

# **Robotics**

## **PS05 – Solutions**

**Andreas Birk**  
Constructor University

# Part 5: Inverse Kinematics

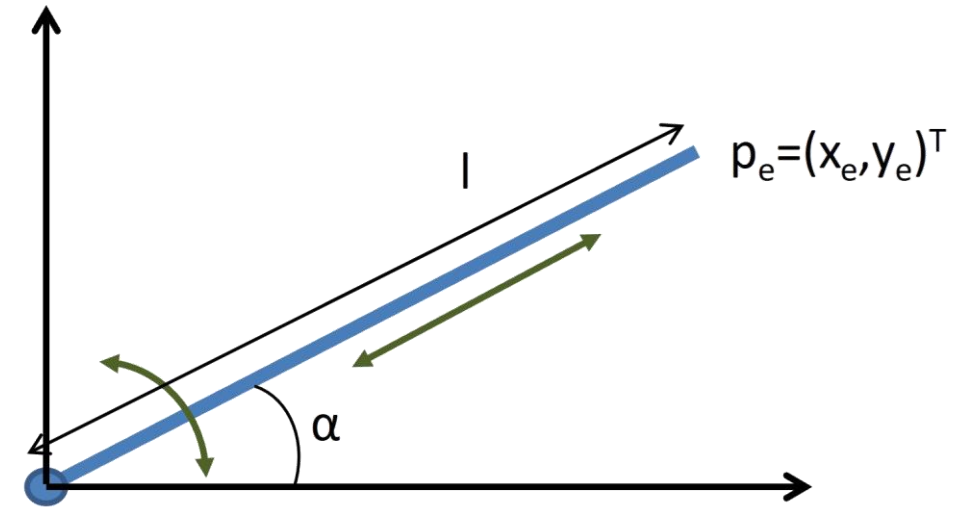
# Problem 1

Take the planar arm from **PS03, Problem 1** with a rotational joint and a prismatic joint linked to it with the DoF  $\alpha$  and  $l$ .

Use its forward kinematics to find

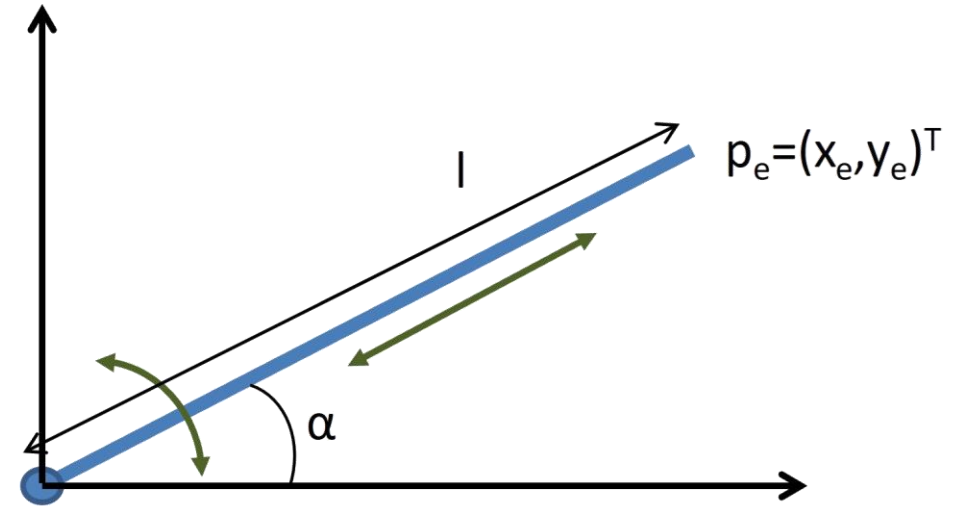
- the proper Jacobian matrix  $J$ , respectively
- the numerical approximation of  $J$  at point  $(1,2)$  with  $\delta = 0.1$

as basis for inverse kinematics.



# Problem 1

$$\begin{aligned}\begin{pmatrix} x_e \\ y_e \\ 1 \end{pmatrix} &= \begin{pmatrix} c\alpha & -s\alpha & 0 \\ s\alpha & c\alpha & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & l \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} c\alpha & -s\alpha & c\alpha \cdot l \\ s\alpha & c\alpha & s\alpha \cdot l \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} c\alpha \cdot l \\ s\alpha \cdot l \\ 1 \end{pmatrix}\end{aligned}$$



FK from PS03, problem 1

# Problem 1

$$f(\alpha, l) = \begin{pmatrix} x_e \\ y_e \end{pmatrix} = \begin{pmatrix} c\alpha \cdot l \\ s\alpha \cdot l \end{pmatrix}$$

$$J = Df(\alpha, l) = \begin{pmatrix} \frac{\partial c\alpha \cdot l}{\partial \alpha} & \frac{\partial c\alpha \cdot l}{\partial l} \\ \frac{\partial s\alpha \cdot l}{\partial \alpha} & \frac{\partial s\alpha \cdot l}{\partial l} \end{pmatrix} = ???$$

