



Robotics 1

Inverse differential kinematics

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Inversion of differential kinematics

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

generalized velocity

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

J square and non-singular at q


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

- problems
 - **near** a singularity of the Jacobian matrix (too high \dot{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

(here with a square, analytic Jacobian)

current

q



current

next

$$r \longrightarrow r + dr$$



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

first, increment the
desired task variables



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

then, solve the inverse
kinematics problem

(possibly, with a numerical method
from the current configuration)



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

first, solve the inverse
differential kinematics problem



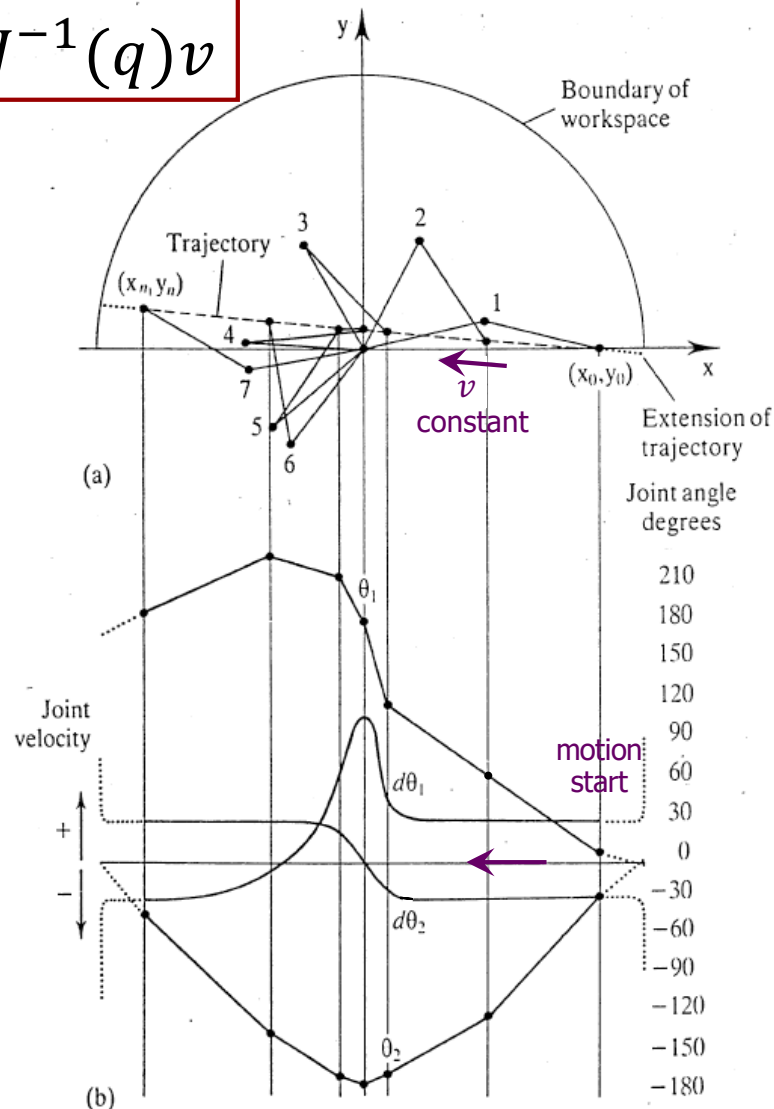
$$q \longrightarrow q + dq$$

then, increment the
original joint variables



Behavior close to 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



Moving close to a singularity in inverse (differential) kinematics problems

- **on-line** inversion of velocities or incremental inverse kinematics
- **singular** configurations for a **6R robot** with **spherical wrist**

wrist

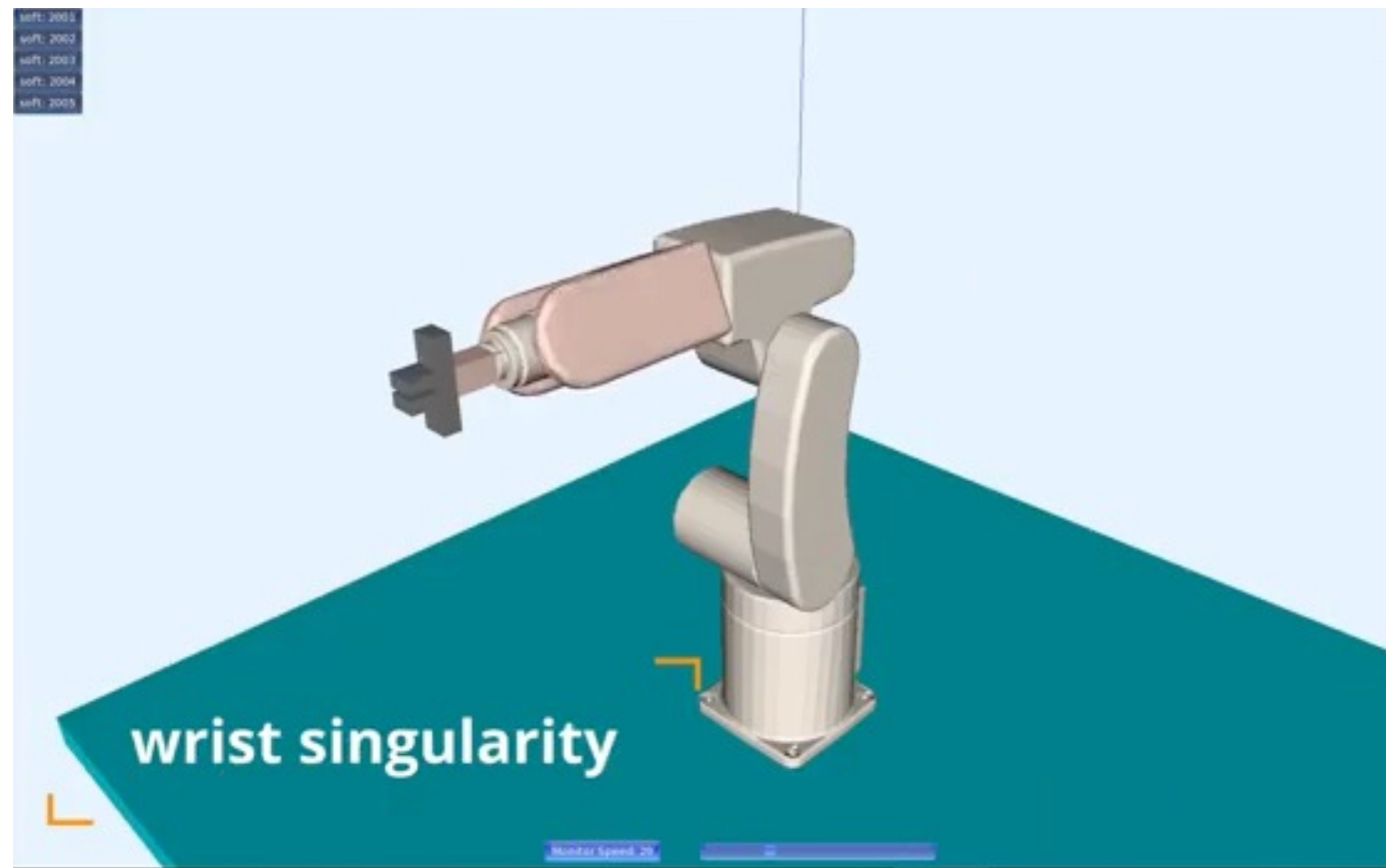
joint axes
4 & 6 aligned

elbow

arm stretched
(or folded)

shoulder

wrist center on
first joint axis



video



Moving close to a singularity

6R KUKA Agile (with spherical wrist)

- wrist, shoulder and elbow **singularities**: feasible **joint** motions **versus** **end-effector** (linear) paths crossing/coming close to critical points



[video](#)

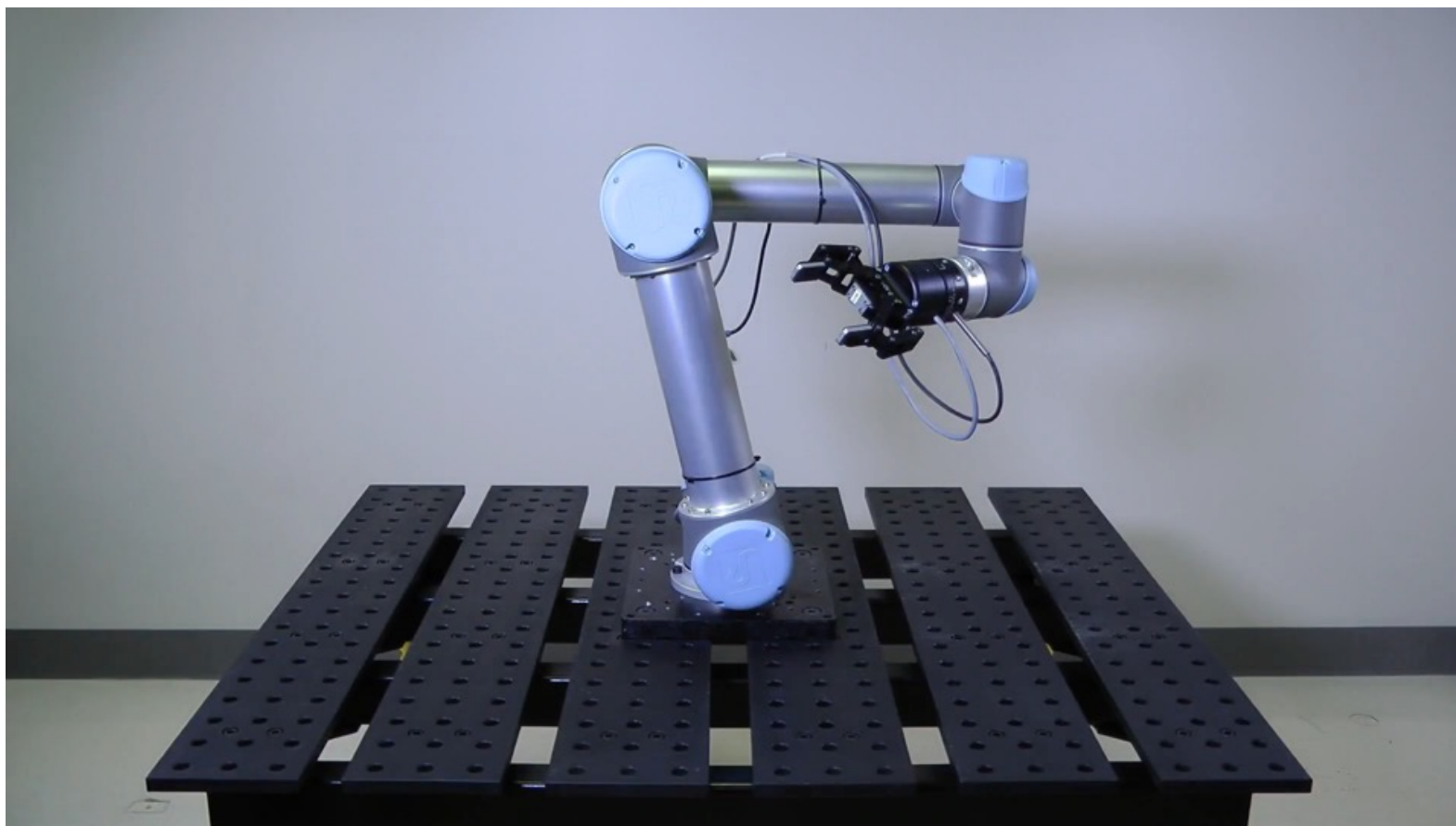
Ecole de Technologie Supérieure, CoR Lab, Montreal



