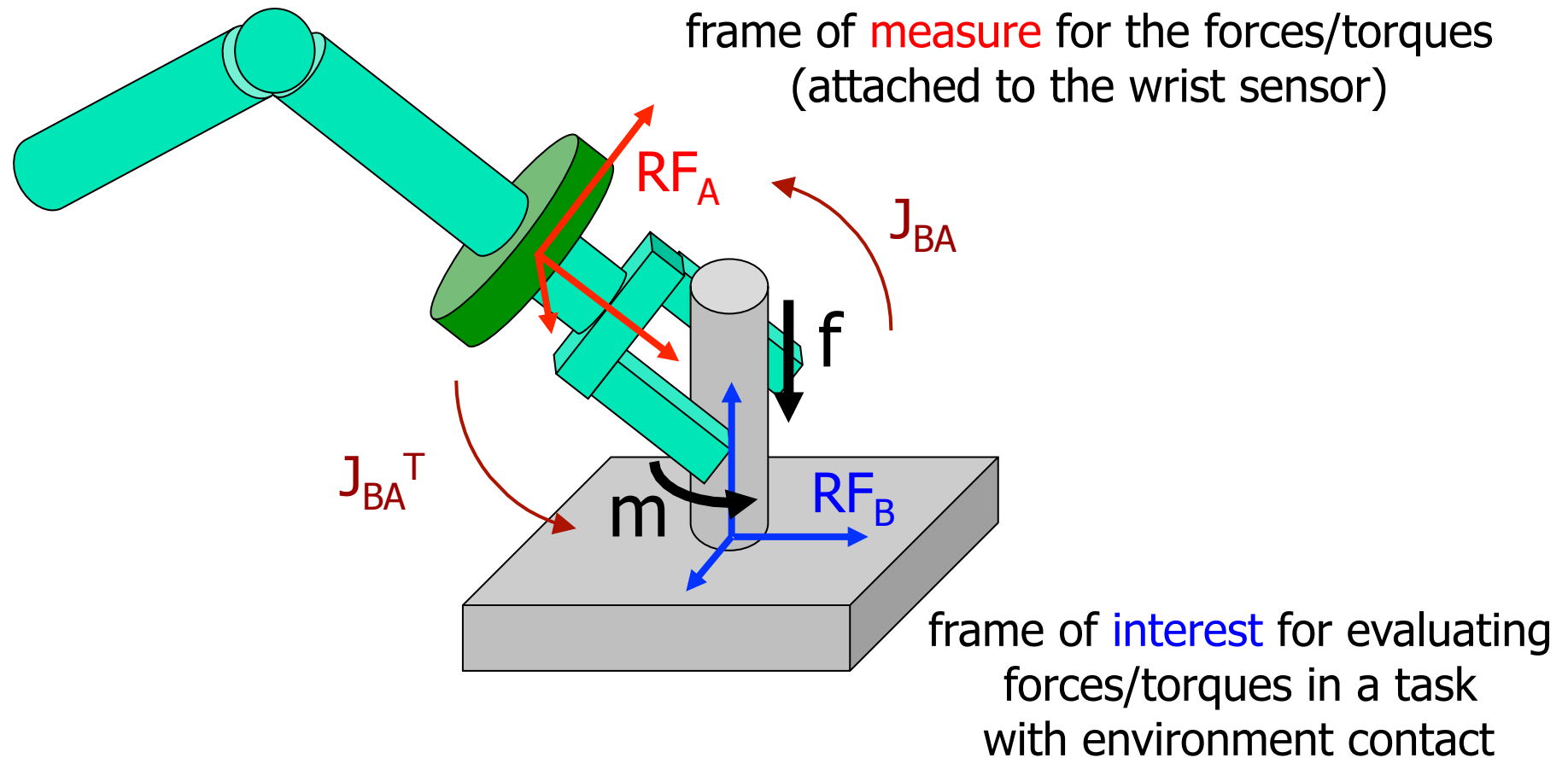


$$\begin{bmatrix} f_B \\ m_B \end{bmatrix} = J_{BA}^T \begin{bmatrix} f_A \\ m_A \end{bmatrix} = \begin{bmatrix} {}^B R_A & 0 \\ S({}^B r_{BA}) {}^B R_A & {}^B R_A \end{bmatrix} \begin{bmatrix} f_A \\ m_A \end{bmatrix}$$

relation among generalized forces

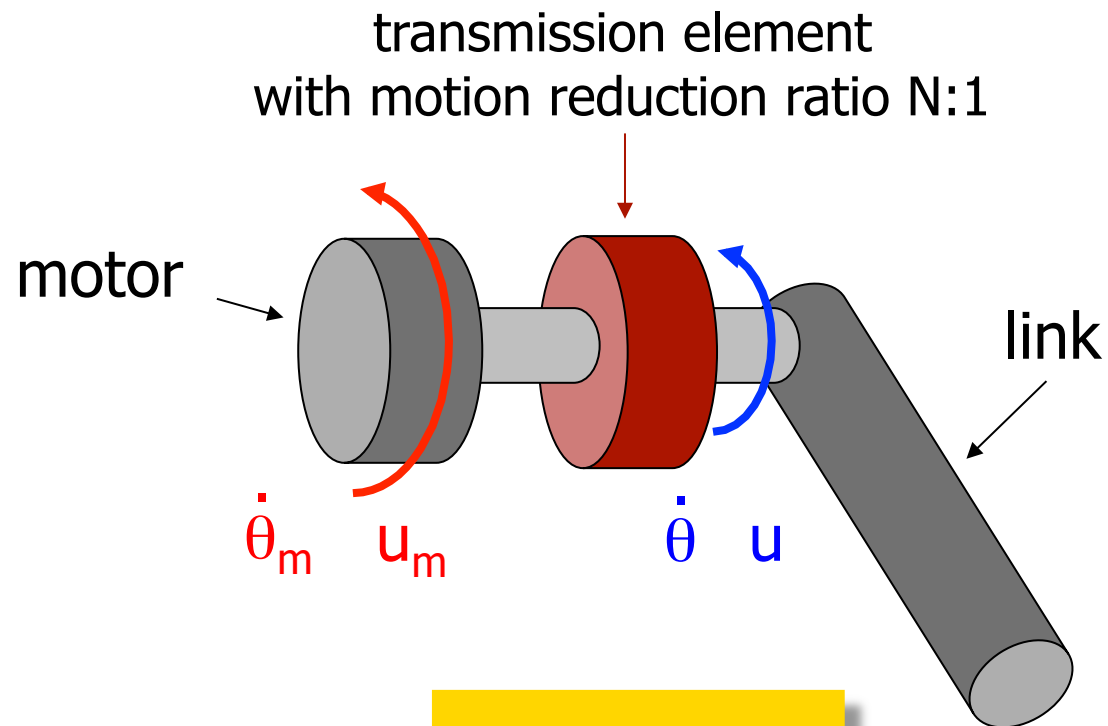


Example 1: 6D force/torque sensor





Example 2: Gear reduction at joints



one of the simplest applications
of the principle of virtual work!

$$\dot{\theta}_m = N\dot{\theta}$$

$$u = Nu_m$$