



## ***Robotics 1***

# **Inverse kinematics**

Prof. Alessandro De Luca

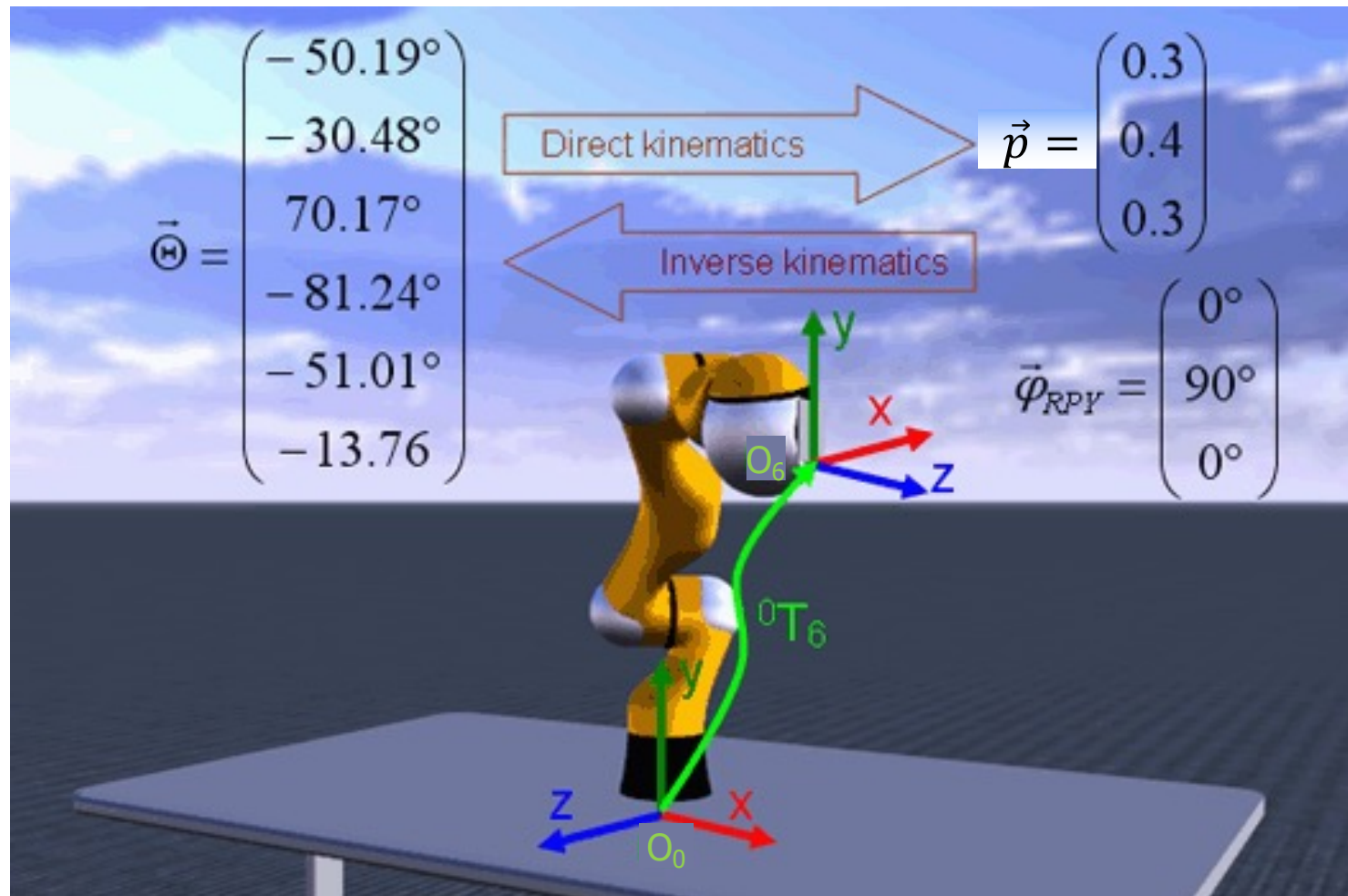
DIPARTIMENTO DI INGEGNERIA INFORMATICA  
AUTOMATICA E GESTIONALE ANTONIO RUBERTI



**SAPIENZA**  
UNIVERSITÀ DI ROMA

# Inverse kinematics

## what are we looking for?



direct kinematics is always unique;  
how about inverse kinematics for this 6R robot?



# Inverse kinematics problem

- given a desired end-effector pose (position + orientation), **find** the values of the joint variables  $q$  that will realize it
- a **synthesis** problem, with **input** data in the form

$$\blacksquare T = \begin{bmatrix} R & p \\ 0^T & 1 \end{bmatrix} = {}^0A_n(q) \quad \blacksquare r = f_r(q), \text{ for a task function}$$

classical formulation:

inverse kinematics for a given end-effector pose  $T$

generalized formulation:

inverse kinematics for a given value  $r$  of task variables

- a typical **nonlinear** problem
  - **existence** of a solution (**workspace** definition)
  - uniqueness/**multiplicity** of solutions ( $r \in \mathbb{R}^m, q \in \mathbb{R}^n$ )
  - solution **methods**



# Solvability and robot workspace

for tasks related to a desired end-effector Cartesian pose

- **primary workspace  $WS_1$** : set of all positions  $p$  that can be reached with **at least one** orientation ( $\phi$  or  $R$ )
  - out of  $WS_1$  there is no solution to the problem
  - if  $p \in WS_1$ , there is a suitable  $\phi$  (or  $R$ ) for which a solution **exists**
- **secondary (or dexterous) workspace  $WS_2$** : set of positions  $p$  that can be reached with **any** orientation (among those **feasible** for the robot direct kinematics)
  - if  $p \in WS_2$ , there exists a solution for **any** feasible  $\phi$  (or  $R$ )
- $WS_2 \subseteq WS_1$





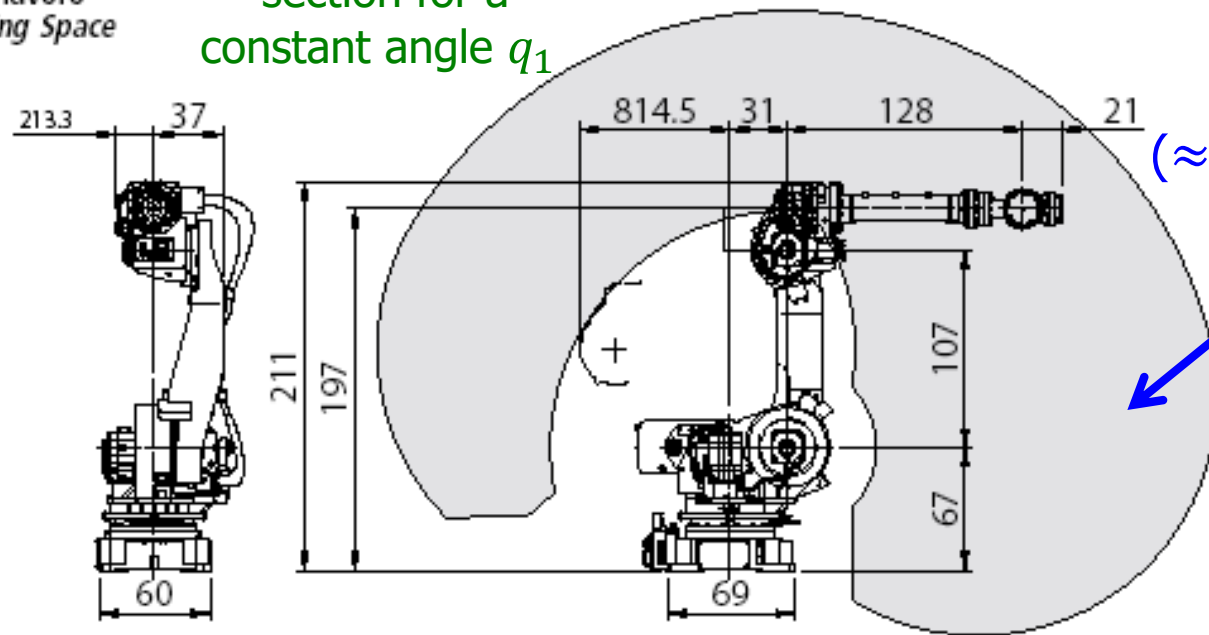
# Workspace of Fanuc R-2000i/165F

Area di lavoro  
Operating Space

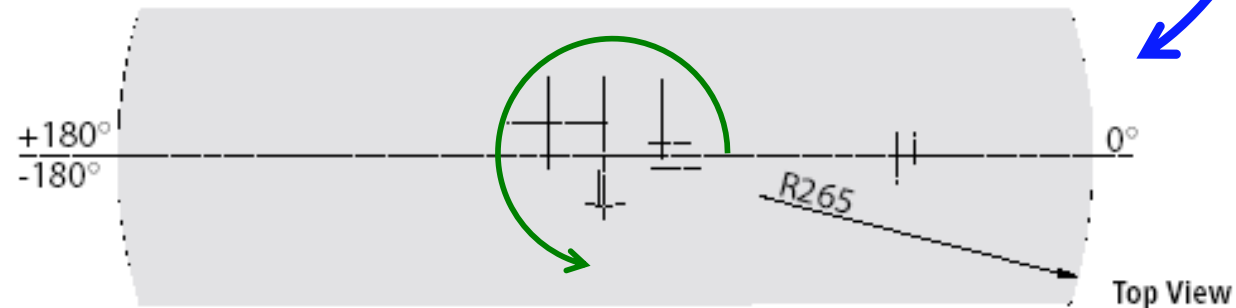
section for a  
constant angle  $q_1$

$$WS_1 \subset \mathbb{R}^3$$

( $\approx WS_2$  for spherical wrist  
without joint limits)



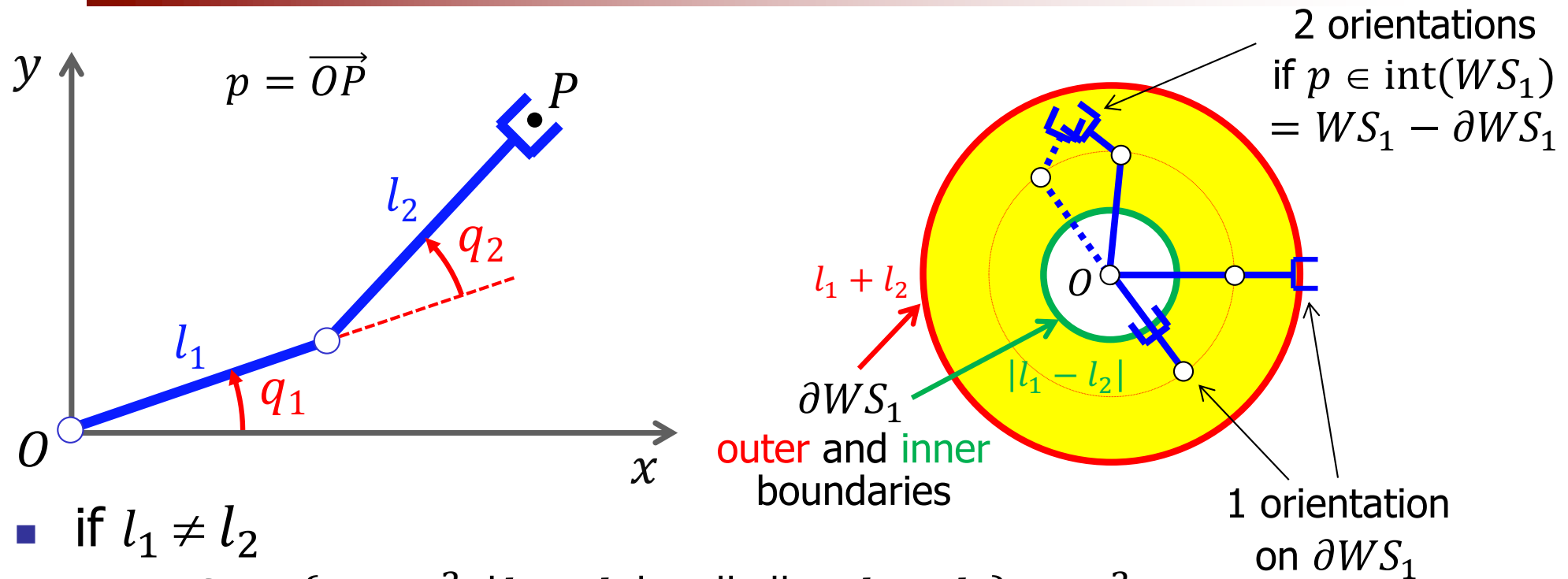
Side View



Top View

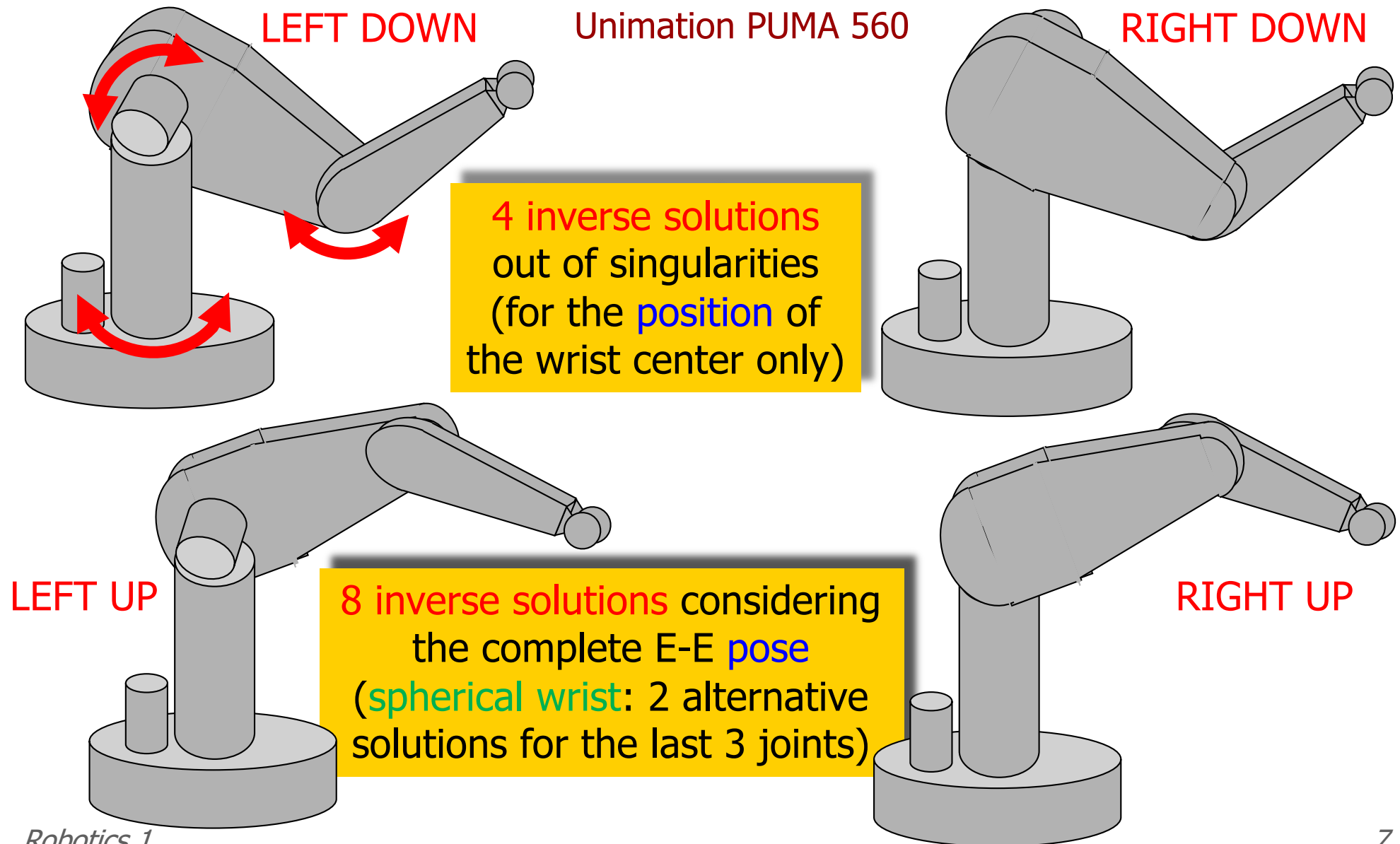
rotating the  
base joint angle  $q_1$

# Workspace of a planar 2R arm



- if  $l_1 \neq l_2$ 
  - $WS_1 = \{p \in \mathbb{R}^2 : |l_1 - l_2| \leq \|p\| \leq l_1 + l_2\} \subset \mathbb{R}^2$
  - $WS_2 = \emptyset$
- if  $l_1 = l_2 = l$ 
  - $WS_1 = \{p \in \mathbb{R}^2 : \|p\| \leq 2l\} \subset \mathbb{R}^2$
  - $WS_2 = \{p = 0\}$  (all **feasible** orientations at the origin!... an **infinite** number)