# Problem 1

note:

$$\sin'(ax + b) = a \cos(ax + b)$$

$$\cos'(ax + b) = -a \sin(ax + b)$$

example:

$$f(\alpha_1, \alpha_2) = \sin(\alpha_1 + \alpha_2) \Rightarrow \frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = \cos(\alpha_1 + \alpha_2)$$

using 1<sup>st</sup> rule with  $a = 1, x = \alpha_1, b = \alpha_2$

# Problem 1

$$f(\alpha, l) = \begin{pmatrix} x_e \\ y_e \end{pmatrix} = \begin{pmatrix} c\alpha \cdot l \\ s\alpha \cdot l \end{pmatrix}$$

$$J = Df(\alpha, l) = \begin{pmatrix} \frac{\partial c\alpha \cdot l}{\partial \alpha} & \frac{\partial c\alpha \cdot l}{\partial l} \\ \frac{\partial s\alpha \cdot l}{\partial \alpha} & \frac{\partial s\alpha \cdot l}{\partial l} \end{pmatrix} = \begin{pmatrix} -s\alpha \cdot l & c\alpha \\ c\alpha \cdot l & s\alpha \end{pmatrix}$$

# Problem 1

$$Df(\alpha, l) = \begin{pmatrix} -s\alpha \cdot l & c\alpha \\ c\alpha \cdot l & s\alpha \end{pmatrix} \Rightarrow Df(1,2) = \begin{pmatrix} -\mathbf{1.683} & \mathbf{0.540} \\ \mathbf{1.081} & \mathbf{0.841} \end{pmatrix}$$

note (1,2):  
 $\alpha = 1$  in radians  
 $l = 2$  in  $m$   
[SI as default]

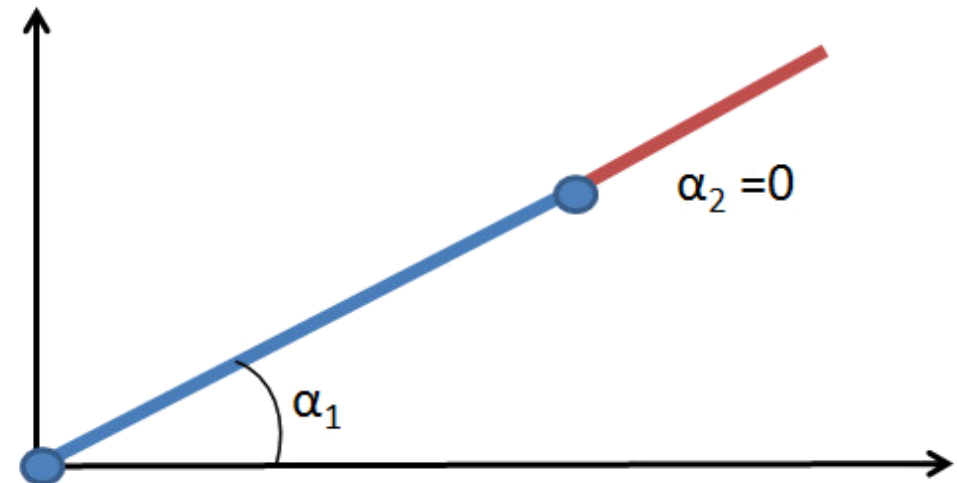
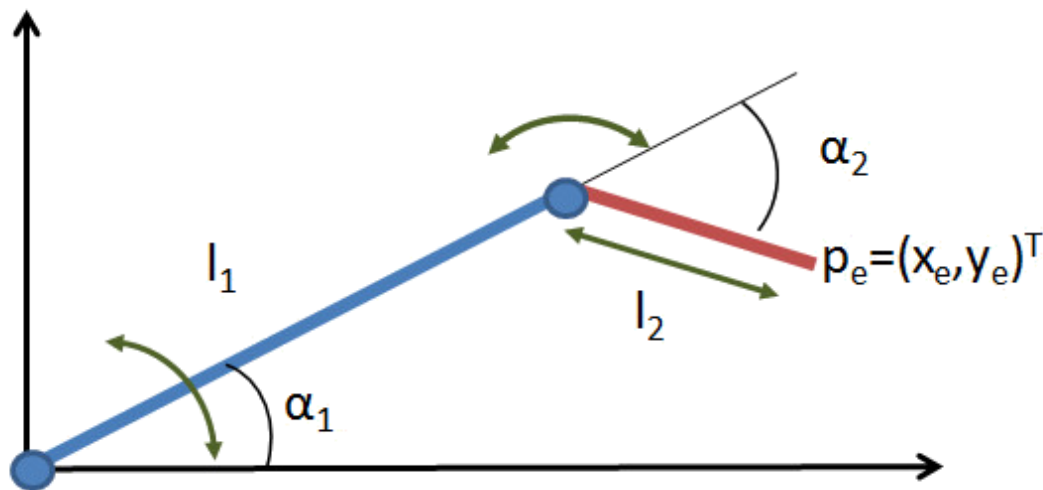
approximation with  $\delta = 0.1$