Moving close to a singularity

6R Universal Robots UR5 (no spherical wrist)

- same 'wrist', shoulder, and elbow singularities, though with slightly different configurations and full rotation of joints 4 & 6 in first case



video

Ecole de Technologie Supérieure, CoR Lab, Montreal

all done
in MATLAB

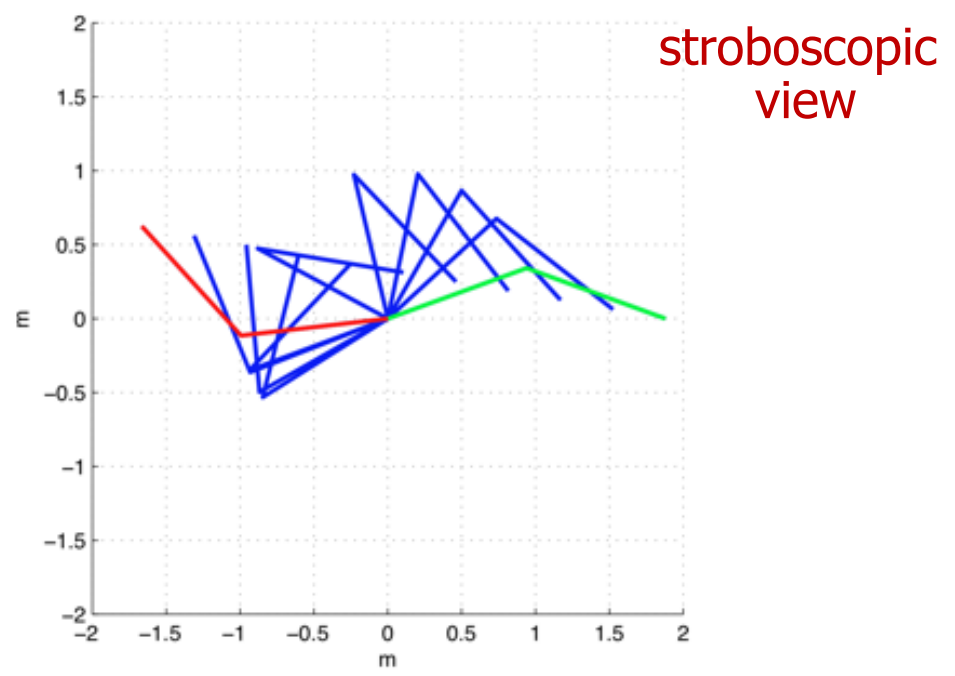
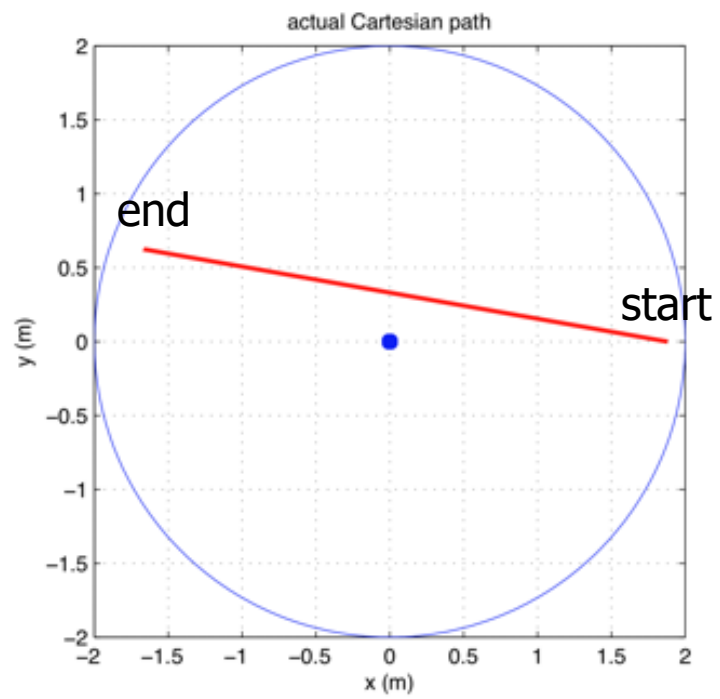


Simulation results

planar 2R robot in straight line Cartesian motion

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

regular case



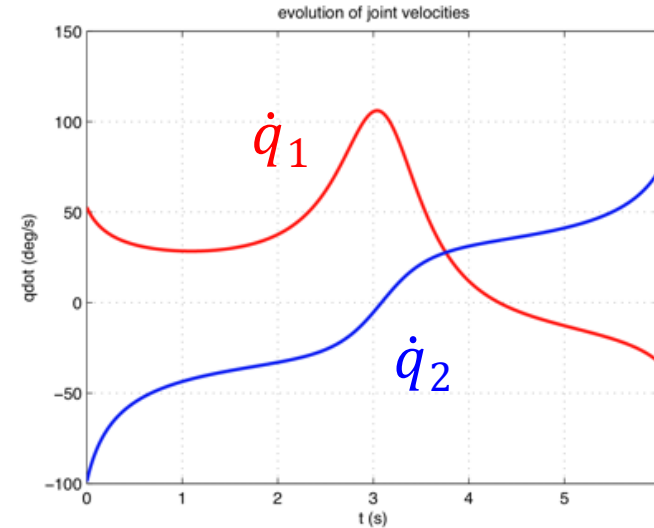
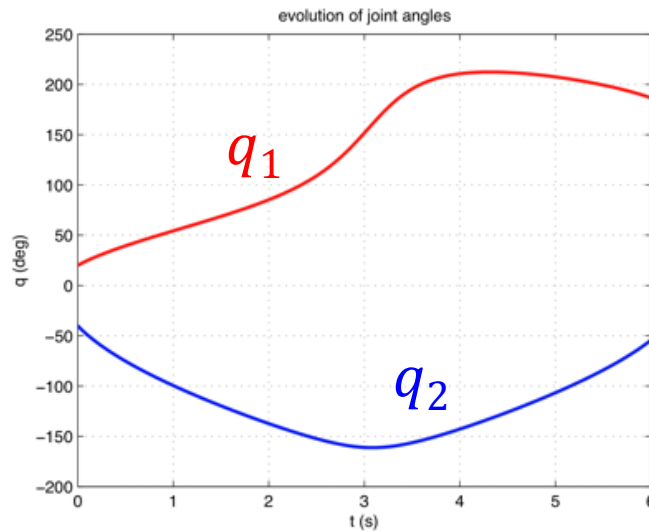
a line from right to left, at $\alpha = 170^\circ$ angle with x -axis, executed at constant speed $v = 0.6$ m/s for $T = 6$ s



Simulation results

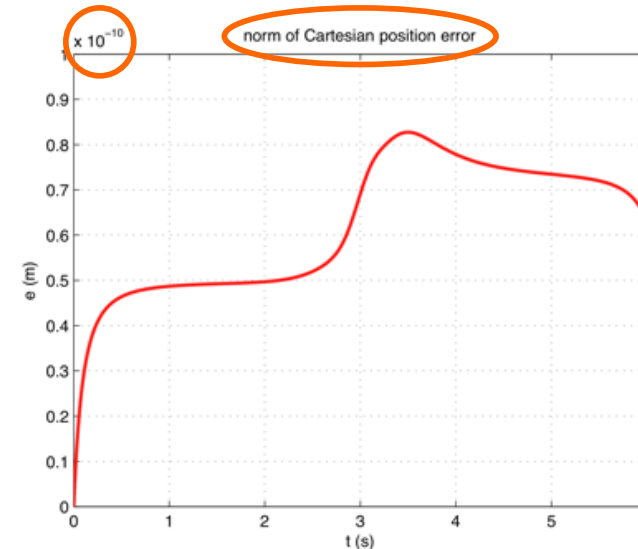
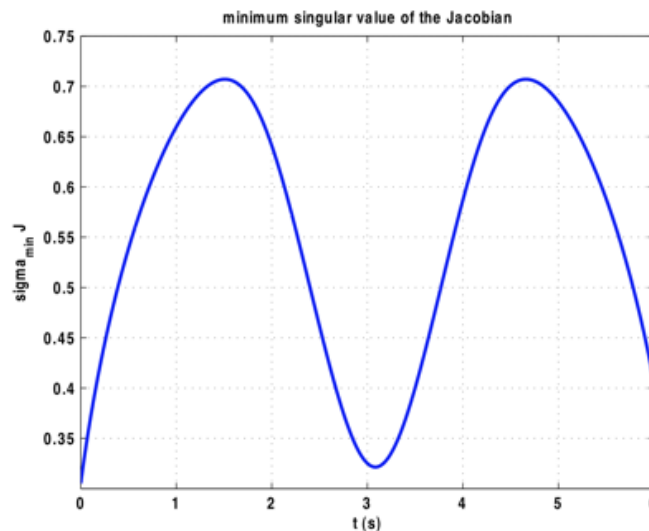
planar 2R robot in straight line Cartesian motion

path at
 $\alpha = 170^\circ$



regular
case

distance to
singularity by
the minimum
singular value
 $\sigma_{min} (= \sigma_2) > 0$
of Jacobian J



error due
only to
numerical
integration
(10^{-10})