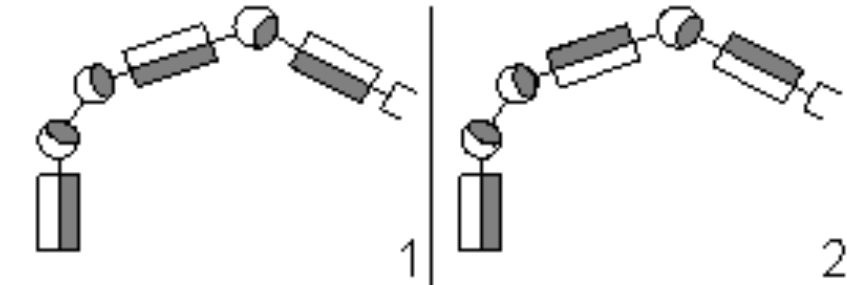
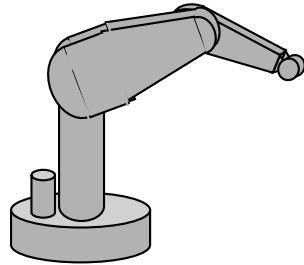
# Wrist position and E-E pose inverse solutions for an articulated 6R robot



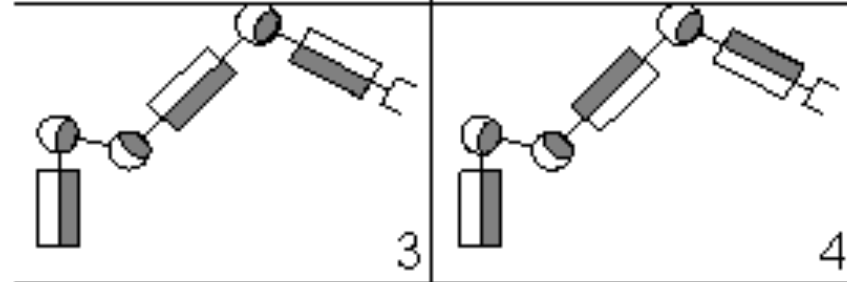
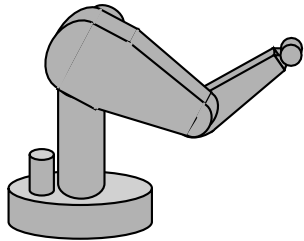
# Counting/visualizing the 8 solutions of the inverse kinematics for a Unimation Puma 560



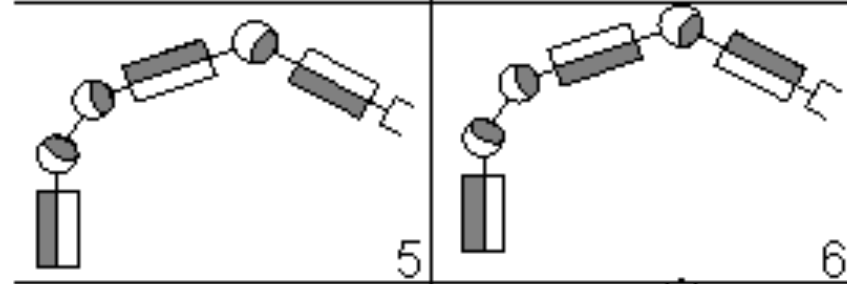
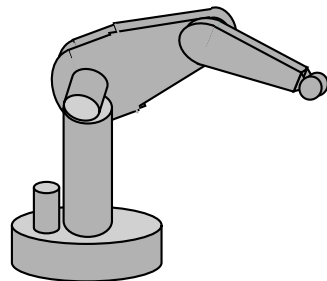
RIGHT UP



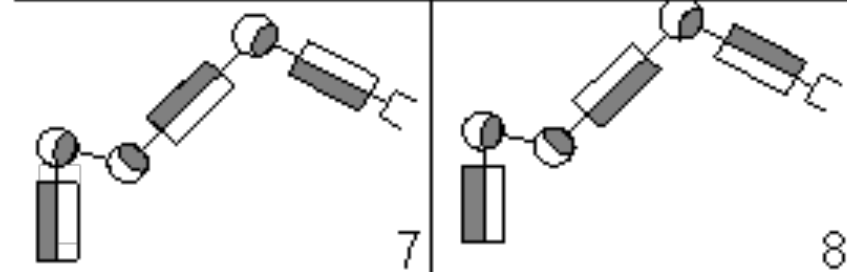
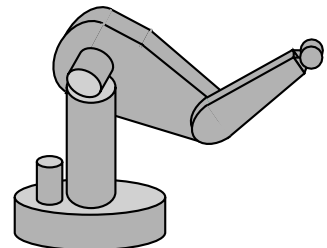
RIGHT DOWN



LEFT UP



LEFT DOWN



# Inverse kinematic solutions of UR10

## 6-dof Universal Robot UR10, with non-spherical wrist



video (slow motion)

desired pose

$$p = \begin{pmatrix} -0.2373 \\ -0.0832 \\ 1.3224 \end{pmatrix} [\text{m}]$$

$$R = \begin{pmatrix} \sqrt{3}/2 & 0.5 & 0 \\ -0.5 & \sqrt{3}/2 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

home configuration at start

$$q = (0 \quad -\pi/2 \quad 0 \quad -\pi/2 \quad 0 \quad 0)^T [\text{rad}]$$





# 8 inverse kinematic solutions of UR10



shoulderRight  
wristDown  
elbowUp

$$q = \begin{pmatrix} 1.0472 \\ -1.2833 \\ -0.7376 \\ -2.6915 \\ -1.5708 \\ 3.1416 \end{pmatrix}$$



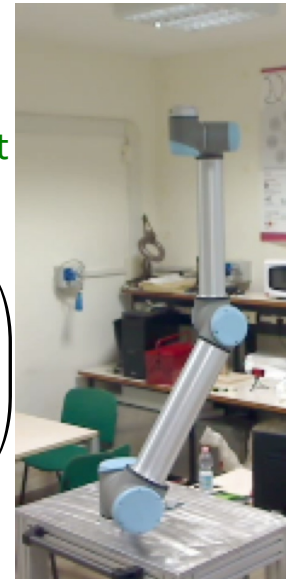
shoulderRight  
wristDown  
elbowDown

$$q = \begin{pmatrix} 1.0472 \\ -1.9941 \\ 0.7376 \\ 2.8273 \\ -1.5708 \\ 3.1416 \end{pmatrix}$$



shoulderRight  
wristUp  
elbowUp

$$q = \begin{pmatrix} 1.0472 \\ -1.5894 \\ -0.5236 \\ 0.5422 \\ 1.5708 \\ 0 \end{pmatrix}$$



shoulderRight  
wristUp  
elbowDown

$$q = \begin{pmatrix} 1.0472 \\ -2.0944 \\ 0.5236 \\ 0 \\ 1.5708 \\ 0 \end{pmatrix}$$



shoulderLeft  
wristDown  
elbowDown

$$q = \begin{pmatrix} 2.7686 \\ -1.0472 \\ -0.5236 \\ 3.1416 \\ -1.5708 \\ 1.4202 \end{pmatrix}$$



shoulderLeft  
wristDown  
elbowUp

$$q = \begin{pmatrix} 2.7686 \\ -1.5522 \\ 0.5236 \\ 2.5994 \\ -1.5708 \\ 1.4202 \end{pmatrix}$$



shoulderLeft  
wristUp  
elbowDown

$$q = \begin{pmatrix} 2.7686 \\ -1.1475 \\ -0.7376 \\ 0.3143 \\ 1.5708 \\ -1.7214 \end{pmatrix}$$



shoulderLeft  
wristUp  
elbowUp

$$q = \begin{pmatrix} 2.7686 \\ -1.8583 \\ 0.7376 \\ -0.4501 \\ 1.5708 \\ -1.7214 \end{pmatrix}$$



# Multiplicity of solutions

few examples

- E-E positioning of planar 2R robot ( $m = n = 2$ )
  - 2 **regular** solutions in  $\text{int}(WS_1)$
  - 1 solution on  $\partial WS_1$
  - for  $l_1 = l_2$ :  $\infty$  solutions in  $WS_2$

} **singular** solutions
- E-E positioning of elbow-type spatial 3R robot ( $m = n = 3$ )
  - 4 **regular** solutions in  $WS_1$  (with **singular** cases yet to be investigated ...)
- spatial 6R robot arms ( $m = n = 6$ )
  - $\leq 16$  **distinct solutions**, out of singularities: this “upper bound” of solutions was shown to be attained by a particular instance of “orthogonal” robot, i.e., with twist angles  $\alpha_i = 0$  or  $\pm\pi/2$  ( $\forall i$ )
  - analysis based on **algebraic transformations** of robot kinematics
    - transcendental equations are transformed into a single polynomial equation in one variable (number of roots = degree of the polynomial)
    - seek for a transformed polynomial equation of the least possible degree



# Algebraic transformations

whiteboard ...



start with some **trigonometric equation** in the joint angle  $\theta$  to be solved ...

$$a \sin \theta + b \cos \theta = c \quad (*)$$

introduce the algebraic transformation (... and the related inverse formulas)

$$u = \tan(\theta/2)$$

$$\Rightarrow \sin \theta = \frac{2u}{1+u^2} \quad \cos \theta = \frac{1-u^2}{1+u^2} \quad (\Rightarrow \sin^2 \theta + \cos^2 \theta = 1)$$

$$\tan \theta = \tan 2(\theta/2) = \frac{2 \tan(\theta/2)}{1 - \tan^2(\theta/2)} = \frac{2u}{1-u^2} \quad (\text{using the duplication formula})$$

substituting in **(\*)**

$$a \frac{2u}{1+u^2} + b \frac{1-u^2}{1+u^2} = c \quad \Rightarrow \quad \text{polynomial equation of second degree in } u$$
$$(b+c) u^2 - 2a u - (b-c) = 0$$

$$\Rightarrow u_{1,2} = \frac{a \pm \sqrt{a^2 + b^2 - c^2}}{b+c} \quad \Rightarrow \quad \theta_{1,2} = 2 \arctan(u_{1,2})$$

**only if argument is real**, else no solution



# A 6R robot with 16 IK solutions

all distinct and non-singular

an **orthogonal** manipulator with DH table

$i$	$d_i$	$\theta_i$	$a_i$	$\alpha_i$
1	0	$\theta_1$	$a_1$	$\pi/2$
2	0	$\theta_2$	$a_2$	0
3	$d_3$	$\theta_3$	0	$\pi/2$
4	0	$\theta_4$	$a_4$	0
5	0	$\theta_5$	0	$\pi/2$
6	0	$\theta_6$	0	0

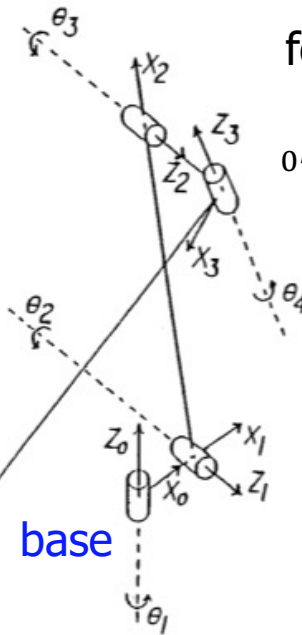
$$a_1 = 0.3, a_2 = 1, a_4 = 1.5, d_3 = 0.2$$

with non-spherical wrist

end-effector

Manseur and Doty:  
International Journal of Robotics Research, 1989

solutions found using a fast  
numerical inversion algorithm ...



for the desired end-effector pose

$${}^0T_6 = \begin{bmatrix} -0.760117 & -0.641689 & 0.102262 & -1.140175 \\ 0.133333 & 0 & 0.991071 & 0 \\ -0.635959 & 0.766965 & 0.085558 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$



there are **16 real solutions**  
of the inverse kinematics!

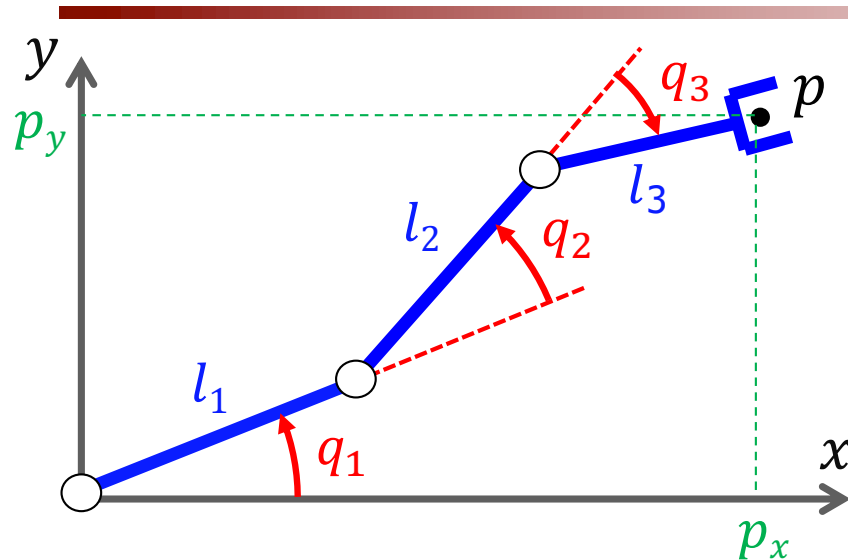
all **non-singular**



$n$	$\theta_1$	$\theta_2$	$\theta_3$	$\theta_4$	$\theta_5$	$\theta_6$	$\det(J^3)$
1	0.000	107.458	112.460	-7.662	0.000	0.000	1.310
2	0.000	107.458	-67.540	-172.338	180.000	180.000	1.310
3	88.670	-176.682	-178.394	-63.284	157.829	139.944	-0.800
4	88.670	-176.682	1.606	-116.716	22.171	-40.056	-0.800
5	113.841	4.741	-179.093	-55.954	-63.659	-42.463	-1.256
6	113.841	4.741	0.907	-124.046	-116.341	137.537	-1.256
7	168.703	-104.205	146.556	-16.393	-170.903	98.216	0.803
8	168.703	-104.205	-33.444	-163.607	-9.097	-81.784	0.803
9	180.000	107.458	-147.375	-7.662	-164.675	180.000	0.732
10	180.000	107.458	32.625	-172.338	-15.325	0.000	0.732
11	-120.748	173.066	-178.472	31.328	-146.087	142.605	-0.717
12	-120.748	173.066	1.528	148.672	-33.913	-37.395	-0.717
13	-96.292	-5.766	-179.142	38.477	51.922	-39.631	-1.441
14	-96.292	-5.766	0.858	141.523	128.078	140.369	-1.441
15	-11.768	-105.495	-114.490	1.243	6.408	-79.398	1.318
16	-11.768	-105.495	65.510	178.757	173.592	100.602	1.318

# A planar 3R arm

workspace and number/type of inverse solutions



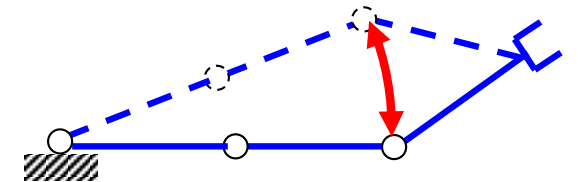
$$l_1 = l_2 = l_3 = l \quad n = 3, m = 2$$

$$WS_1 = \{p \in \mathbb{R}^2 : \|p\| \leq 3l\} \subset \mathbb{R}^2$$

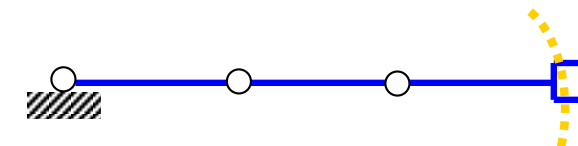
$$WS_2 = \{p \in \mathbb{R}^2 : \|p\| \leq l\} \subset \mathbb{R}^2$$

any planar orientation is feasible in  $WS_2$

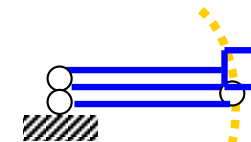
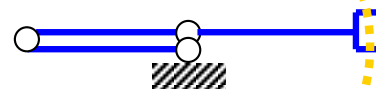
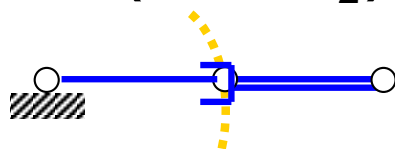
1. in  $\text{int}(WS_1) - \partial WS_2$ :  $\infty^1$  **regular** solutions at which the E-E can take a **continuum** of  $\infty$  orientations (but **not all** orientations in the plane!)



2. if  $\|p\| = 3l$  (at  $\partial WS_1$ ): only 1 solution, **singular**



3. if  $\|p\| = l$  (at  $\partial WS_2$ ):  $\infty^1$  solutions, 3 of which **singular**



4. if  $\|p\| < l$  (in  $\text{int}(WS_2)$ ):  $\infty^1$  **regular** solutions (that are **never singular**)

# Workspace of a planar 3R arm with generic link lengths

$$l_{max} = \max \{l_i, i = 1, 2, 3\}$$

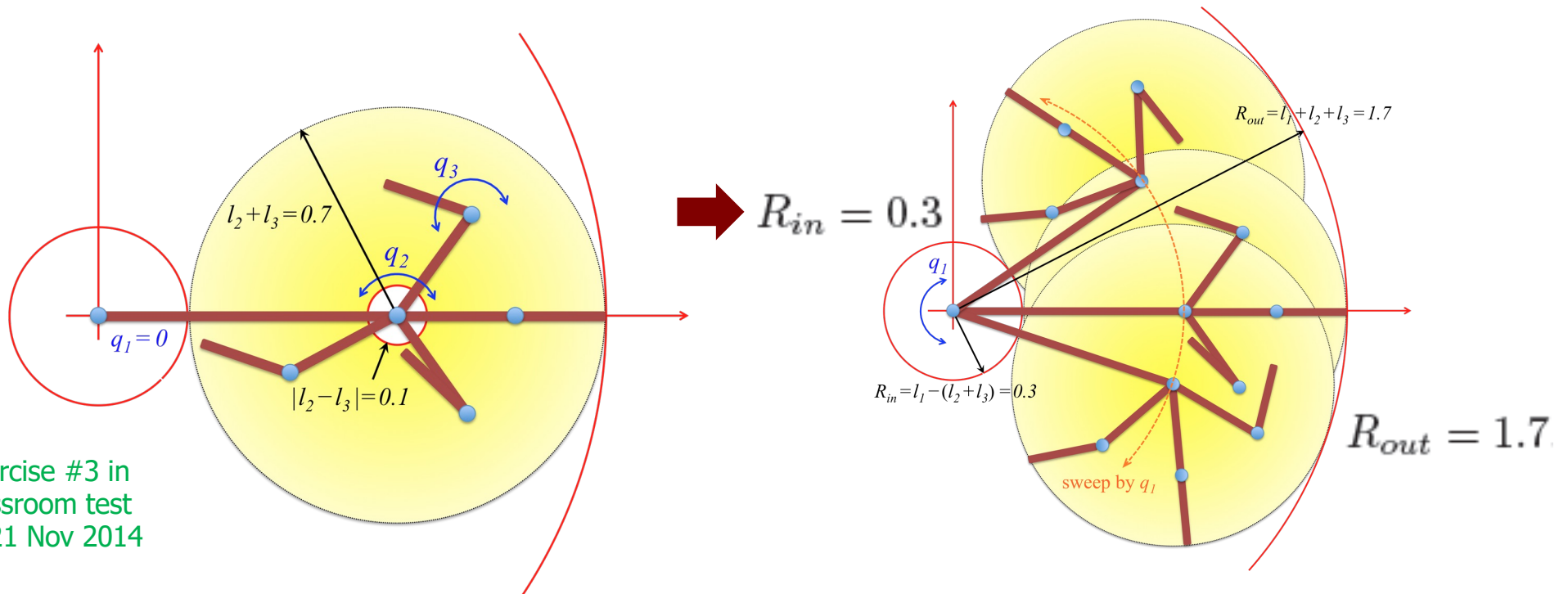
$$l_{min} = \min \{l_i, i = 1, 2, 3\}$$



$$R_{out} = l_{min} + l_{med} + l_{max} = l_1 + l_2 + l_3$$

$$R_{in} = \max \{0, l_{max} - (l_{med} + l_{min})\}$$

a)  $l_1 = 1, l_2 = 0.4, l_3 = 0.3$  [m]  $\Rightarrow l_{max} = l_1 = 1, l_{med} = l_2 = 0.4, l_{min} = l_3 = 0.3$



b)  $l_1 = 0.5, l_2 = 0.7, l_3 = 0.5$  [m]  $\Rightarrow l_{max} = l_2 = 0.7, l_{med} = l_{min} = l_1(\text{or } l_3) = 0.5$



$$R_{in} = 0, R_{out} = 1.7$$



# Multiplicity of solutions

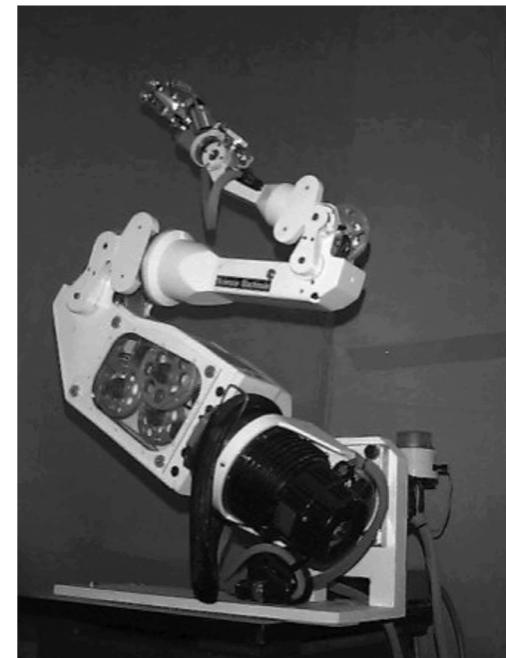
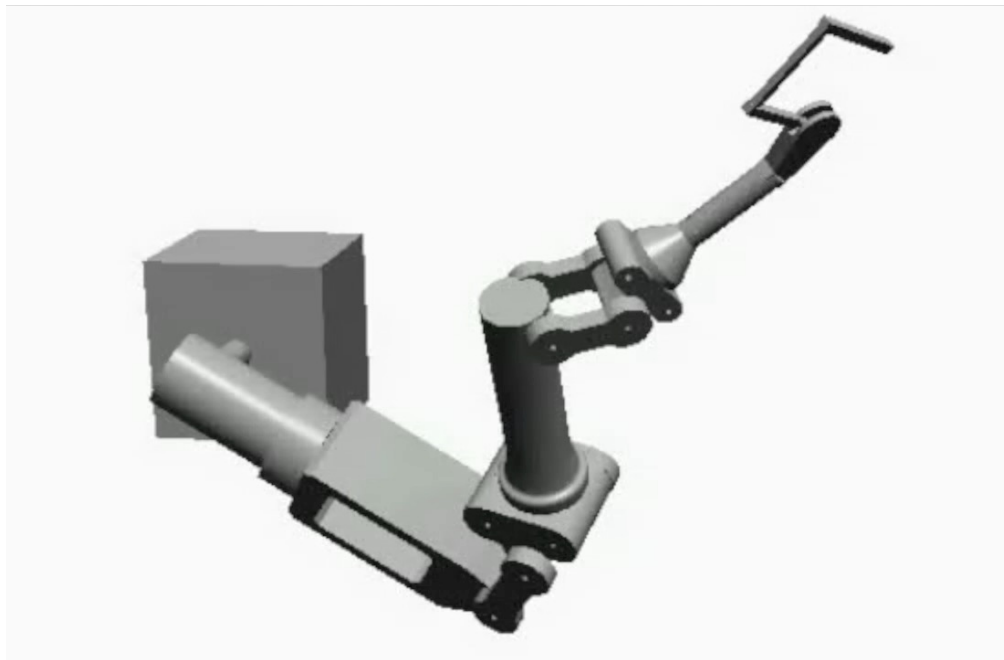
summary of the general cases

- if  $m = n$ 
  - $\nexists$  solutions
  - a finite number of solutions (regular/generic case)
  - “degenerate” solutions: infinite or finite set, but anyway different in number from the generic case (singularity)
- if  $m < n$  (robot is kinematically redundant for the task)
  - $\nexists$  solutions
  - $\infty^{n-m}$  solutions (regular/generic case)
  - a finite or infinite number of singular solutions
- use of the term singularity will become clearer when dealing with differential kinematics
  - instantaneous velocity mapping from joint to task velocity
  - lack of full rank of the associated  $m \times n$  Jacobian matrix  $J(q)$

# Dexter 8R robot arm

- $m = 6$  (position and orientation of E-E)
- $n = 8$  (all revolute joints)
- $\infty^2$  inverse kinematic solutions (redundancy degree =  $n - m = 2$ )

video

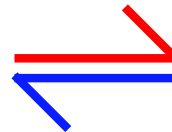


exploring inverse kinematic solutions by a robot self-motion



# Solution methods

## ANALYTICAL solution (in closed form)



## NUMERICAL solution (in iterative form)

- preferred, if it can be found\*
- use ad-hoc geometric inspection
- algebraic methods (solution of polynomial equations)
- systematic ways for generating a reduced set of equations to be solved

- it is certainly needed if  $n > m$  (redundant case) or at/close to singularities
- slower, but easier to be set up
- in its basic form, it uses the (analytical) **Jacobian matrix** of the direct kinematics map

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

- **Newton** method, **Gradient** method ...

### \* sufficient conditions for 6-dof arms

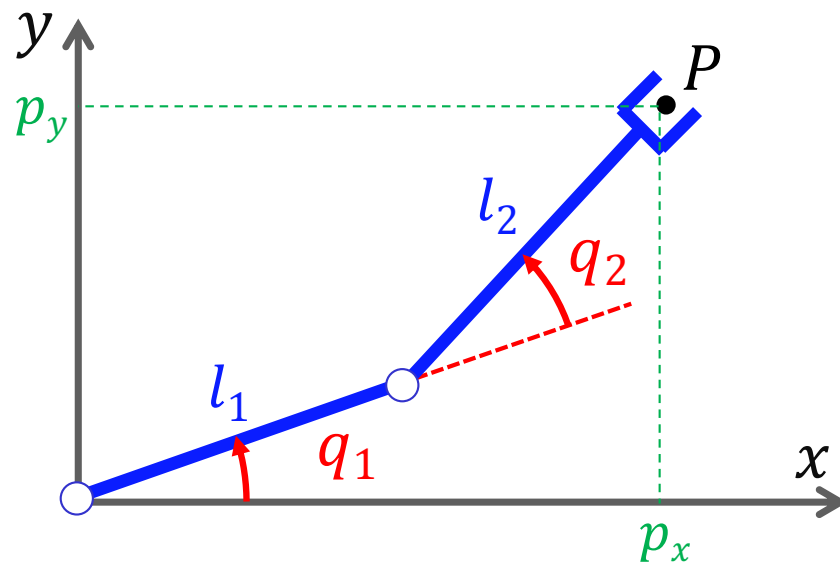
- 3 consecutive rotational joint axes are incident (e.g., spherical wrist), **or**
- 3 consecutive rotational joint axes are parallel

D. Pieper, PhD thesis, Stanford University, 1968





# Inverse kinematics of planar 2R arm



direct kinematics

$$p_x = l_1 c_1 + l_2 c_{12}$$

$$p_y = l_1 s_1 + l_2 s_{12}$$

$\underbrace{\quad}_{\text{data}}$   $q_1, q_2$  unknowns

“squaring and summing” the two equations of the direct kinematics

$$p_x^2 + p_y^2 - (l_1^2 + l_2^2) = 2l_1l_2(c_1c_{12} + s_1s_{12}) = 2l_1l_2c_2$$

and from this

$$c_2 = (p_x^2 + p_y^2 - (l_1^2 + l_2^2)) / 2l_1l_2, \quad s_2 = \pm \sqrt{1 - c_2^2}$$

$$q_2 = \text{atan2}\{s_2, c_2\}$$

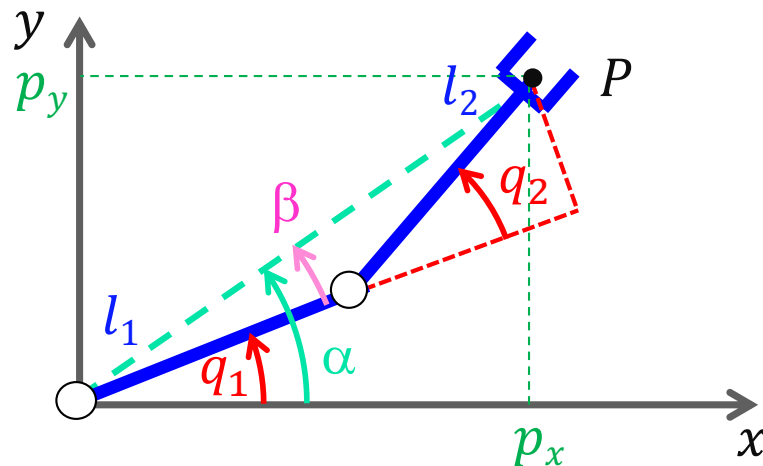
↑  
must be in  $[-1, 1]$  (else, point  $P$   
is outside robot workspace!)

↑  
2 solutions

↑  
in analytical form



# Inverse kinematics of 2R arm (cont'd)



by geometric inspection

$$q_1 = \alpha - \beta$$



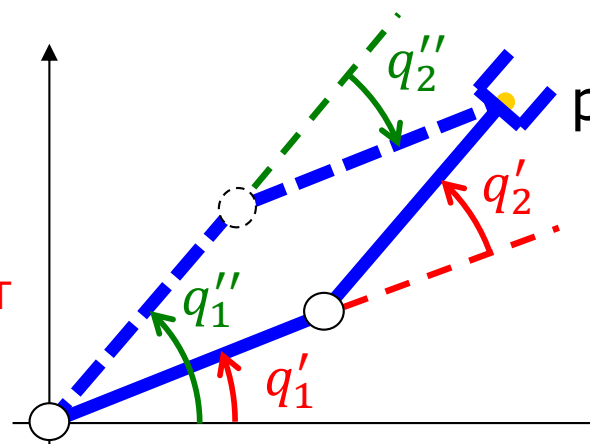
2 solutions  
(one for each value of  $s_2$ )

$$q_1 = \text{atan2}\{p_y, p_x\} - \text{atan2}\{l_2 s_2, l_1 + l_2 c_2\}$$

note: difference of atan2's needs to be re-expressed in  $(-\pi, \pi]$ !