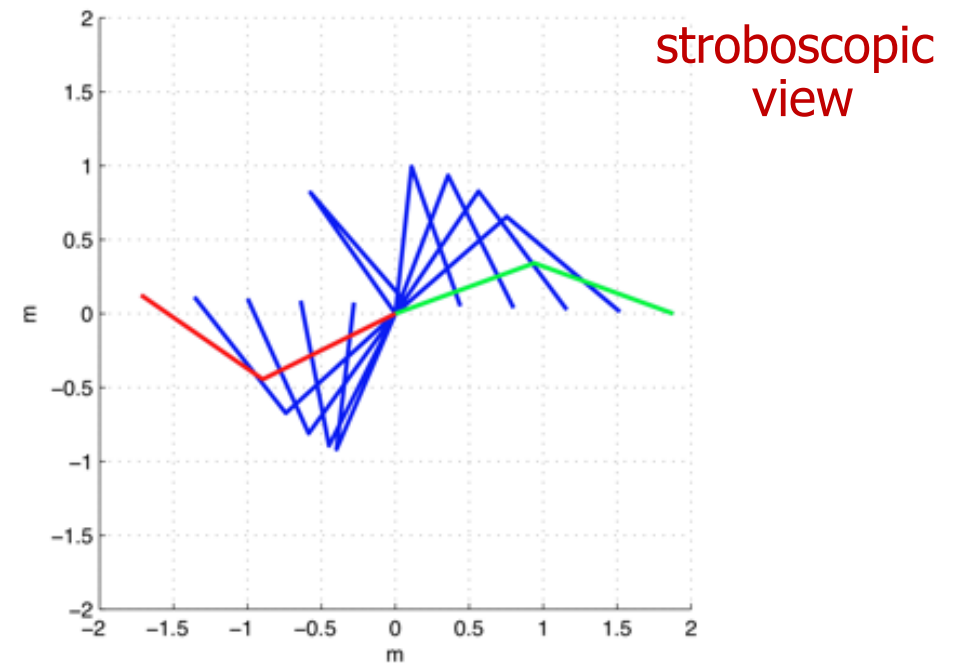
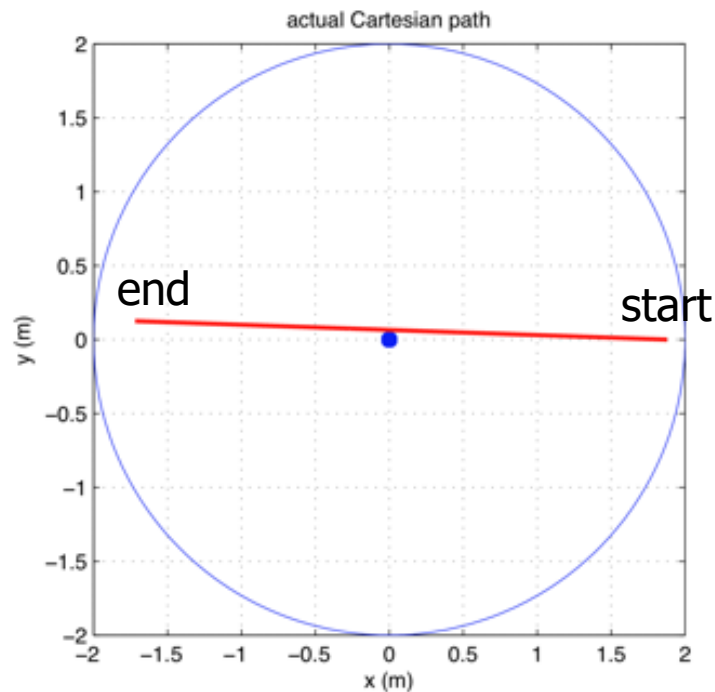


Simulation results

planar 2R robot in straight line Cartesian motion

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

close to singular case



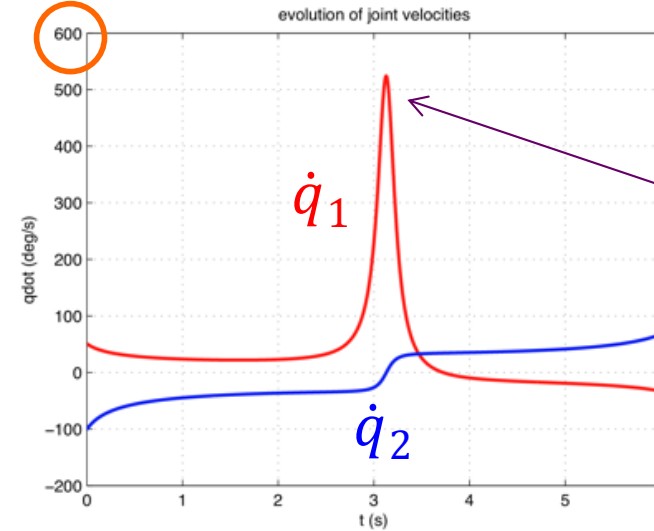
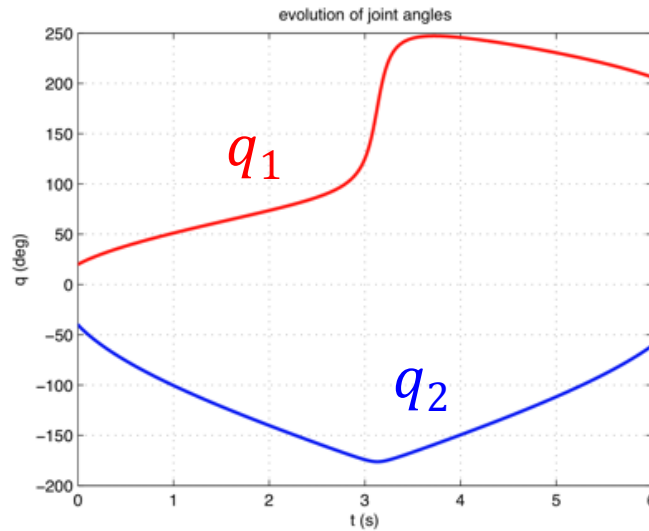
a line from right to left, at $\alpha = 178^\circ$ angle with x -axis,
executed at constant speed $v = 0.6$ m/s for $T = 6$ s



Simulation results

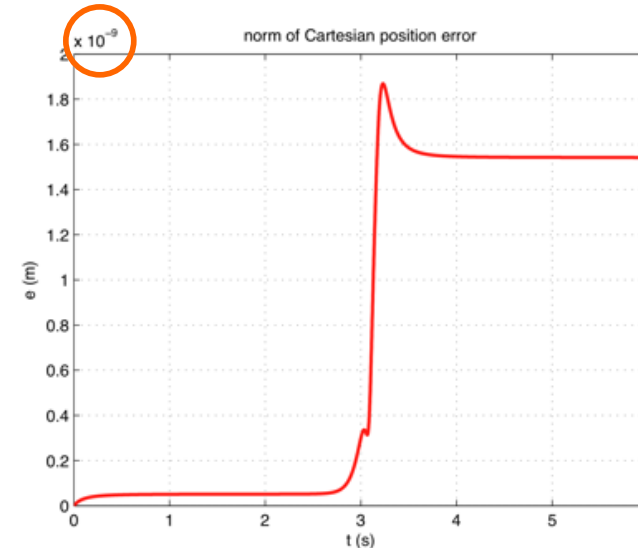
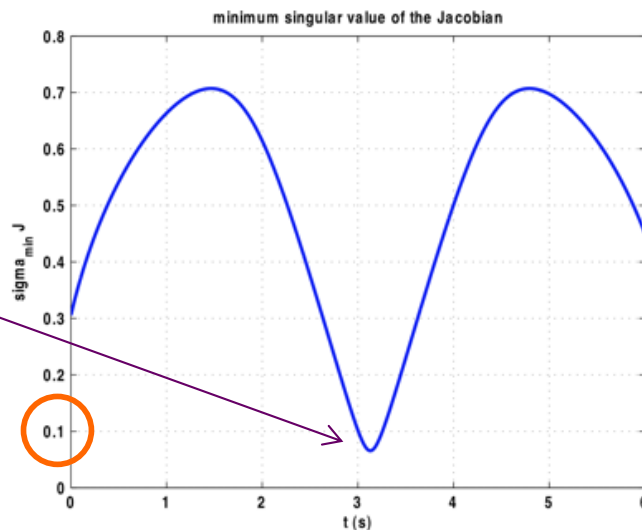
planar 2R robot in straight line Cartesian motion