$\{q_1, q_2\}_{\text{UP/LEFT}}$

$\{q_1, q_2\}_{\text{DOWN/RIGHT}}$



$q_2'$  and  $q_2''$  have same absolute value, but opposite signs

$q_1'$  and  $q_1''$  are in general unrelated to each other



# Algebraic solution for $q_1$

another  
solution  
method...

$$\left. \begin{aligned} p_x &= l_1 c_1 + l_2 c_{12} = l_1 c_1 + l_2 (c_1 c_2 - s_1 s_2) \\ p_y &= l_1 s_1 + l_2 s_{12} = l_1 s_1 + l_2 (s_1 c_2 + c_1 s_2) \end{aligned} \right\} \text{linear in } s_1 \text{ and } c_1$$

$$\underbrace{\begin{bmatrix} l_1 + l_2 c_2 & -l_2 s_2 \\ l_2 s_2 & l_1 + l_2 c_2 \end{bmatrix}} \begin{bmatrix} c_1 \\ s_1 \end{bmatrix} = \begin{bmatrix} p_x \\ p_y \end{bmatrix}$$

$$\det = l_1^2 + l_2^2 + 2l_1 l_2 c_2 > 0$$

except if  $l_1 = l_2$  and  $c_2 = -1$   
being then  $q_1$  undefined  
(singular case:  $\infty^1$  solutions)

$$q_1 = \text{atan2}\{s_1, c_1\}$$

$$= \text{atan2}\left\{\frac{p_y(l_1 + l_2 c_2) - p_x l_2 s_2}{\det}, \frac{p_x(l_1 + l_2 c_2) + p_y l_2 s_2}{\det}\right\}$$

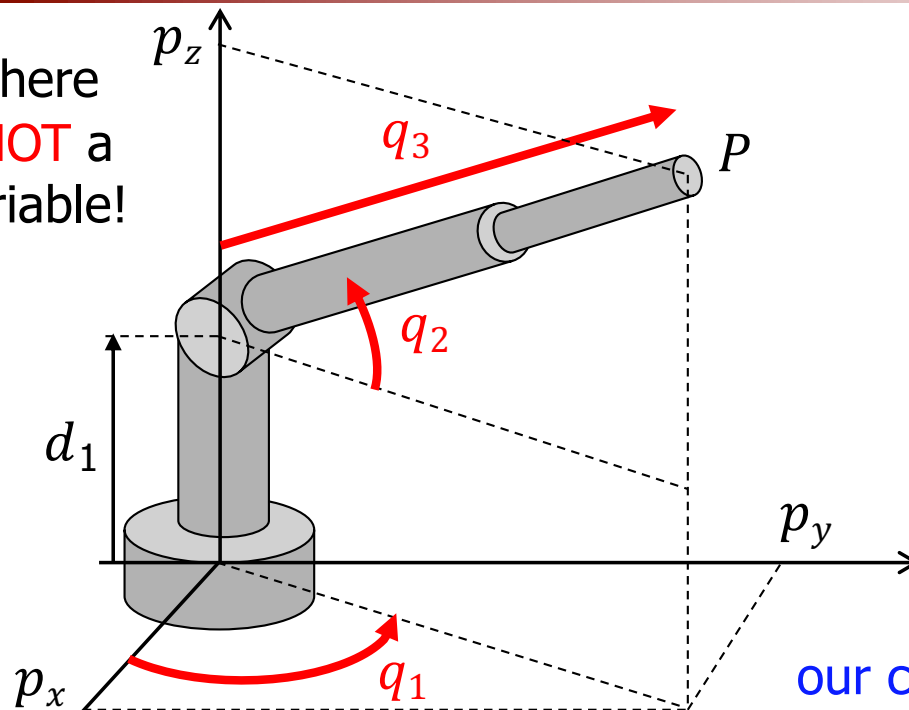
notes: a) this method provides directly the result in  $(-\pi, \pi]$

b) when evaluating atan2,  $\det > 0$  can be simply eliminated  
from the expressions of  $s_1$  and  $c_1$  (not changing the result)



# Inverse kinematics of polar (RRP) arm

note: here  $q_2$  is **NOT** a DH variable!



direct kinematics

$$p_x = q_3 c_2 c_1$$

$$p_y = q_3 c_2 s_1$$

$$p_z = d_1 + q_3 s_2$$

$$p_x^2 + p_y^2 + (p_z - d_1)^2 = q_3^2$$

$$q_3 = + \sqrt{p_x^2 + p_y^2 + (p_z - d_1)^2}$$

our choice: take here only the positive value...

if  $q_3 = 0$ , then  $q_1$  and  $q_2$  remain both undefined (**stop**); **else**

$$q_2 = \text{atan2} \left\{ (p_z - d_1) / q_3, \pm \sqrt{p_x^2 + p_y^2} / q_3 \right\}$$

(if we **stop**, it is a **singular** case:  
 $\infty^2$  or  $\infty^1$   
solutions)

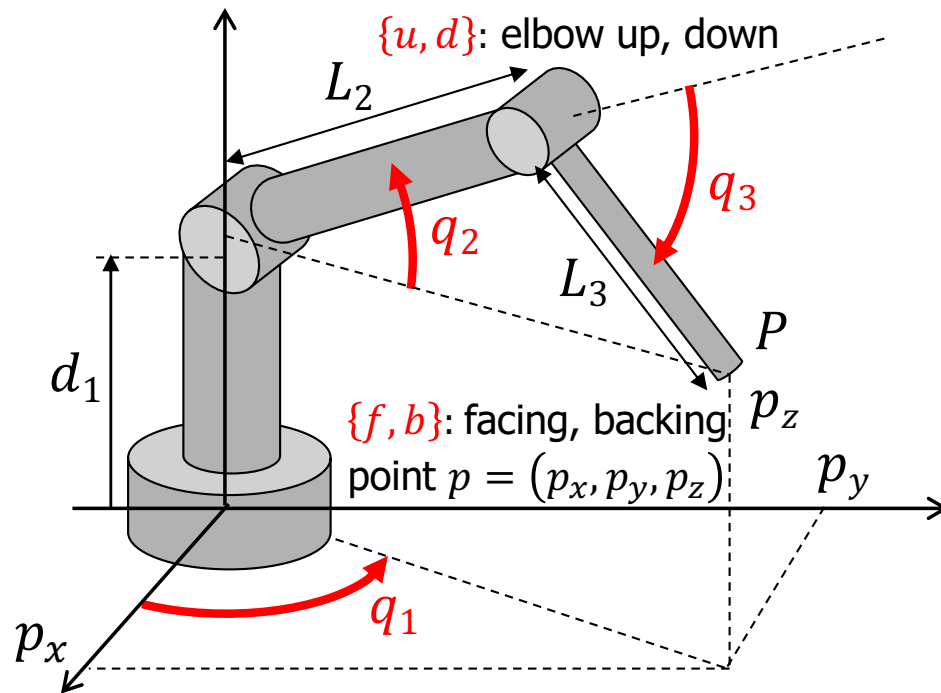
if  $p_x^2 + p_y^2 = 0$ , then  $q_1$  remains undefined (**stop**); **else**

$$q_1 = \text{atan2} \{ p_y / c_2, p_x / c_2 \}$$

(2 **regular** solutions  $\{q_1, q_2, q_3\}$ )

eliminating  $q_3 > 0$  from both arguments

# Inverse kinematics of 3R elbow-type arm



symmetric structure **without** offsets  
e.g., first 3 joints of Mitsubishi PA10 robot

$WS_1 = \{\text{spherical shell centered at } (0,0,d_1),$   
with outer radius  $R_{out} = L_2 + L_3$   
and inner radius  $R_{in} = |L_2 - L_3|\}$

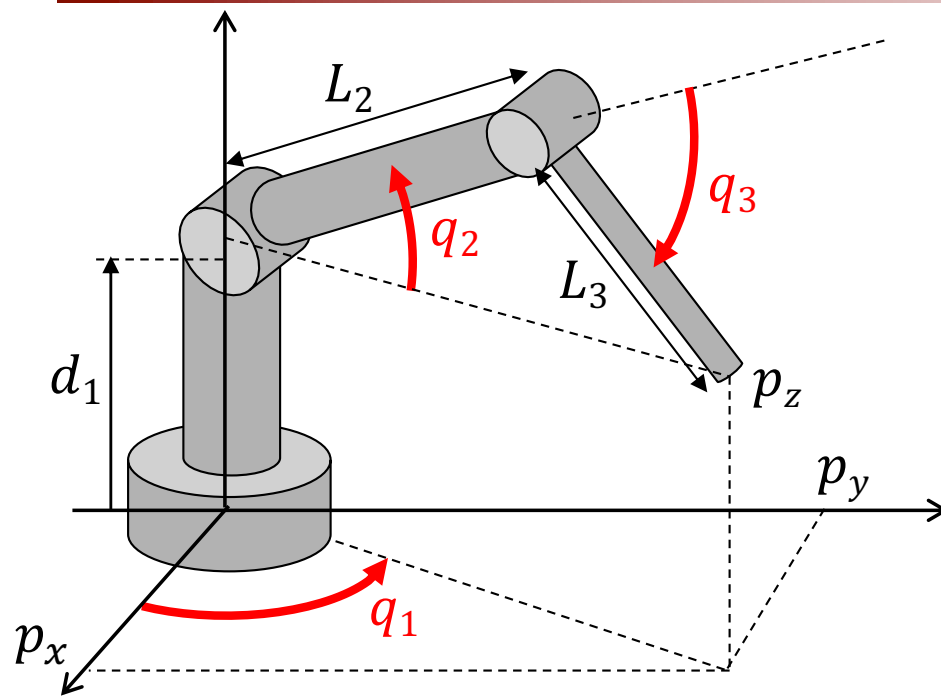


4 **regular** inverse  
kinematics solutions in  
 $\text{int}(WS_1) - \{\text{axis } z_0\}$

more details (e.g., full handling of **singular cases**)  
can be found in the solution of Exercise #1  
in written exam of 11 Apr 2017

# Inverse kinematics of 3R elbow-type arm

## step 1



direct  
kinematics

$$\begin{aligned} p_x &= c_1(L_2c_2 + L_3c_{23}) \\ p_y &= s_1(L_2c_2 + L_3c_{23}) \\ p_z &= d_1 + L_2s_2 + L_3s_{23} \end{aligned}$$

$$\begin{aligned} p_x^2 + p_y^2 + (p_z - d_1)^2 &= c_1^2(L_2c_2 + L_3c_{23})^2 + s_1^2(L_2c_2 + L_3c_{23})^2 + (L_2s_2 + L_3s_{23})^2 \\ &= \dots = L_2^2 + L_3^2 + 2L_2L_3(c_2c_{23} + s_2s_{23}) = L_2^2 + L_3^2 + 2L_2L_3c_3 \end{aligned}$$

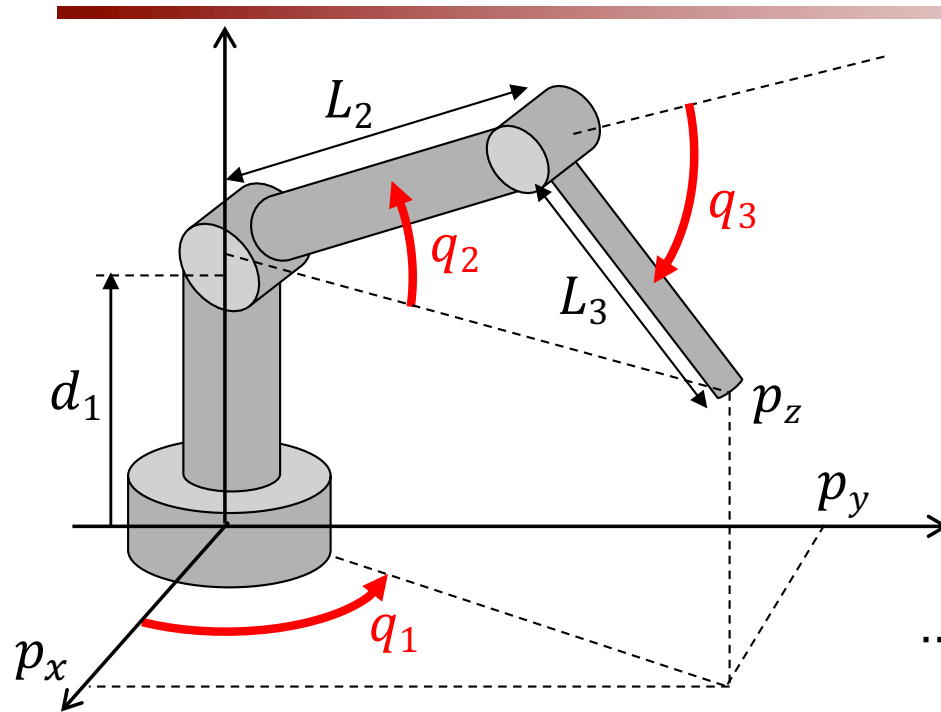
$$c_3 = (p_x^2 + p_y^2 + (p_z - d_1)^2 - L_2^2 - L_3^2) / 2L_2L_3 \in [-1, +1] \text{ (else, } p \text{ is out of workspace!)}$$

$$\downarrow$$

$$\pm s_3 = \pm \sqrt{1 - c_3^2} \quad \Rightarrow \quad \text{two solutions} \quad \left\{ \begin{array}{l} q_3^{\{+\}} = \text{atan2}\{s_3, c_3\} \\ q_3^{\{-\}} = \text{atan2}\{-s_3, c_3\} = -q_3^{\{+\}} \end{array} \right.$$

# Inverse kinematics of 3R elbow-type arm

## step 2



direct  
kinematics

$$\begin{aligned} p_x &= c_1(L_2c_2 + L_3c_{23}) \\ p_y &= s_1(L_2c_2 + L_3c_{23}) \\ p_z &= d_1 + L_2s_2 + L_3s_{23} \end{aligned}$$

... being  $p_x^2 + p_y^2 = (L_2c_2 + L_3c_{23})^2 > 0$

**only** when  $p_x^2 + p_y^2 > 0$  ...  
else  $q_1$  is **undefined** —infinite solutions!

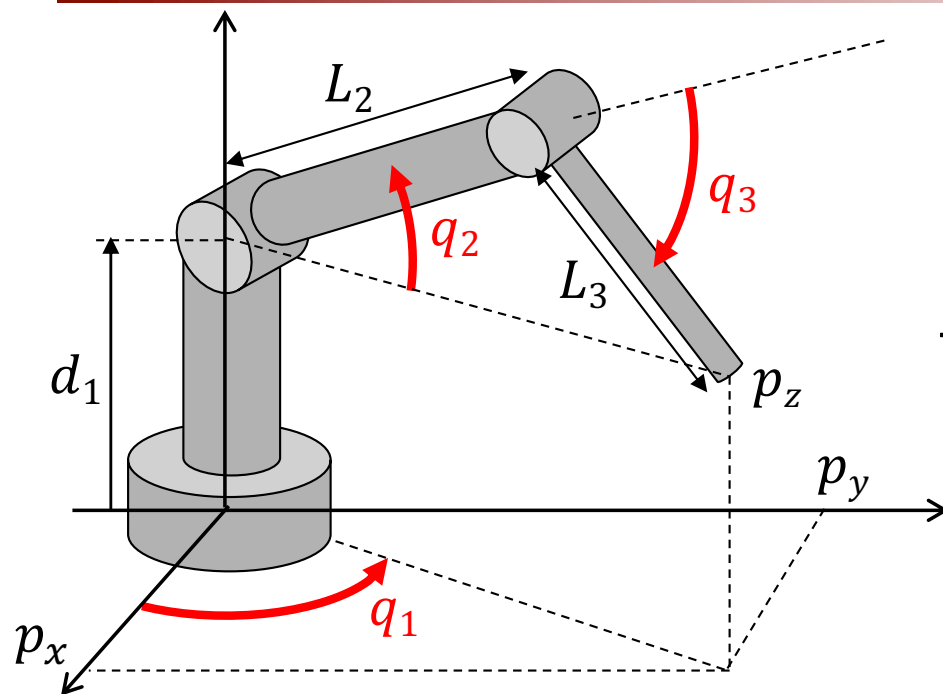
$$\Rightarrow \begin{cases} c_1 = p_x / \pm \sqrt{p_x^2 + p_y^2} \\ s_1 = p_y / \pm \sqrt{p_x^2 + p_y^2} \end{cases}$$

again, two solutions  $\Rightarrow \begin{cases} q_1^{\{+\}} = \text{atan2}\{p_y, p_x\} \\ q_1^{\{-\}} = \text{atan2}\{-p_y, -p_x\} \end{cases}$



# Inverse kinematics of 3R elbow-type arm

## step 3



combine first the two equations of direct kinematics and rearrange the last one

$$\begin{cases} c_1 p_x + s_1 p_y = L_2 c_2 + L_3 c_{23} \\ \quad \quad \quad = (L_2 + L_3 c_3) c_2 - L_3 s_3 s_2 \\ p_z - d_1 = L_2 s_2 + L_3 s_{23} \\ \quad \quad \quad = L_3 s_3 c_2 + (L_2 + L_3 c_3) s_2 \end{cases}$$

define and solve a **linear system**  $Ax = b$  in the **algebraic** unknowns  $x = (c_2, s_2)$

$$\begin{bmatrix} L_2 + L_3 c_3 & -L_3 s_3^{\{+,-\}} \\ L_3 s_3^{\{+,-\}} & L_2 + L_3 c_3 \end{bmatrix} \begin{bmatrix} c_2 \\ s_2 \end{bmatrix} = \begin{bmatrix} c_1^{\{+,-\}} p_x + s_1^{\{+,-\}} p_y \\ p_z - d_1 \end{bmatrix}$$

coefficient matrix  $A$

known vector  $b$

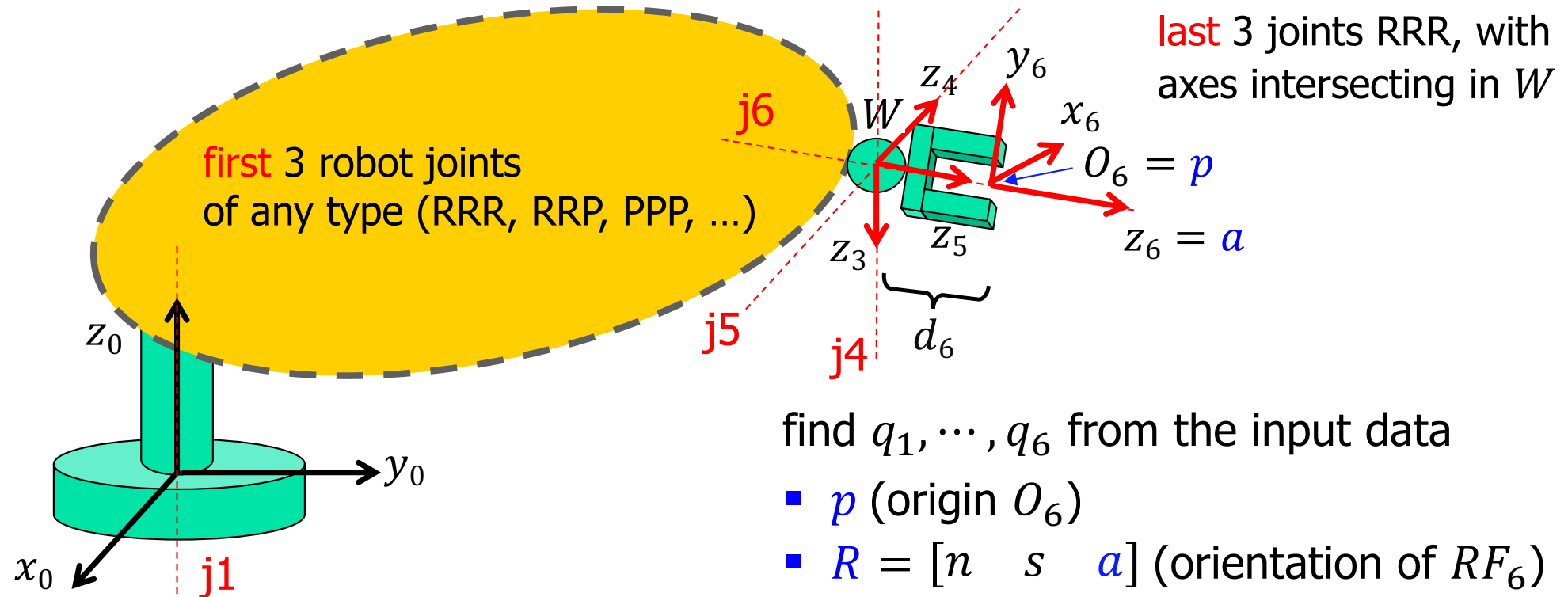
provided  $\det A = p_x^2 + p_y^2 + (p_z - d_1)^2 \neq 0$   
(**always** true if  $L_2 \neq L_3$ !) ...

else,  $q_2$  is **undefined** —infinite solutions!

4 **regular** solutions for  $q_2$ ,  
depending on the combinations  
of  $\{+, -\}$  from  $q_1$  and  $q_3$

$$q_2^{\{\{f,b\},\{u,d\}\}} = \text{atan2} \left\{ s_2^{\{\{f,b\},\{u,d\}\}}, c_2^{\{\{f,b\},\{u,d\}\}} \right\}$$

# Inverse kinematics for robots with spherical wrist

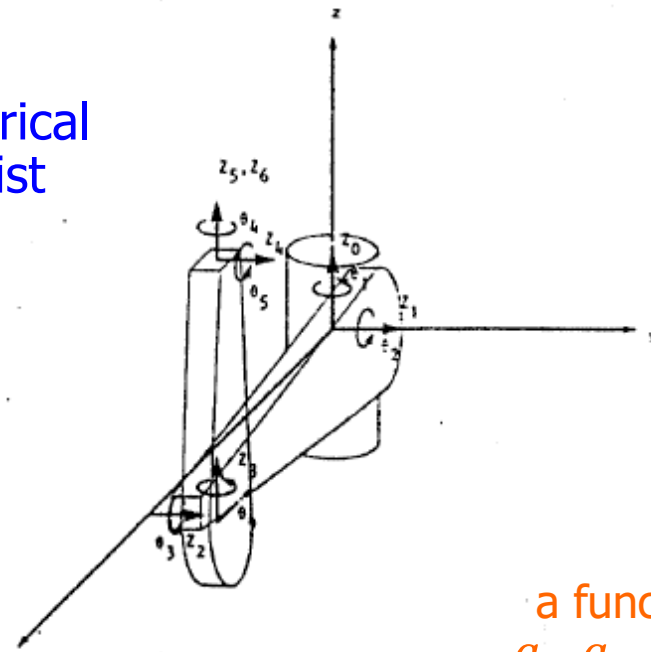


1.  $W = p - d_6 a \Rightarrow q_1, q_2, q_3$  (inverse "position" kinematics for main axes)
  2.  $R = {}^0R_3(q_1, q_2, q_3) \underbrace{{}^3R_6(q_4, q_5, q_6)}_{\text{Euler ZYZ or ZXZ rotation matrix with } q_4, q_5, q_6 (\theta_4, \theta_5, \theta_6)} \Rightarrow {}^3R_6(q_4, q_5, q_6) = {}^0R_3^T R \Rightarrow q_4, q_5, q_6$  (inverse "orientation" kinematics for the wrist)
- given  $\uparrow$  known, after step 1  $\uparrow$  two regular solutions  $\rightarrow$



# 6R robot Unimation PUMA 600

spherical wrist



a function of  $q_1, q_2, q_3$  only!

TABLE I  
LINK PARAMETERS FOR PUMA ARM

Joint	$\alpha^a$	$\theta^a$	$d$	$a$	Range
1	$-90^\circ$	$\theta_1$	0	0	$\theta_1: +/ - 160^\circ$
2	0	$\theta_2$	0	$a_2$	$\theta_2: +45^\circ \rightarrow -225^\circ$
3	$90^\circ$	$\theta_3$	$d_3$	$a_3$	$\theta_3: 225^\circ \rightarrow -45^\circ$
4	$-90^\circ$	$\theta_4$	$d_4$	0	$\theta_4: +/ - 170^\circ$
5	$90^\circ$	$\theta_5$	0	0	$\theta_5: +/ - 135^\circ$
6	0	$\theta_6$	0	0	$\theta_6: +/ - 170^\circ$

$a_1 = 17.000$     $a_3 = 0.75$   
 $d_3 = 4.937$     $d_4 = 17.000$

here  $d_6 = 0$ ,  
so that  $O_6 = W$  directly

$$\begin{aligned}
 n_x &= C_1[C_{23}(C_4C_5C_6 - S_4S_6) - S_{23}S_5C_6] - S_1[S_4C_5C_6 + C_4S_6] \\
 n_y &= S_1[C_{23}(C_4C_5C_6 - S_4S_6) - S_{23}S_5C_6] + C_1[S_4C_5C_6 + C_4S_6] \\
 n_z &= -S_{23}(C_4C_5C_6 - S_4S_6) - C_{23}S_5C_6 \\
 o_x &= C_1[-C_{23}(C_4C_5S_6 + S_4C_6) + S_{23}S_5S_6] - S_1[-S_4C_5S_6 + C_4C_6] \\
 o_y &= S_1[-C_{23}(C_4C_5S_6 + S_4C_6) + S_{23}S_5S_6] + C_1[-S_4C_5S_6 + C_4C_6] \\
 o_z &= S_{23}(C_4C_5S_6 + S_4C_6) + C_{23}S_5S_6 \\
 a_x &= C_1(C_{23}C_4S_5 + S_{23}C_5) - S_1S_4S_5 \\
 a_y &= S_1(C_{23}C_4S_5 + S_{23}C_5) + C_1S_4S_5 \\
 a_z &= -S_{23}C_4S_5 + C_{23}C_5 \\
 p_x &= C_1(d_4S_{23} + a_3C_{23} + a_2C_2) - S_1d_3 \\
 p_y &= S_1(d_4S_{23} + a_3C_{23} + a_2C_2) + C_1d_3 \\
 p_z &= -(-d_4C_{23} + a_3S_{23} + a_2S_2)
 \end{aligned}$$

$n = {}^0x_6(q)$   
 $s = {}^0y_6(q)$   
 $a = {}^0z_6(q)$   
 $p = O_6(q)$

8 different (regular) inverse solutions  
that can be found in closed form



# Finding nice kinematic relations

## whiteboard ...

- the most complex inverse kinematics that can be solved in principle in closed form (i.e., **analytically**) is that of a **6R serial manipulator**, with arbitrary DH table
  - ways to systematically generate equations from the direct kinematics that could be easier to solve  $\Rightarrow$  some scalar equations may contain perhaps **a single unknown variable!**

method used for the Unimation PUMA 600 in (\*)

$${}^0T_6 = {}^0A_1(\theta_1) {}^1A_2(\theta_2) \cdots {}^5A_6(\theta_6) = U_0$$

$${}^0A_1^{-1} {}^0T_6 = U_1 (= {}^1A_2 \cdots {}^5A_6)$$

$${}^1A_2^{-1} {}^0A_1^{-1} {}^0T_6 = U_2 (= {}^2A_3 \cdots {}^5A_6)$$

$$\cdots$$

$${}^4A_5^{-1} \cdots {}^1A_2^{-1} {}^0A_1^{-1} {}^0T_6 = U_5 (= {}^5A_6)$$

or also ...

$${}^0T_6 {}^5A_6^{-1} = V_5 (= {}^0A_1 \cdots {}^4A_5)$$

$${}^0T_6 {}^5A_6^{-1} {}^4A_5^{-1} = V_4 (= {}^0A_1 \cdots {}^3A_4)$$

$$\cdots$$

$${}^0T_6 {}^5A_6^{-1} {}^4A_5^{-1} \cdots {}^1A_2^{-1} = V_1 (= {}^0A_1)$$

(\*) Paul, Shimano, and Mayer: IEEE Transactions on Systems, Man, and Cybernetics, 1981

- generating from the direct kinematics a reduced set of equations to be solved (setting w.l.o.g.  $d_1 = d_6 = 0$ )  $\Rightarrow$  **4 compact scalar equations** in the 4 unknowns  $\theta_2, \dots, \theta_5$

$${}^0T_6 = \begin{bmatrix} n & s & a & p \\ 0 & 0 & 0 & 1 \end{bmatrix} = {}^0A_6(\theta) \longrightarrow \begin{aligned} a_z &= a^T(\theta) z & \|p\|^2 &= p^T(\theta) p(\theta) \\ p_z &= p^T(\theta) z & p^T a &= p^T(\theta) a(\theta) \end{aligned}$$

$z = [0 \quad 0 \quad 1]^T$

... then solve easily for the remaining  $\theta_1$  and  $\theta_6$

solved analytically or numerically ...

Manseur and Doty: International Journal of Robotics Research, 1988

# Numerical solution of inverse kinematics problems



- use when a closed-form solution  $q$  to  $r_d = f_r(q)$  does not exist or is “too hard” to be found
  - all methods are **iterative** and need the matrix  $J_r(q) = \frac{\partial f_r(q)}{\partial q}$  (analytical Jacobian)
  - **Newton method** (here only for  $m = n$ , at the  $k$ -th iteration)
    - $r_d = f_r(q) = f_r(q^k) + J_r(q^k)(q - q^k) + o(\|q - q^k\|)$  ← neglected in Taylor expansion
- $$q^{k+1} = q^k + J_r^{-1}(q^k) [r_d - f_r(q^k)]$$
- convergence for  $q^0$  (initial guess) **close enough** to some  $q^*$ :  $f_r(q^*) = r_d$
  - problems near **singularities** of the Jacobian matrix  $J_r(q)$
  - in case of robot redundancy ( $m < n$ ), use the **pseudoinverse**  $J_r^\#(q)$
  - has **quadratic** (fast!) convergence rate when near to a solution (under a Lipschitz condition for  $J_r$ , otherwise **superlinear** convergence)



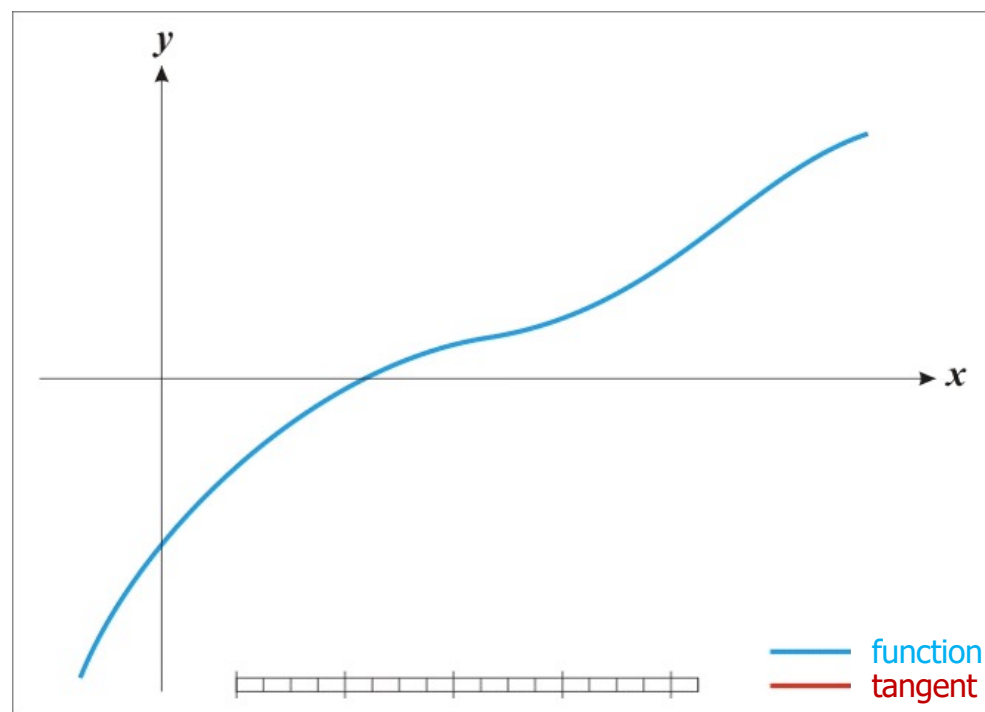
# Operation of Newton method

- in the **scalar** case, also known as “method of the tangent”
- for a differentiable function  $f(x)$ , find a root  $x^*$  of  $f(x^*) = 0$  by iterating as