path at $\alpha = 178^\circ$



large peak of \dot{q}_1

close to singular case



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



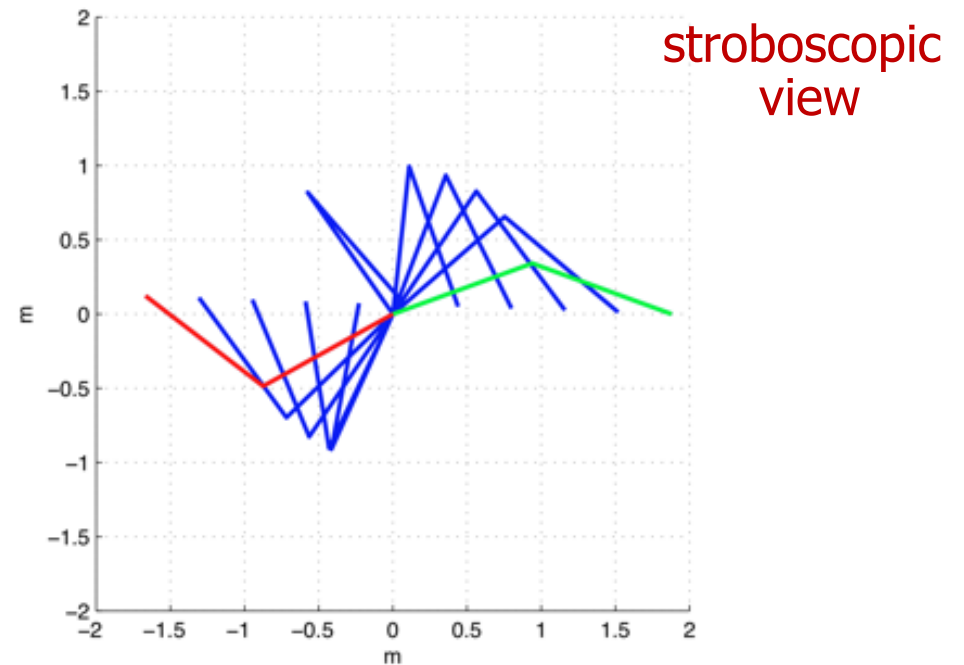
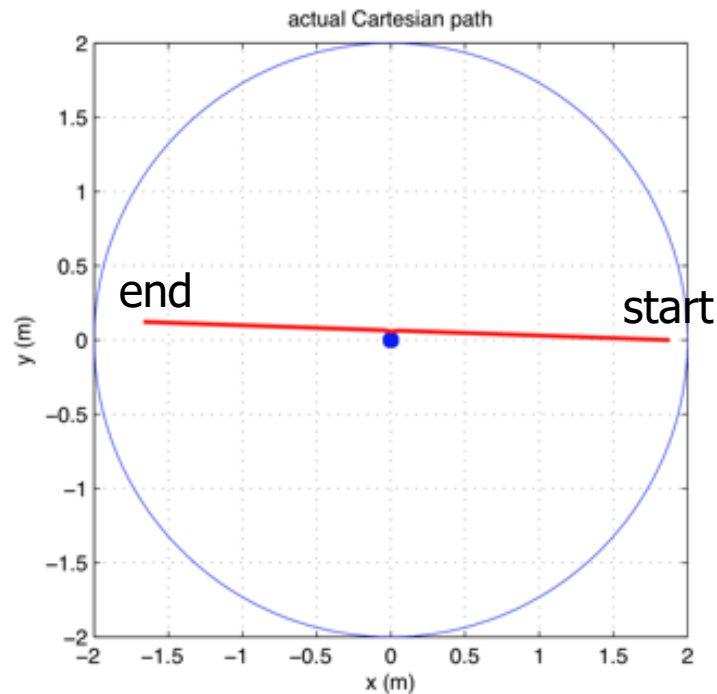
Simulation results

planar 2R robot in straight line Cartesian motion

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

close to singular case

with joint velocity saturation at $V_i = 300^\circ/s$



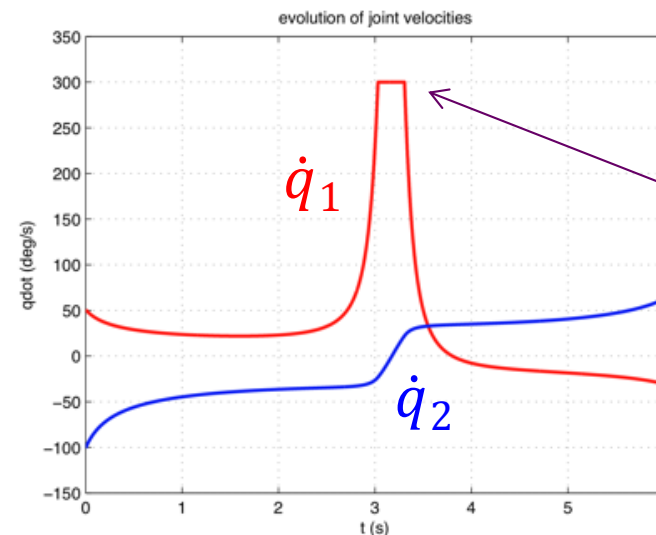
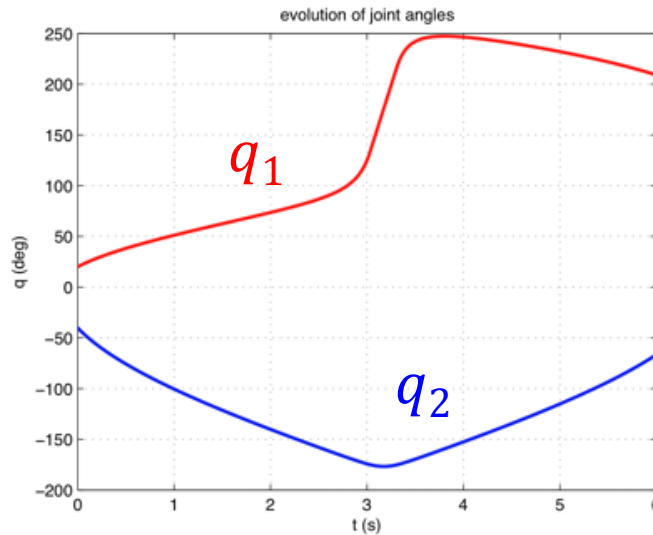
a line from right to left, at $\alpha = 178^\circ$ angle with x -axis, executed at constant speed $v = 0.6$ m/s for $T = 6$ s



Simulation results

planar 2R robot in straight line Cartesian motion

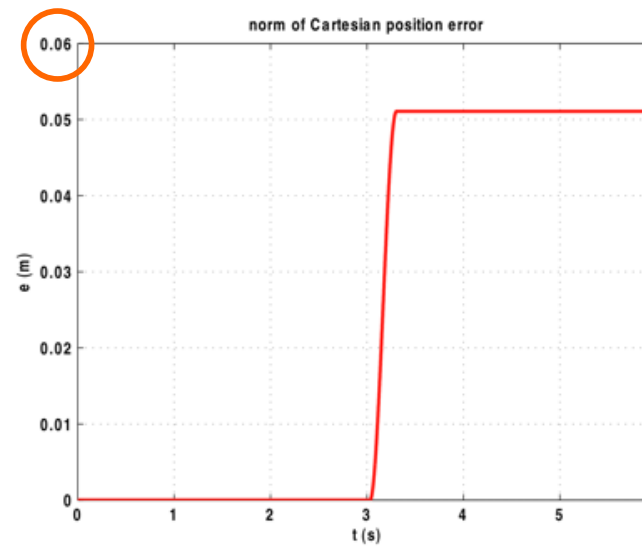
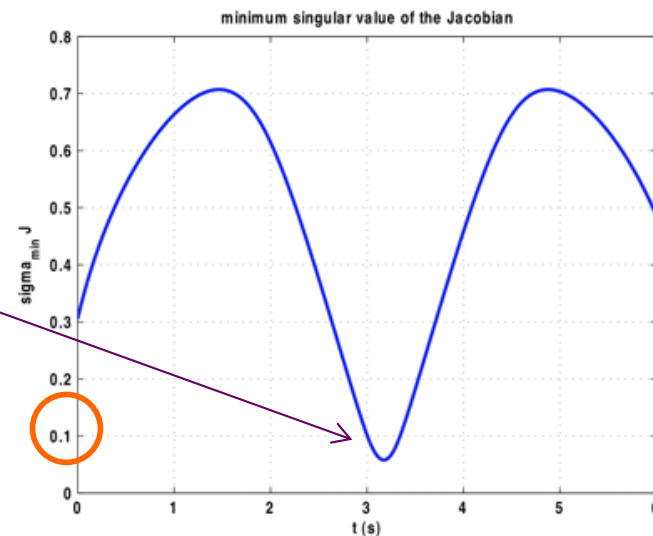
path at $\alpha = 178^\circ$



saturated value of \dot{q}_1



close to singular case



actual position error!! (6 cm)

to be recovered using an error feedback control action!



Damped Least Squares (DLS) method

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

prove it!

prove it!

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

two **equivalent** expressions, but the second is more convenient in redundant robots!

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

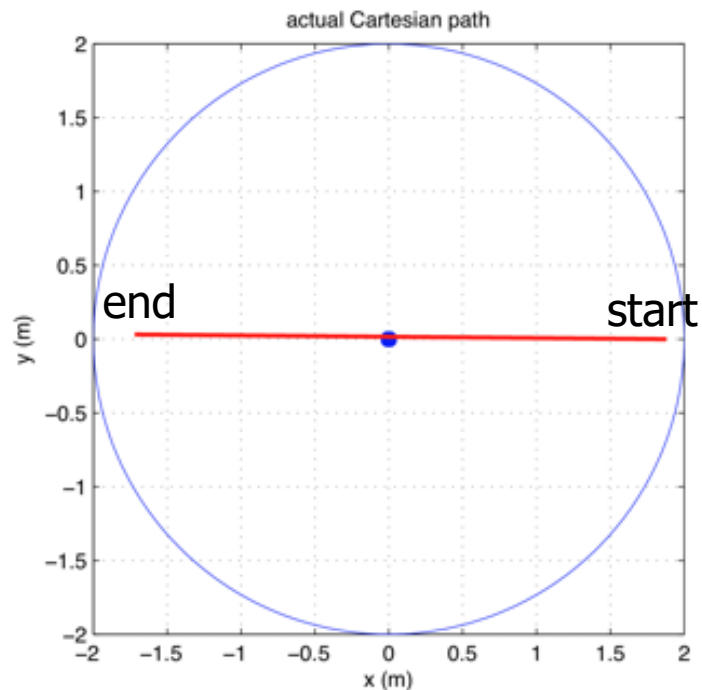


Simulation results

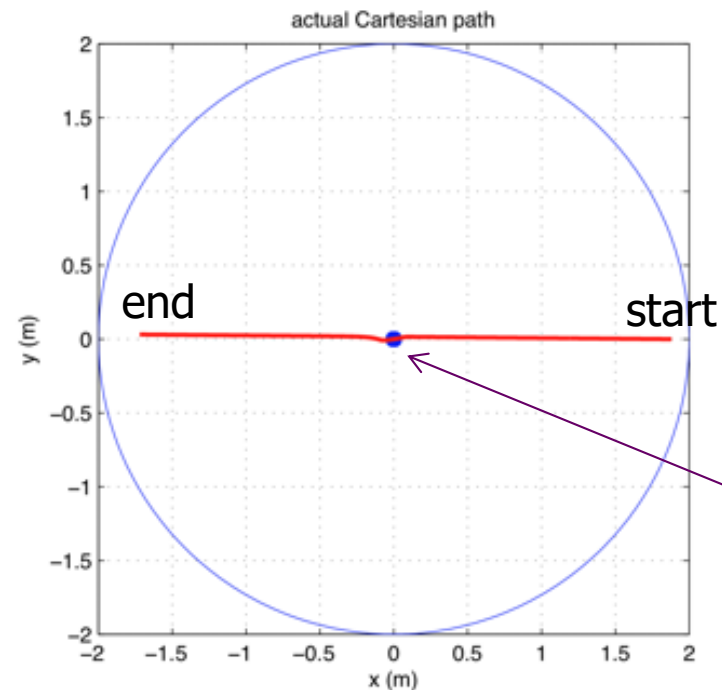
planar 2R robot in straight line Cartesian motion

a comparison of inverse and damped inverse Jacobian methods
even closer to singular case (removing joint velocity saturation)

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



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



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



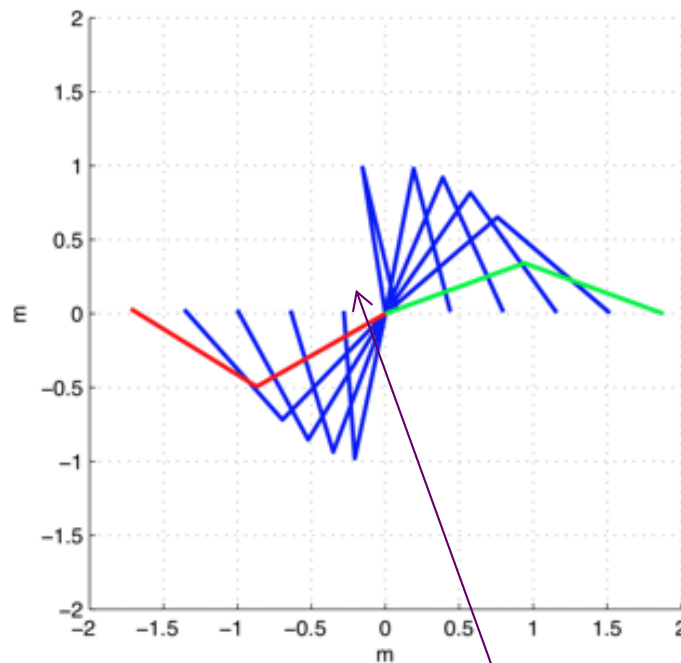
Simulation results

planar 2R robot in straight line Cartesian motion

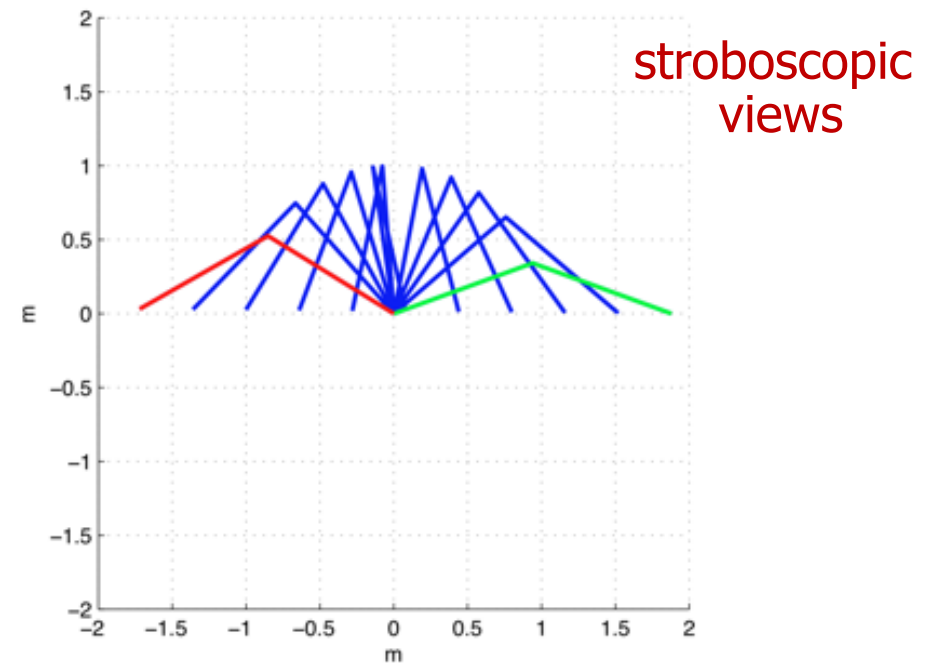
$$\dot{q} = J^{-1}(q)v$$

path at
 $\alpha = 179.5^\circ$

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



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



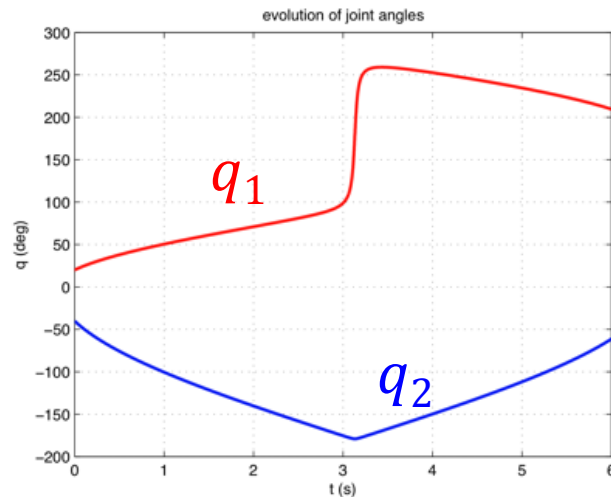
a completely **different inverse solution**, around/after crossing the region close to the folded singularity



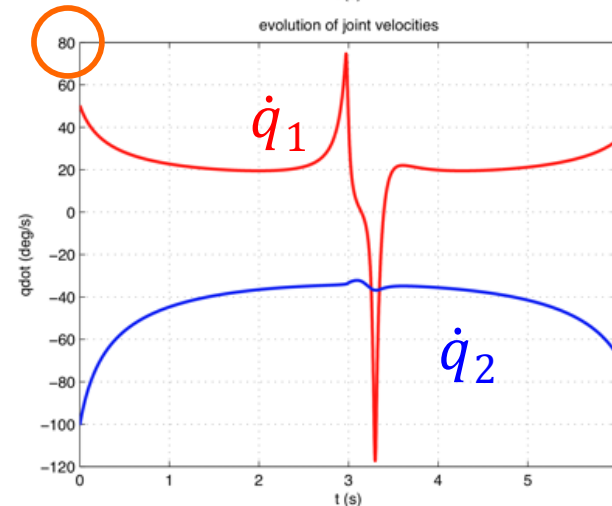
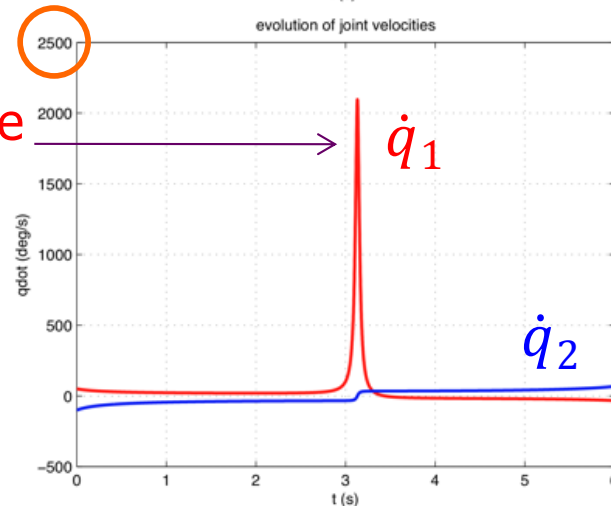
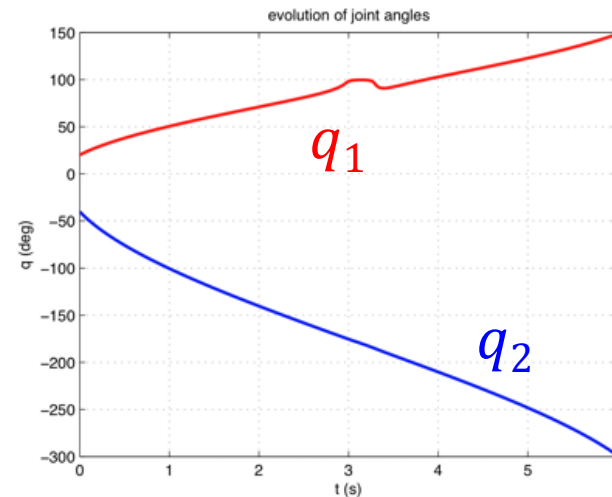
Simulation results

planar 2R robot in straight line Cartesian motion

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



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



smoother joint motion with limited joint velocities!

extremely large peak velocity of first joint!!