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)} \quad \rightarrow$$

an approximating sequence

$$\{x_1, x_2, x_3, x_4, x_5, \dots\} \rightarrow x^*$$



animation from  
[http://en.wikipedia.org/wiki/File:NewtonIteration\\_Ani.gif](http://en.wikipedia.org/wiki/File:NewtonIteration_Ani.gif)

# Numerical solution of inverse kinematics problems (cont'd)



- **Gradient method** (steepest descent)

- minimize the **error** function

$$H(q) = \frac{1}{2} \|r_d - f_r(q)\|^2 = \frac{1}{2} (r_d - f_r(q))^T (r_d - f_r(q))$$

$$q^{k+1} = q^k - \alpha \nabla_q H(q^k)$$

from

$$\nabla_q H(q) = (\partial H(q) / \partial q)^T = - \left( (r_d - f_r(q))^T (\partial f_r(q) / \partial q) \right)^T = -J_r^T(q) (r_d - f_r(q))$$

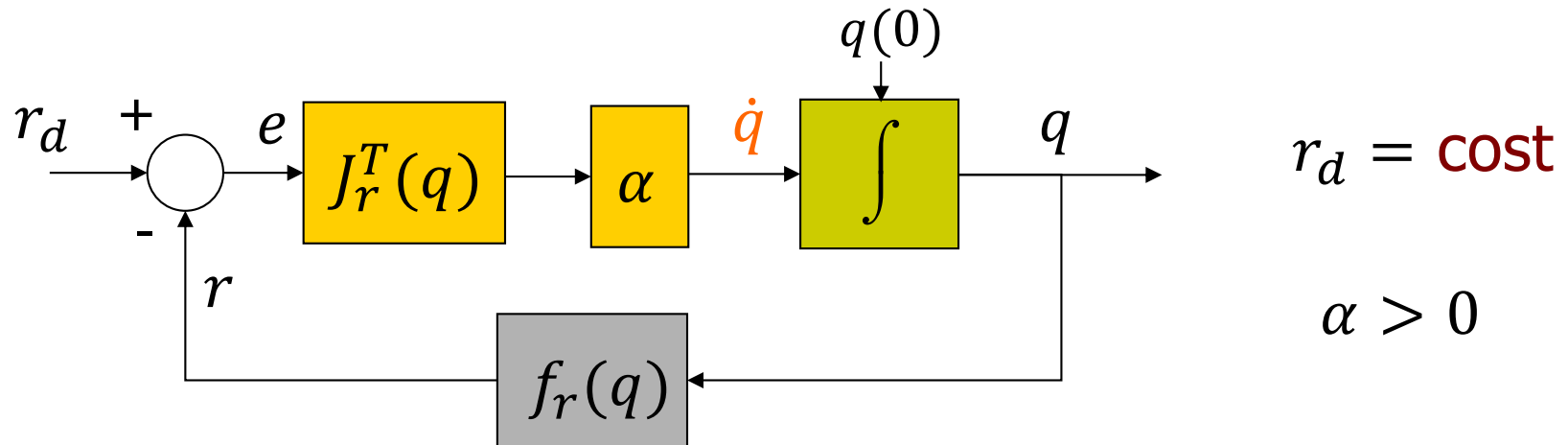
we get

$$q^{k+1} = q^k + \alpha J_r^T(q^k) (r_d - f_r(q^k))$$

- the scalar **step size**  $\alpha > 0$  should be chosen so as to guarantee a decrease of the error function at each iteration: too large values for  $\alpha$  may lead the method to “miss” the minimum
- when the step size is too small, convergence is extremely **slow**



# Revisited as a feedback scheme



$e = r_d - f_r(q) \rightarrow 0 \iff$  closed-loop **equilibrium**  $e = 0$   
is **asymptotically stable**

$V = \frac{1}{2} e^T e \geq 0$  is a **Lyapunov** candidate function

$$\dot{V} = e^T \dot{e} = e^T \frac{d}{dt} (r_d - f_r(q)) = -e^T J_r(q) \dot{q} = -\alpha e^T J_r(q) J_r^T(q) e \leq 0$$

$\dot{V} = 0 \iff e \in \mathcal{N}(J_r^T(q))$  in particular,  **$e = 0$**

↑  
null space

asymptotic stability



# Properties of Gradient method

- **computationally simpler**: use the **Jacobian transpose**, rather than its (pseudo)inverse
- same use also for robots that are **redundant** ( $n > m$ ) for the task
- may not converge to a solution, but it **never diverges**
- the **discrete-time** evolution of the continuous scheme

$$q^{k+1} = q^k + \Delta T J_r^T(q^k)(r_d - f_r(q^k)), \quad \alpha = \Delta T$$

is equivalent to an iteration of the Gradient method

- the scheme can be accelerated by using a gain matrix  $K > 0$

$$\dot{q} = J_r^T(q) K e = J_r^T(q) K (r_d - f_r(q))$$

**note**:  $K \rightarrow K + K_s$ , with  $K_s$  skew-symmetric, can be used also to “escape” from being stuck in a **stationary point** of  $V = \frac{1}{2} e^T K e$ , by **rotating** the error  $K e$  out of the null space of  $J_r^T$  (when a **singularity** is encountered)



# A case study

## analytic expressions of Newton and gradient iterations

- 2R robot with  $l_1 = l_2 = 1$ , desired end-effector position  $r_d = p_d = (1,1)$
- direct kinematic function and error

$$f_r(q) = \begin{pmatrix} c_1 + c_{12} \\ s_1 + s_{12} \end{pmatrix} \quad e = p_d - f_r(q) = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - f_r(q)$$

- Jacobian matrix

$$J_r(q) = \frac{\partial f_r(q)}{\partial q} = \begin{pmatrix} -(s_1 + s_{12}) & -s_{12} \\ c_1 + c_{12} & c_{12} \end{pmatrix}$$

- **Newton** versus **Gradient** iteration

$$q^{k+1} = q^k + \begin{pmatrix} \frac{1}{s_2} \begin{pmatrix} c_{12} & s_{12} \\ -(c_1 + c_{12}) & -(s_1 + s_{12}) \end{pmatrix} \\ \alpha \begin{pmatrix} -(s_1 + s_{12}) & c_1 + c_{12} \\ -s_{12} & c_{12} \end{pmatrix} \end{pmatrix}_{|q=q^k} \times \begin{pmatrix} 1 - (c_1 + c_{12}) \\ 1 - (s_1 + s_{12}) \end{pmatrix}_{|q=q^k} \times e_k$$

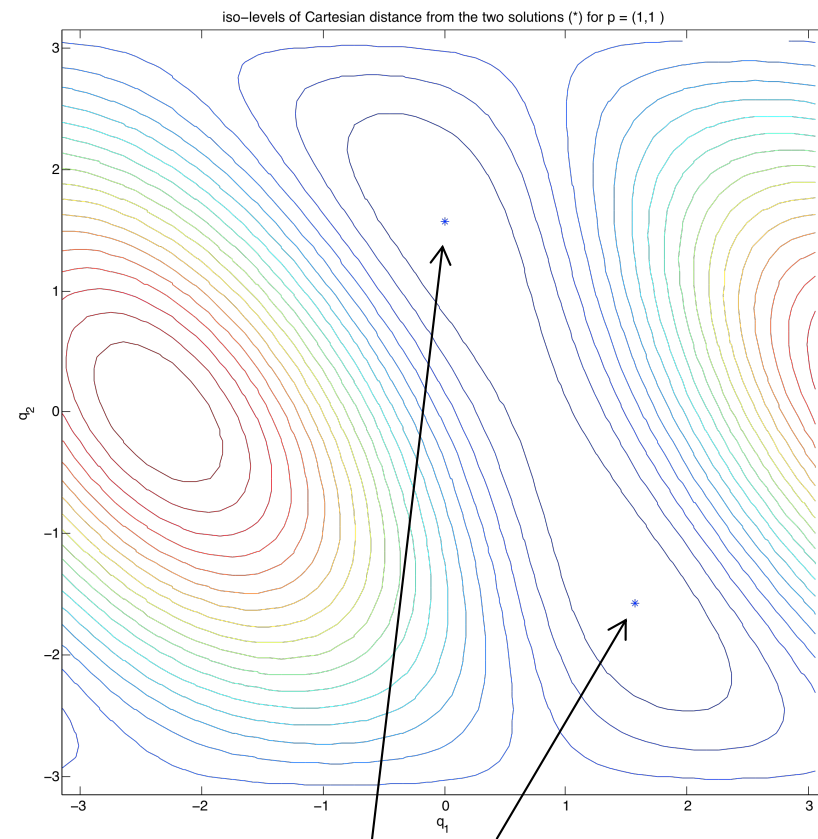
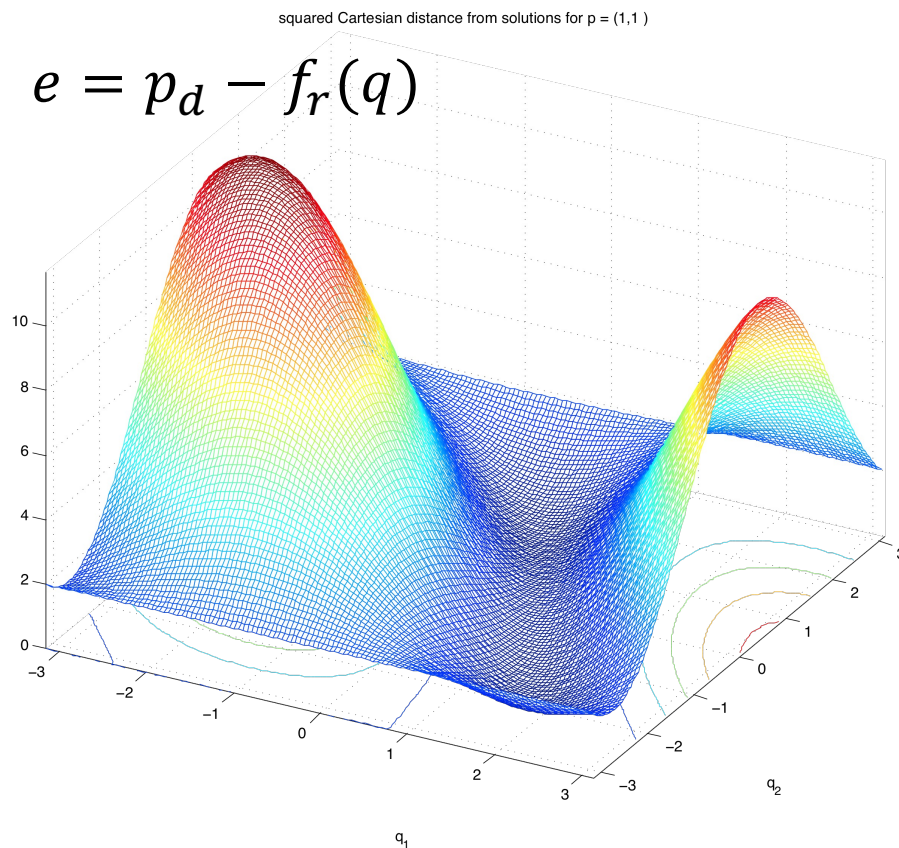
$\det J_r(q)$  points to  $\frac{1}{s_2}$

$J_r^{-1}(q^k)$  points to the first matrix

$J_r^T(q^k)$  points to the second matrix

# Error function

- 2R robot with  $l_1 = l_2 = 1$  and desired end-effector position  $p_d = (1,1)$



plot of  $\|e\|^2$  as a function of  $q = (q_1, q_2)$

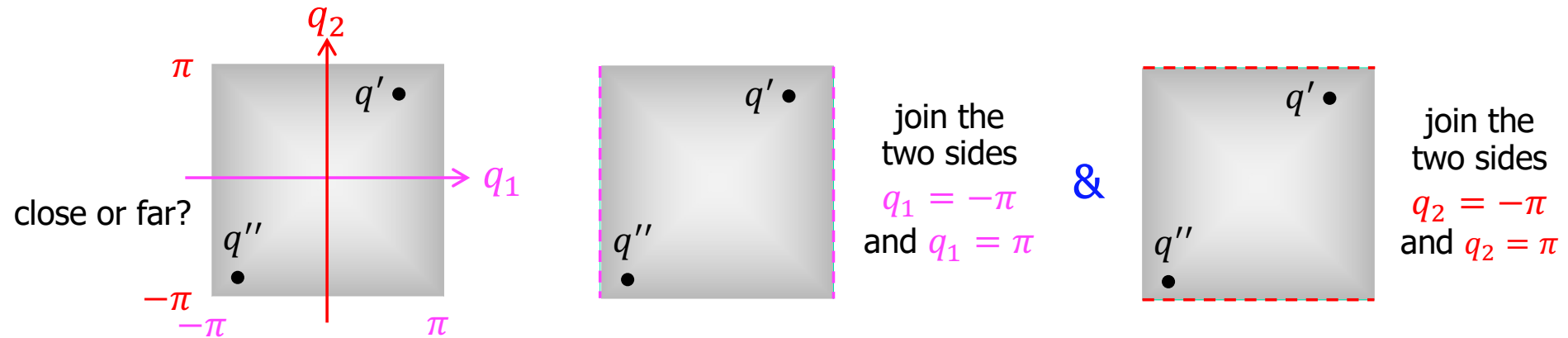
two local minima  
(inverse kinematic solutions)



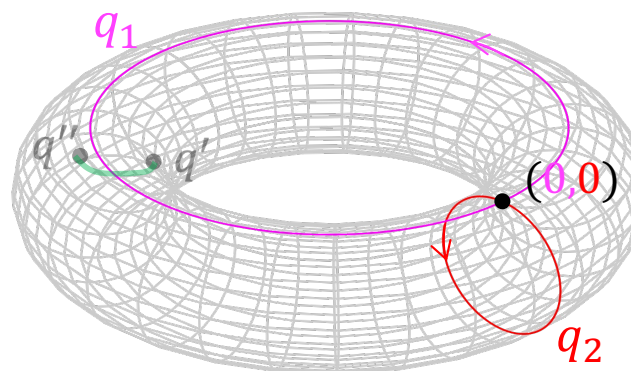
# Configuration space of 2R robot

## whiteboard ...

- can we represent the correct “distance” between two configurations  $q'$  and  $q''$  of this robot on a (square) region in  $\mathbb{R}^2$ ?



- configuration space is a **torus**  $SO(1) \times SO(1)$ , i.e., the surface of a “donut”

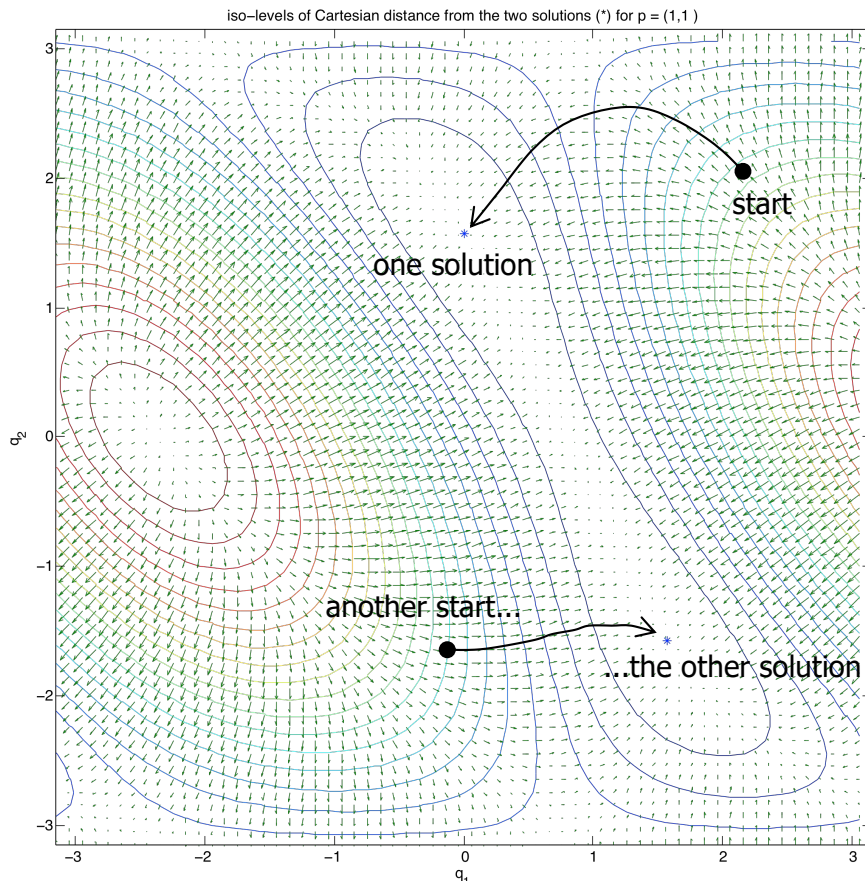


- the right metric is a **geodesic** on the torus ...

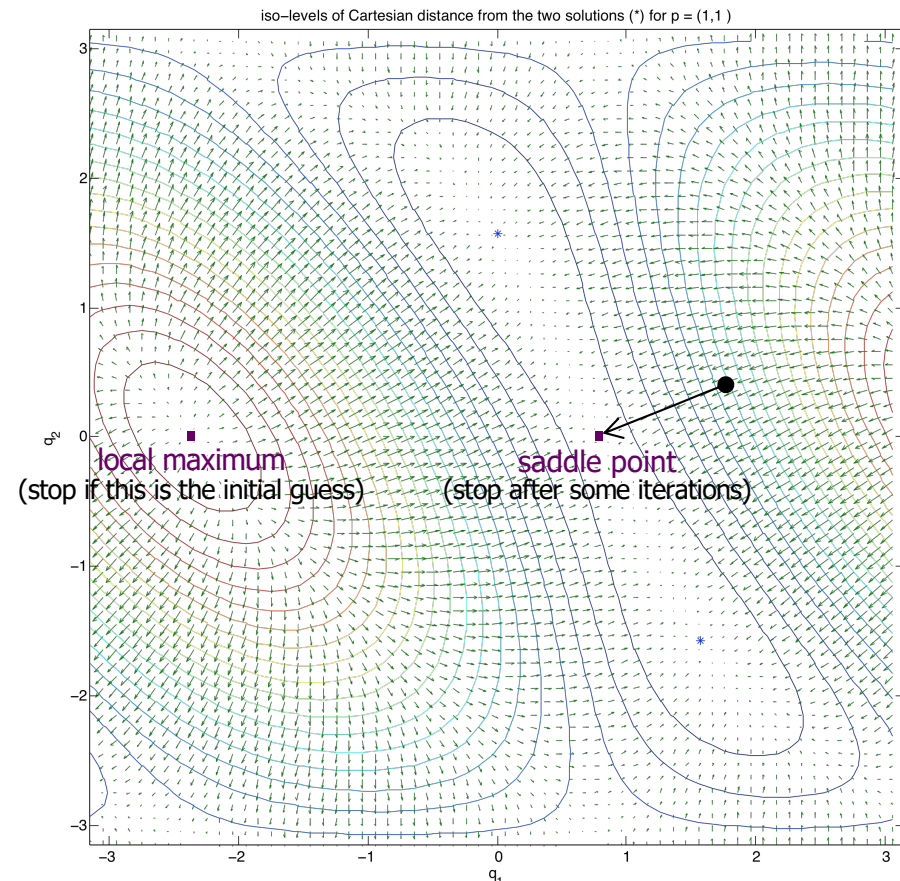


# Error reduction by Gradient method

- flow of iterations along the **negative** (or anti-) gradient
- two possible cases: convergence or stuck (at **zero gradient**)



$$(q_1, q_2)' = (0, \pi/2) \quad (q_1, q_2)'' = (\pi/2, -\pi/2)$$



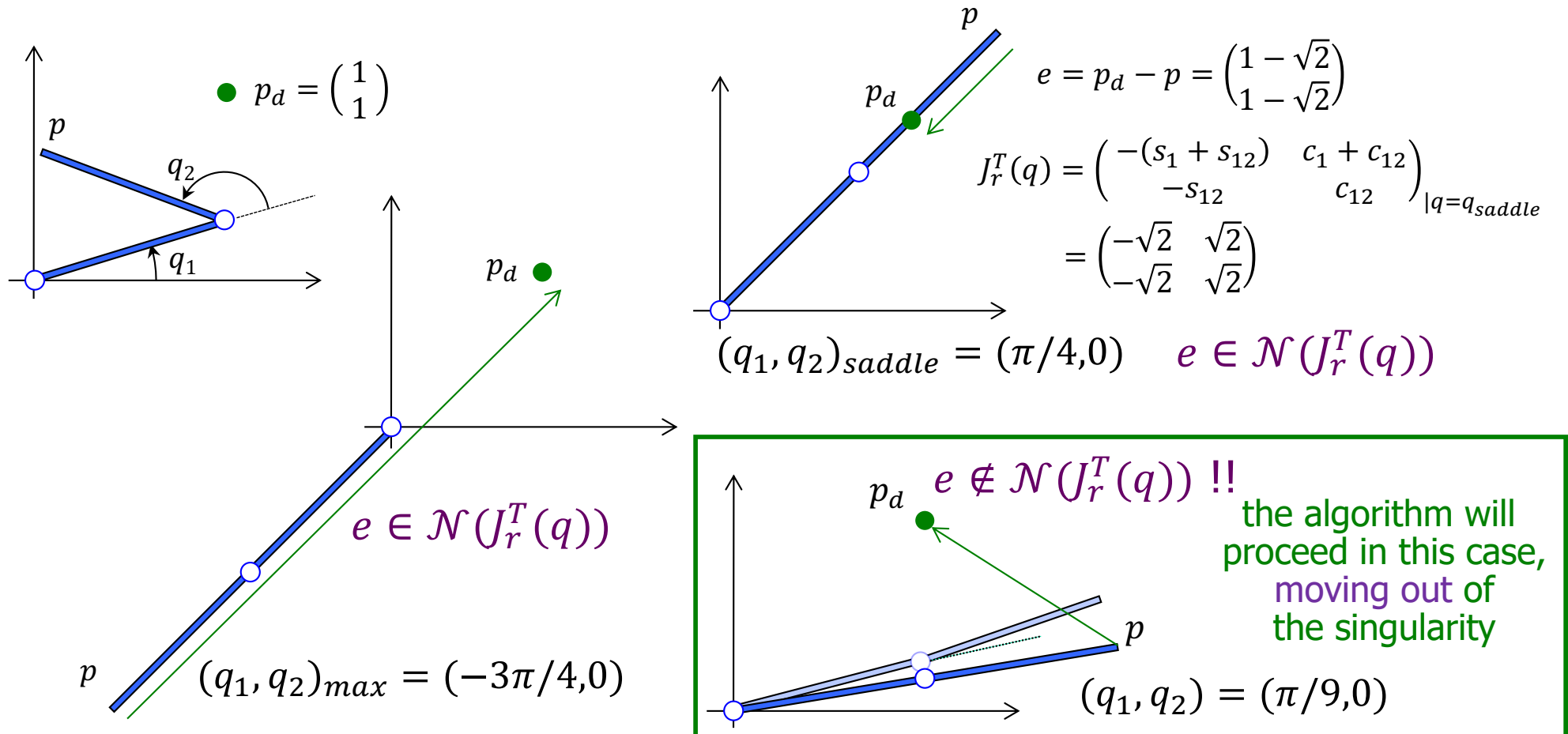
$$(q_1, q_2)_{max} = (-3\pi/4, 0) \quad (q_1, q_2)_{saddle} = (\pi/4, 0)$$

$$e \in \mathcal{N}(J_r^T(q)) !$$

# Convergence analysis

when does the gradient method get stuck?

- lack of convergence occurs when
  - the Jacobian matrix  $J_r(q)$  is not full rank (the robot is in a "singular configuration")
  - **AND** the error  $e$  is in the null space of  $J_r^T(q)$





# Issues in implementation

- initial guess  $q^0$ 
  - only **one** inverse solution is generated for each guess
  - multiple initializations for obtaining other solutions
- optimal step size  $\alpha > 0$  in Gradient method
  - a constant step may work good initially, but not close to the solution (or vice versa)
  - an **adaptive** one-dimensional line search (e.g., Armijo's rule) could be used to choose the best  $\alpha$  at each iteration

- stopping criteria

**Cartesian error**  
(possibly, separate for position and orientation)  $\|r_d - f_r(q^k)\| \leq \varepsilon$       **algorithm increment**  $\|q^{k+1} - q^k\| \leq \varepsilon_q$

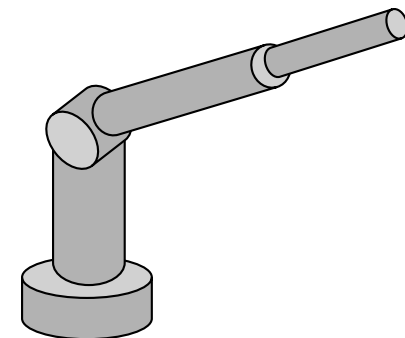
- understanding closeness to singularities

$\sigma_{\min}\{J_r(q^k)\} \geq \sigma_0$       **good numerical conditioning of Jacobian matrix (SVD)**  
(or a simpler test on its determinant, for  $m = n$ )





# Numerical tests on RRP robot

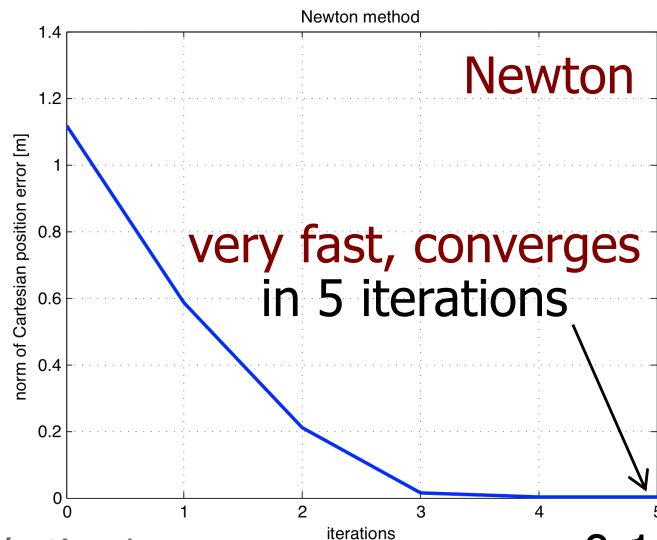
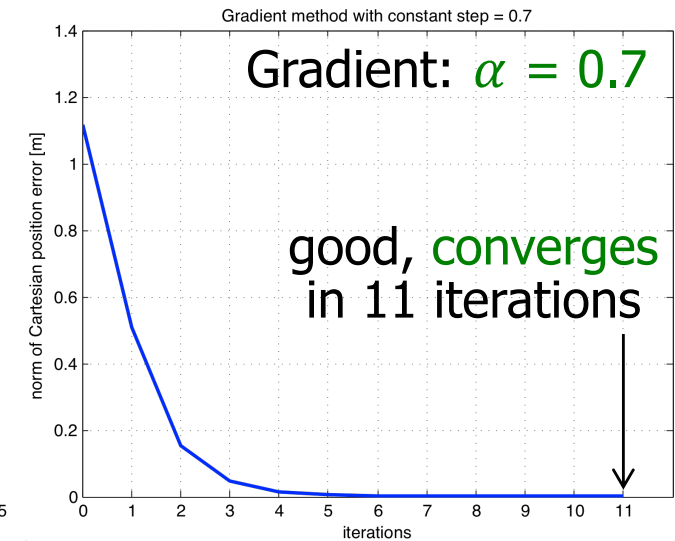
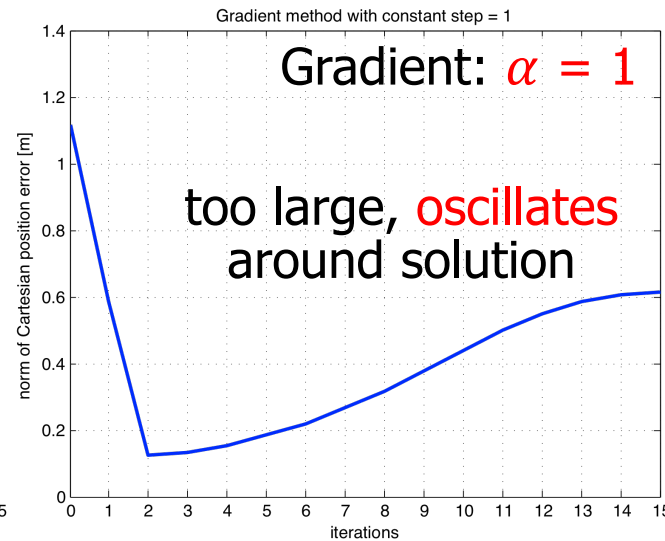
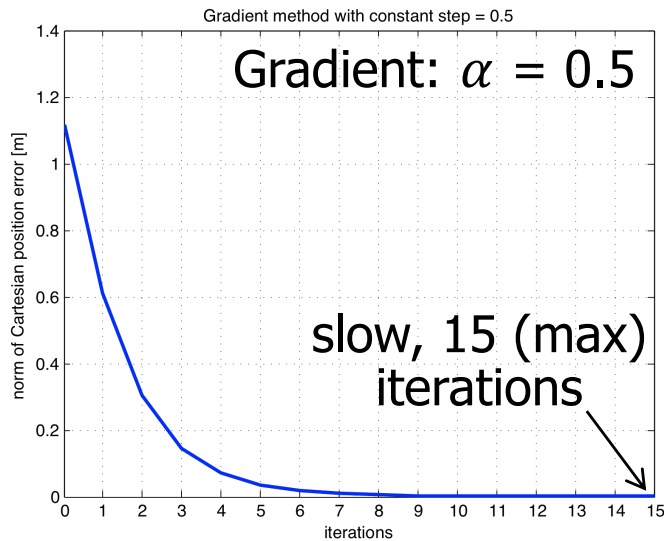


- **RRP/polar robot**: desired E-E position  $r_d = p_d = (1, 1, 1)$   
—see **slide #22**, with  $d_1 = 0.5$
- the two (known) **analytical** solutions, with  $q_3 \geq 0$ , are
$$q^* = (0.7854, 0.3398, 1.5)$$
$$q^{**} = (q_1^* - \pi, \pi - q_2^*, q_3^*) = (-2.3562, 2.8018, 1.5)$$
- norms  $\varepsilon = 10^{-5}$  (max Cartesian error),  $\varepsilon_q = 10^{-6}$  (min joint increment)
- $k_{max} = 15$  (max # iterations),  $|\det J_r(q)| \leq 10^{-4}$  (singularity closeness)
- **numerical** performance of Gradient (with different steps  $\alpha$ ) vs. Newton
  - **test 1**:  $q^0 = (0, 0, 1)$  as initial guess
  - **test 2**:  $q^0 = (-\pi/4, \pi/2, 1)$  — “singular” start, since  $c_2 = 0$  (see **slide #22**)
  - **test 3**:  $q^0 = (0, \pi/2, 0)$  — “doubly singular” start, since also  $q_3 = 0$
  - solution and plots with MATLAB code