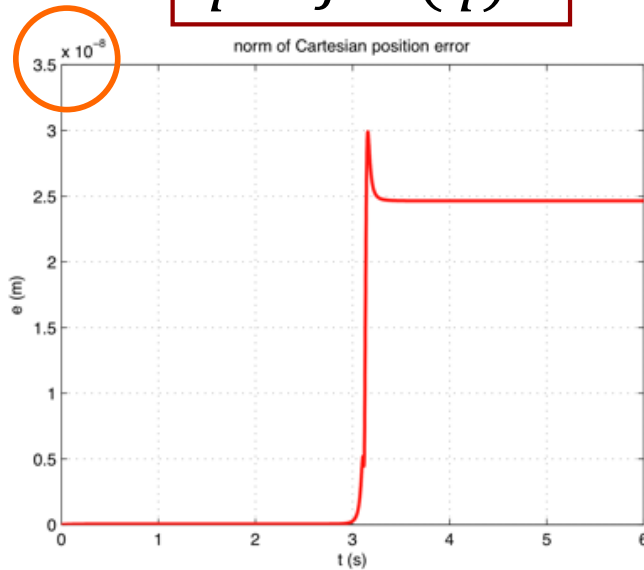


Simulation results

planar 2R robot in straight line Cartesian motion

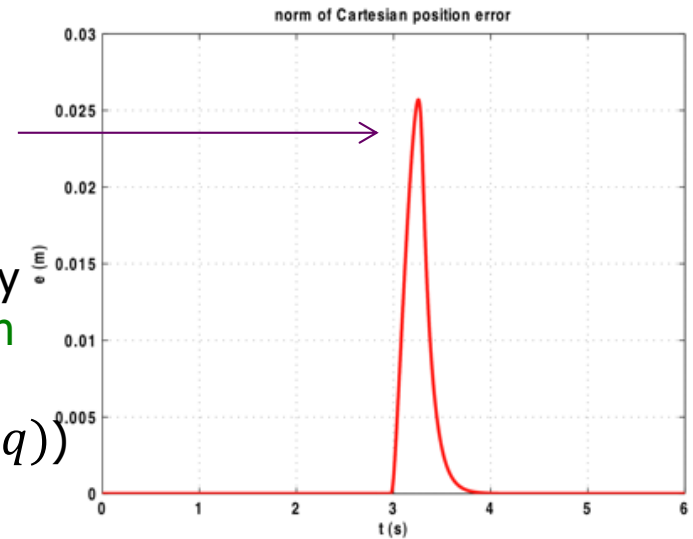
$$\dot{q} = J^{-1}(q)v$$

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



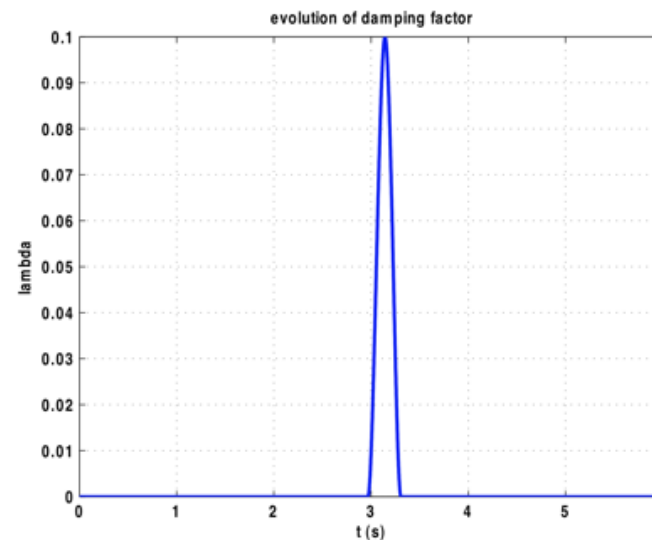
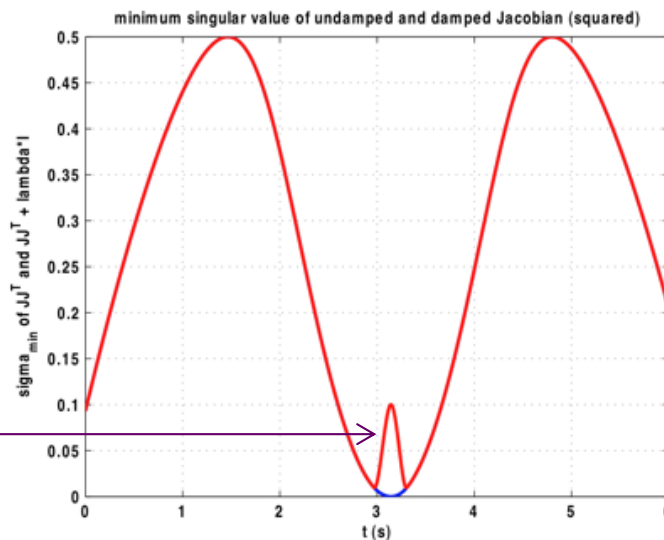
increased numerical integration error ($3 \cdot 10^{-8}$)

error (25 mm) when crossing the singularity, later recovered by a **feedback action** ($v \Rightarrow v + K_p e_p$ with $e_p = p_d - p(q)$)



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

they differ only when damping factor is non-zero



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



Pseudoinverse method

a constrained optimization (minimum norm) problem

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

$$S = \left\{ \begin{array}{l} \dot{q} \in R^n : \\ \|J\dot{q} - v\| \text{ is minimum} \end{array} \right\}$$

solution

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

pseudoinverse of J

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

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



Definition of the pseudoinverse

given J , is the **unique** matrix $J^\#$ satisfying the **four** relationships

$$J J^\# J = J$$

$$J^\# J J^\# = J^\#$$

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

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

- explicit expressions for **full rank** cases
 - if $\rho(J) = m = n$: $J^\# = J^{-1}$
 - if $\rho(J) = m < n$: $J^\# = J^T (J J^T)^{-1}$
 - if $\rho(J) = n < m$: $J^\# = (J^T J)^{-1} J^T$
- $J^\#$ **always** exists and is computed in general numerically using the SVD = Singular Value Decomposition of J
 - e.g., with the **MATLAB** function **pinv** (which uses in turn **svd**)



Numerical example

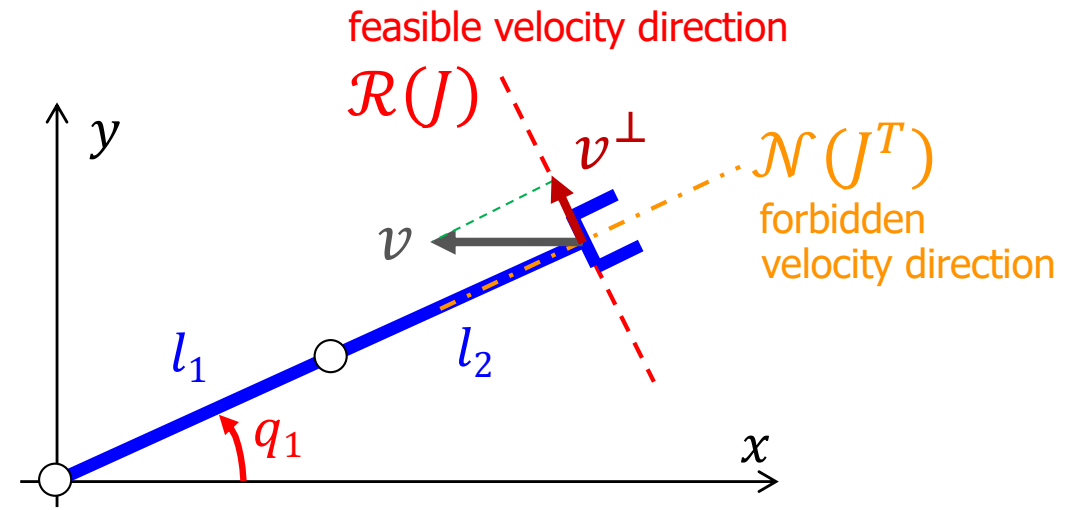
Jacobian of 2R robot with $l_1 = l_2 = 1$ at $q_2 = 0$ (rank $\rho(J) = 1$)

$$J = \begin{pmatrix} -2s_1 & -s_1 \\ 2c_1 & c_1 \end{pmatrix}$$

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

$$JJ^\# = \begin{pmatrix} s_1^2 & -s_1c_1 \\ -s_1c_1 & c_1^2 \end{pmatrix} \quad J^\#J = \begin{pmatrix} 0.8 & 0.4 \\ 0.4 & 0.2 \end{pmatrix}$$

both symmetric ...



$\dot{q} = J^\# v$ is the **minimum** norm joint velocity vector that **realizes exactly** v^\perp

- at $q_1 = \pi/6$: for $v = \begin{pmatrix} -0.5 \\ 0 \end{pmatrix}$ [m/s], $\dot{q} = J^\# v = \begin{pmatrix} 0.1 \\ 0.05 \end{pmatrix}$ [rad/s] $\Rightarrow v^\perp = JJ^\# v = \begin{pmatrix} -1/8 \\ \sqrt{3}/8 \end{pmatrix}$ [m/s]
- at $q_1 = \pi/2$: $J = \begin{pmatrix} -2 & -1 \\ 0 & 0 \end{pmatrix} \Rightarrow J^\# = \begin{pmatrix} -0.4 & 0 \\ -0.2 & 0 \end{pmatrix}$; now the **same** $v \in \mathcal{R}(J)$, $\dot{q} = \begin{pmatrix} 0.2 \\ 0.1 \end{pmatrix} \Rightarrow v^\perp = v$ (no error!)