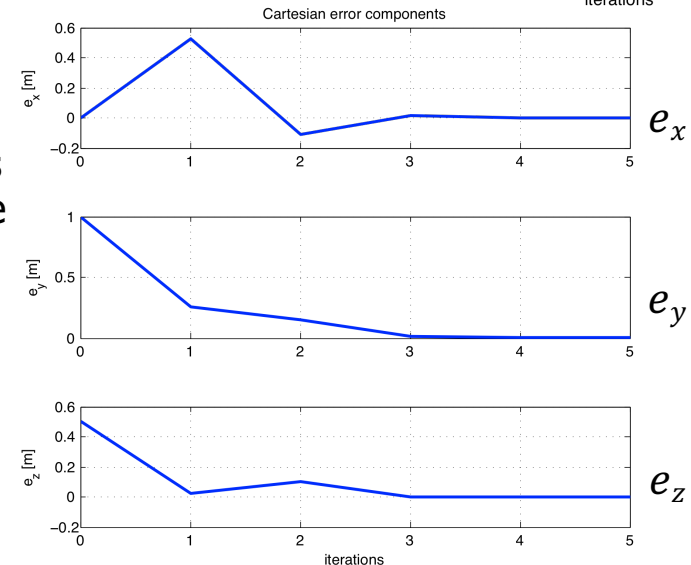


# Numerical test - 1

- **test 1:**  $q^0 = (0, 0, 1)$  as initial guess; evolution of the **error norm**



Cartesian errors component-wise



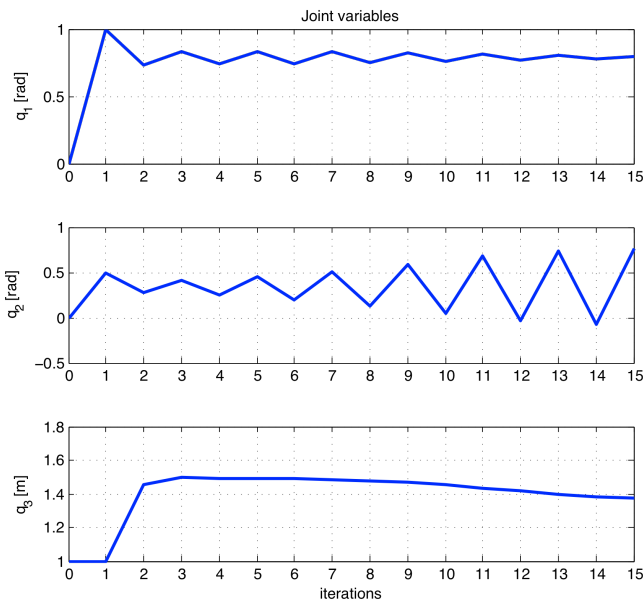
$0.57 \cdot 10^{-5}$

$0.15 \cdot 10^{-8}$



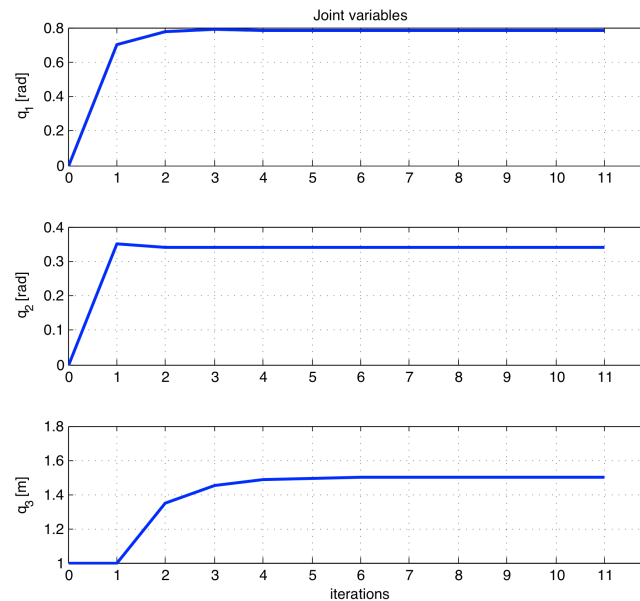
# Numerical test - 1

- test 1:  $q^0 = (0, 0, 1)$  as initial guess; evolution of joint variables



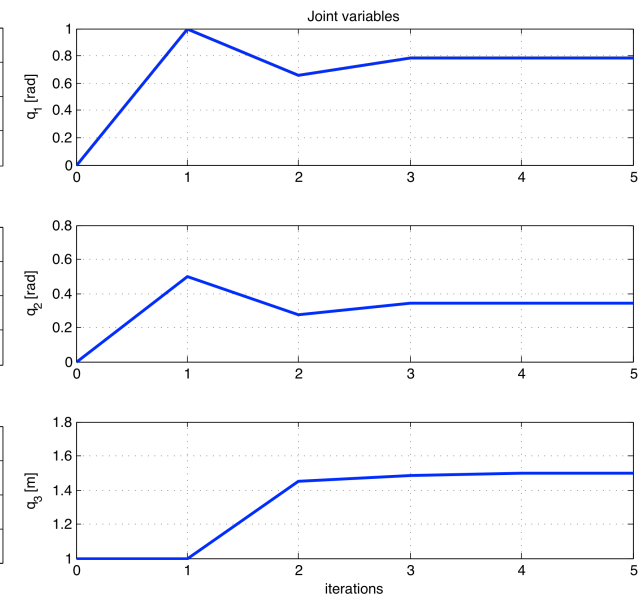
Gradient:  $\alpha = 1$

not converging  
to a solution



Gradient:  $\alpha = 0.7$

converges in  
11 iterations



Newton

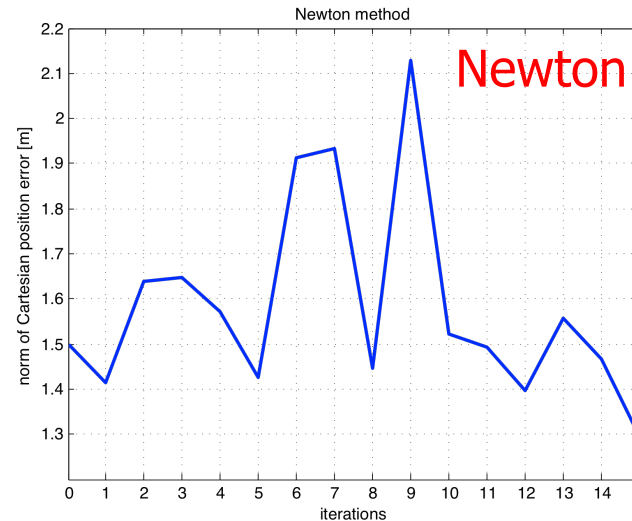
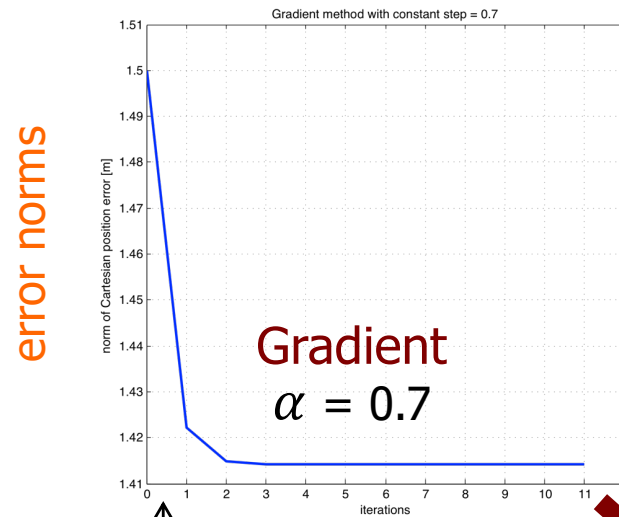
converges in  
5 iterations

both to the same solution  $q^* = (0.7854, 0.3398, 1.5)$



# Numerical test - 2

- **test 2:**  $q^0 = (-\pi/4, \pi/2, 1)$ : singular start

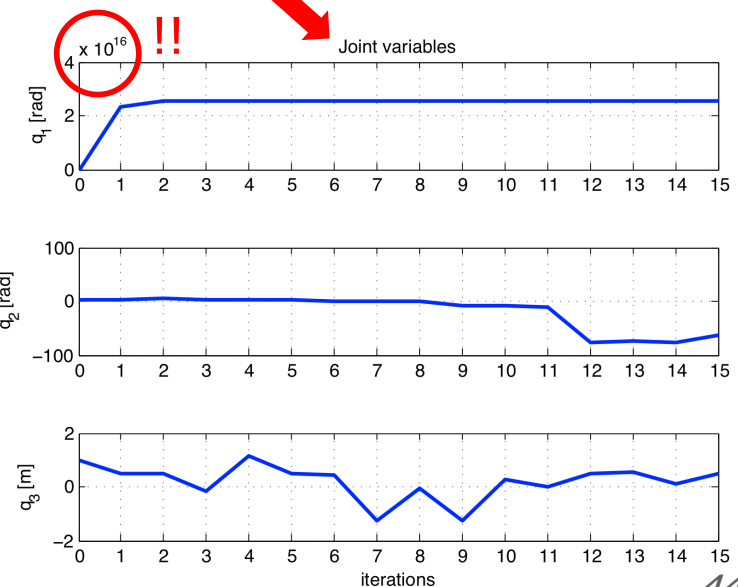
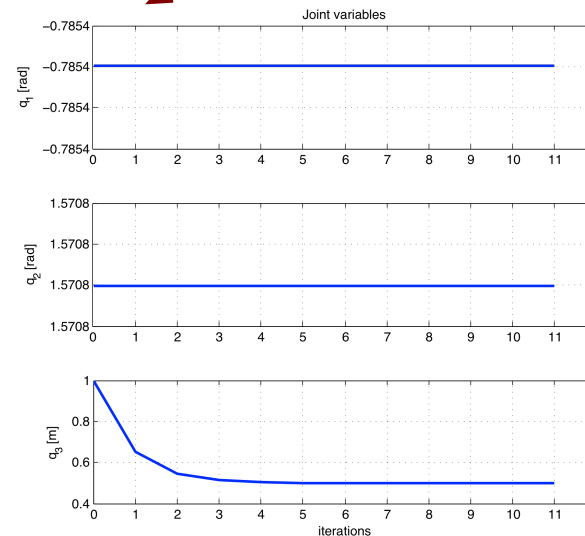


with check of singularity:  
blocked at start

without check:  
it diverges!

starts toward solution, but slowly stops  
(in singularity):  
when Cartesian error vector  $e \in \mathcal{N}(J_r^T(q))$

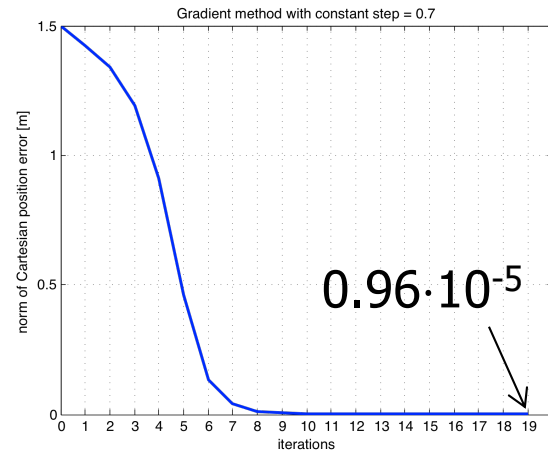
joint variables



# Numerical test - 3

- **test 3:**  $q^0 = (-\pi/4, \pi/2, 1)$ : doubly singular start

error norm



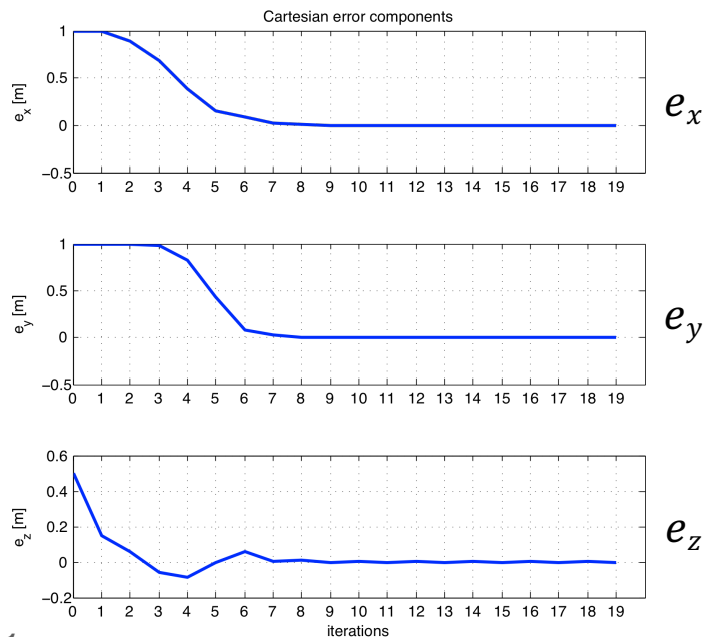
**Gradient** (with  $\alpha = 0.7$ )

- ① starts toward solution
- ② **exits** the double singularity
- ③ slowly **converges** in 19 iterations to the solution

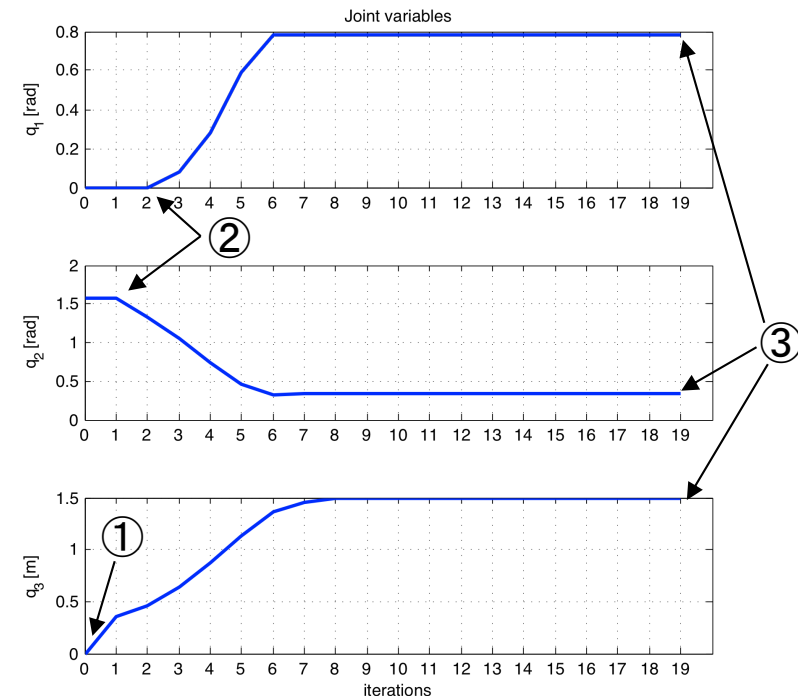
$q^* = (0.7854, 0.3398, 1.5) \Rightarrow$  **"NaN"** in MATLAB

**Newton** is either **blocked at start** or (w/o check) **explodes!**

Cartesian errors



joint variables





# Final remarks

- an **efficient** iterative scheme can be devised by combining
  - **initial iterations** using Gradient (“sure but slow”, **linear** convergence rate)
  - **switch then** to Newton method (**quadratic** terminal convergence rate)
- **joint range limits** are considered only at the end
  - check if the solution found is **feasible**, as for analytical methods
- or, an **optimization** criterion and/or **constraints** included in the search
  - drive iterations toward an inverse kinematic solution with nicer properties
- if the problem has to be solved **on-line**
  - execute iterations and associate an actual robot motion: **repeat steps** at times  $t_0, t_1 = t_0 + T, \dots, t_k = t_{k-1} + T$  (e.g., every  $T = 40$  ms)
  - a “good” choice for the initial guess  $q^0$  at  $t_k$  is the solution of the previous problem at  $t_{k-1}$  (provides continuity, requires only 1-2 Newton iterations)
  - crossing of singularities and handling of joint range limits need special care
- Jacobian-based inversion schemes are used also for **kinematic control**, moving along/tracking a continuous task trajectory  $r_d(t)$