General solution for $m < n$

ALL solutions of the inverse differential kinematics problem can be written as

$$\dot{q} = J^\# v + (I - J^\# J) \xi \leftarrow \text{any joint velocity...}$$

projection matrix in the null space $\mathcal{N}(J)$

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

$$\min_{\dot{q}} H = \frac{1}{2} \|\dot{q} - \xi\|^2 \text{ such that } J\dot{q} = v \iff \min_{\dot{q} \in S} H = \frac{1}{2} \|\dot{q} - \xi\|^2$$
$$S = \left\{ \begin{array}{l} \dot{q} \in \mathbb{R}^n : \\ \|J\dot{q} - v\| \text{ is minimum} \end{array} \right\}$$

verification of the **actual** task velocity that is being obtained

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

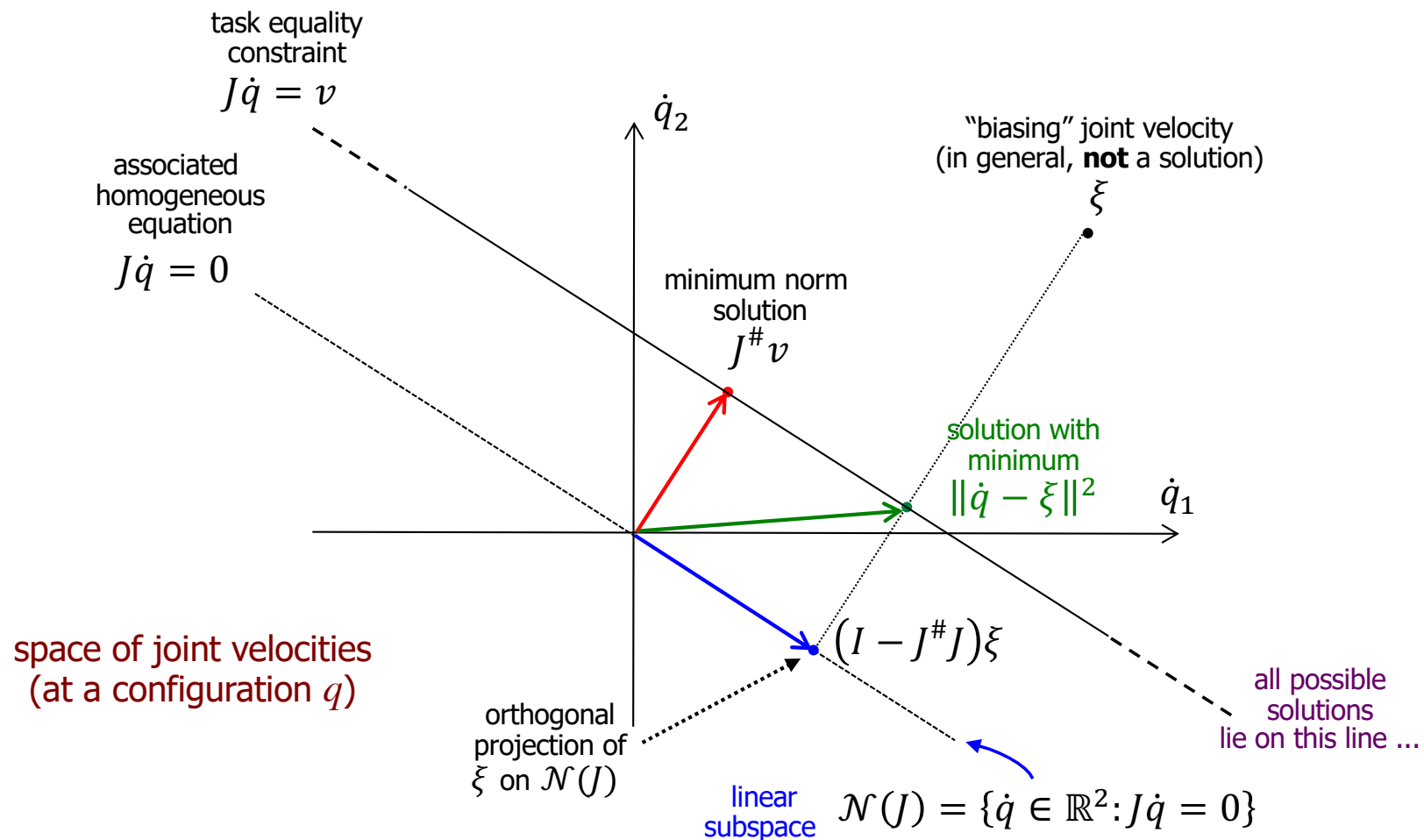
if $v \in \mathcal{R}(J) \Rightarrow v = Jw$ for some $w \in \mathbb{R}^n$



Geometric interpretation for $m < n$

a simple case with $n = 2, m = 1$
at a given configuration

$$J\dot{q} = [j_1 \quad j_2] \begin{bmatrix} \dot{q}_1 \\ \dot{q}_2 \end{bmatrix} = v \in \mathbb{R}$$

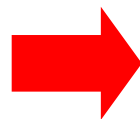




Velocity manipulability

- in a given configuration, evaluate how effective is the **transformation** between joint and end-effector velocities
 - “how easily” can the end-effector be moved in 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**

$$\dot{q}^T \dot{q} = 1$$



$$v^T J^{\#T} J^{\#} v = 1$$

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

if $\rho(J) = m$

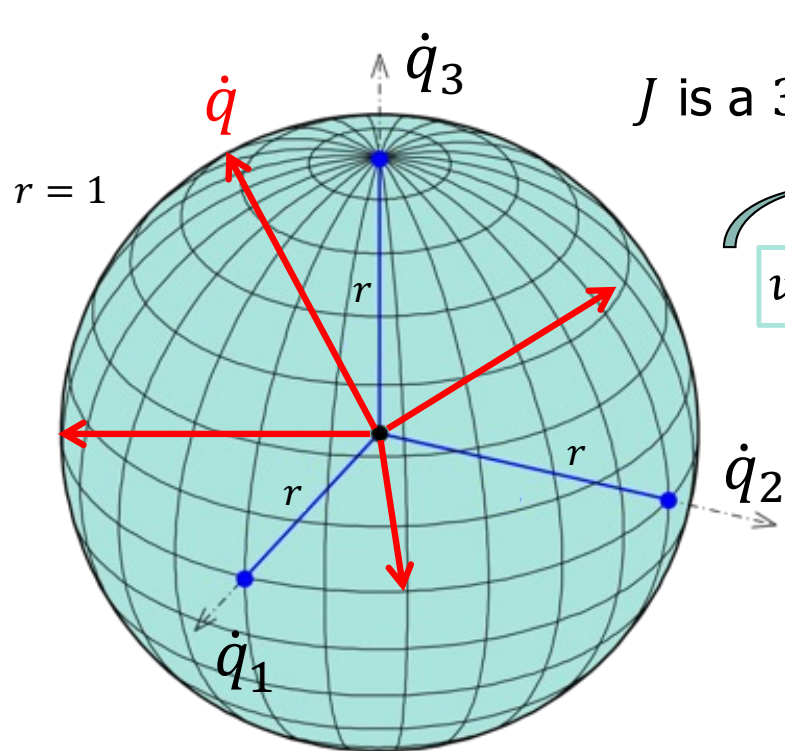
$$(J J^T)^{-1}$$

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



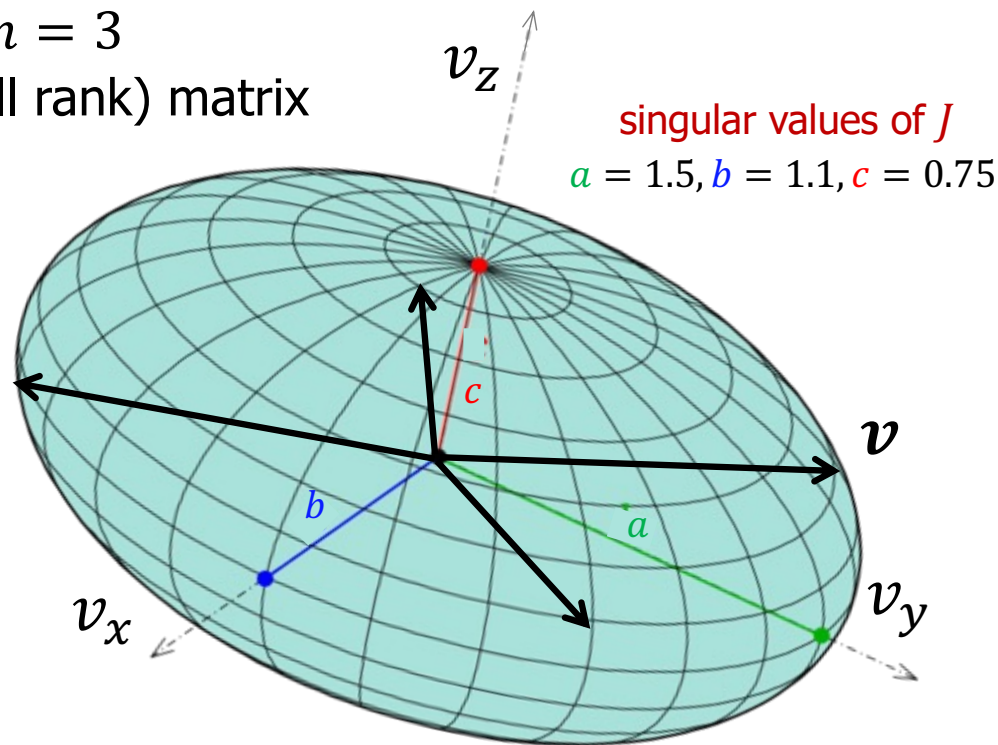
(Hyper-)Spheres and Ellipsoids

whiteboard ...



$m = n = 3$
 J is a 3×3 (full rank) matrix

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



singular values of J
 $a = 1.5, b = 1.1, c = 0.75$

$$\dot{q}_1^2 + \dot{q}_2^2 + \dot{q}_3^2 = \dot{q}^T \begin{pmatrix} 1 & & \\ & 1 & \\ & & 1 \end{pmatrix} \dot{q} = 1$$

$$\frac{v_x^2}{a^2} + \frac{v_y^2}{b^2} + \frac{v_z^2}{c^2} = v^T \begin{pmatrix} a^2 & & \\ & b^2 & \\ & & c^2 \end{pmatrix}^{-1} v = 1$$

$$\dot{q}^T \dot{q} = 1$$

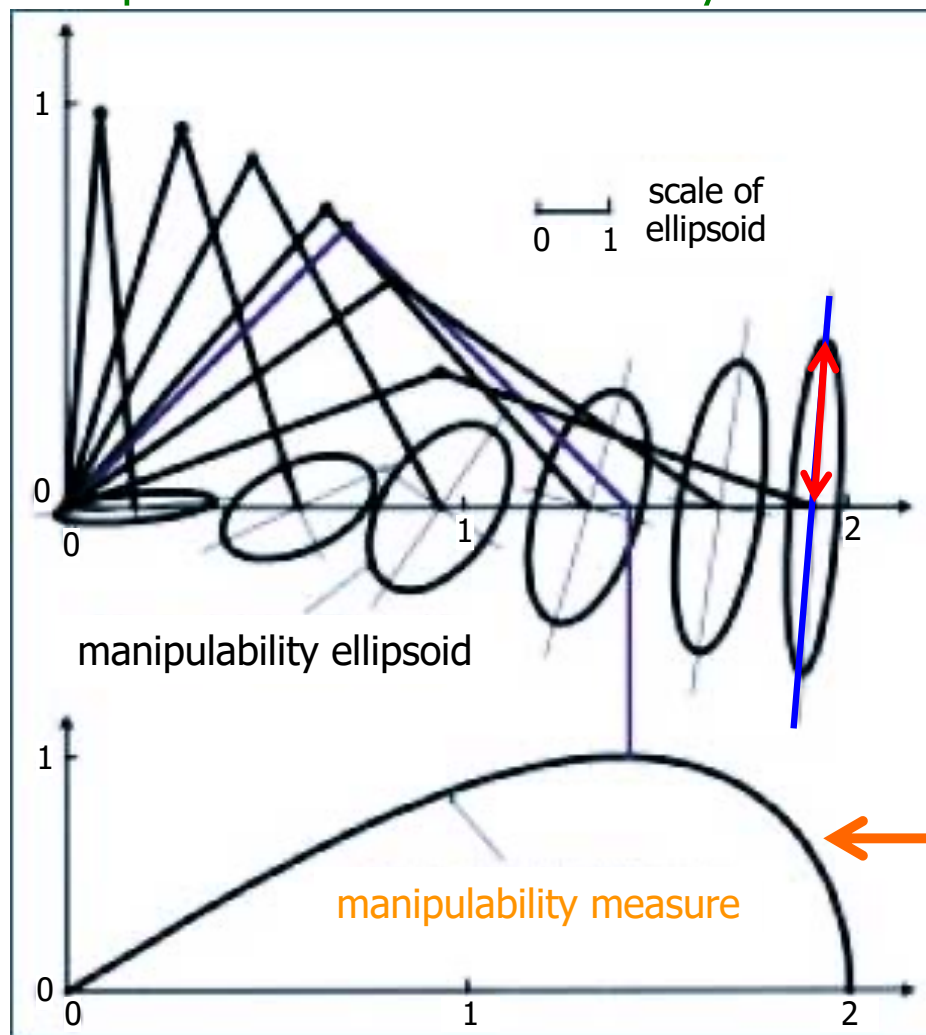


$$v^T (J J^T)^{-1} v = 1$$



Manipulability ellipsoid in velocity

planar 2R arm with unitary links



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

$$\sigma_i(J) = \sqrt{\lambda_i(J J^T)}$$

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

direction of principal axes
eigenvectors associated to λ_i

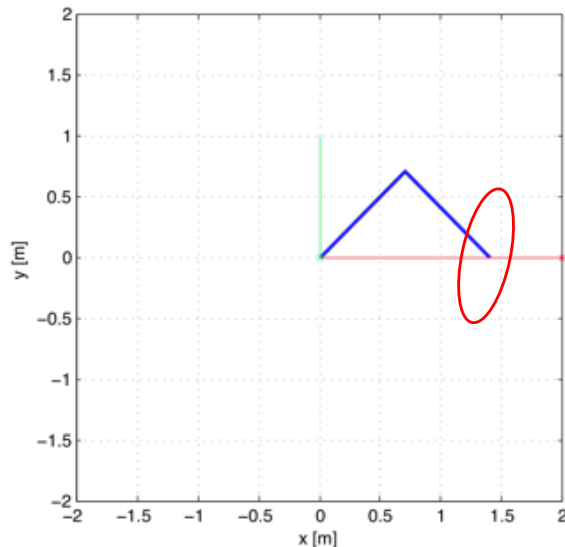
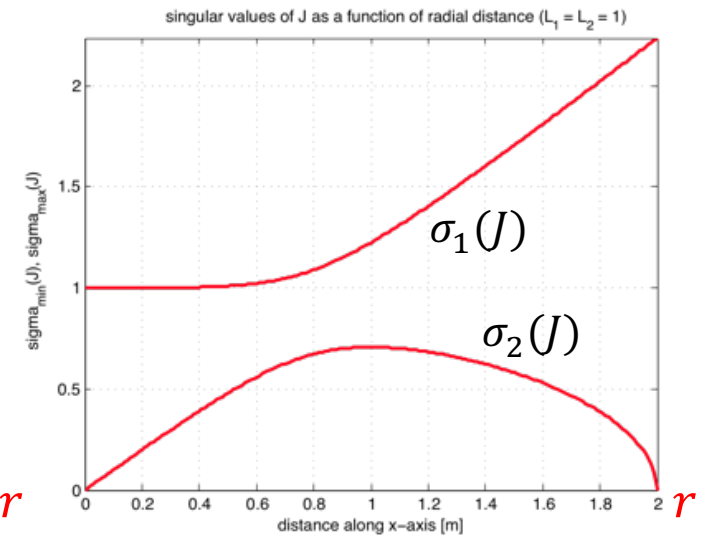
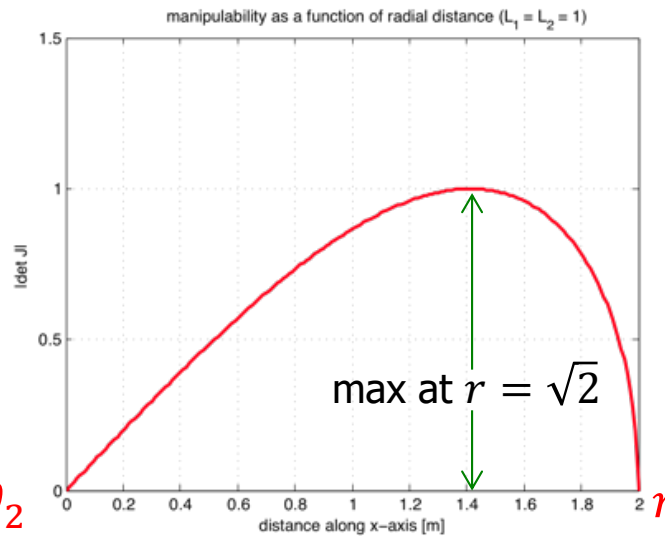
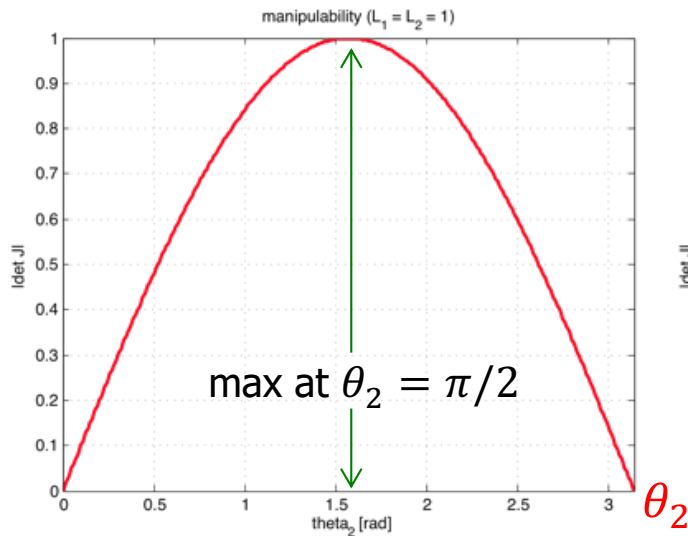
$$w = \sqrt{\det(J J^T)} = \prod_{i=1}^m \sigma_i \geq 0$$

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



Manipulability measure

planar 2R arm (with $l_1 = l_2 = 1$): $\sqrt{\det(J J^T)} = \sqrt{\det(J) \cdot \det(J^T)} = |\det J| = \prod_{i=1}^2 \sigma_i$



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

no full isotropy (i.e., a circle)
is obtained in this case
since it is always $\sigma_1 \neq \sigma_2$





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}

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

- (pseudo-)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 $\ddot{r}, \dddot{r}, \dots$)