$$\begin{aligned} D_{\delta=0.1}f(1,2) &= \begin{pmatrix} \frac{f_1(1+\delta, 2) - f_1(1,2)}{\delta} & \frac{f_1(1,2+\delta) - f_1(1,2)}{\delta} \\ \frac{f_2(1+\delta, 2) - f_2(1,2)}{\delta} & \frac{f_2(1,2+\delta) - f_2(1,2)}{\delta} \end{pmatrix} \\ &= \begin{pmatrix} \frac{c(1.1) \cdot 2 - c(1) \cdot 2}{0.1} & \frac{c(1) \cdot 2.1 - c(1) \cdot 2}{0.1} \\ \frac{s(1.1) \cdot 2 - s(1) \cdot 2}{0.1} & \frac{s(1) \cdot 2.1 - s(1) \cdot 2}{0.1} \end{pmatrix} = \begin{pmatrix} -\mathbf{1.734} & \mathbf{0.540} \\ \mathbf{0.995} & \mathbf{0.841} \end{pmatrix} \end{aligned}$$



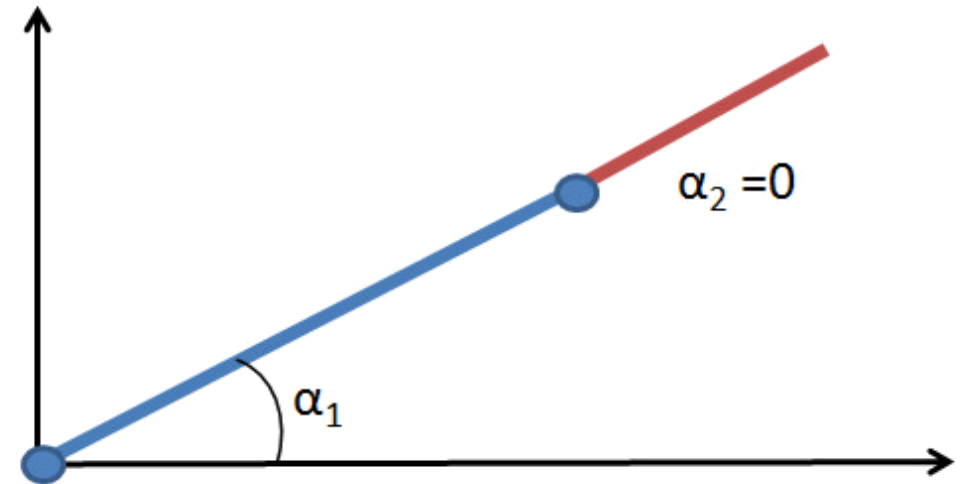
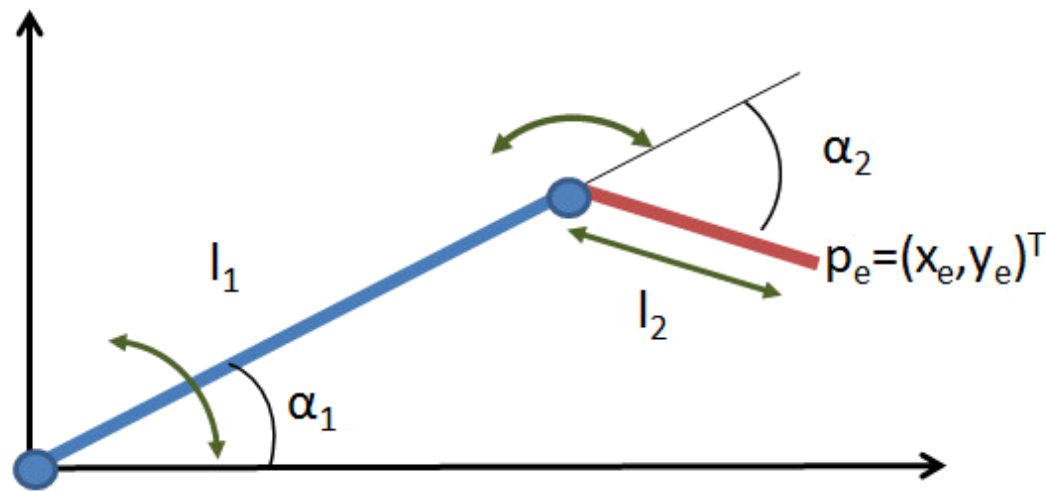
# Problem 2

Take the planar arm from **PS03, Problem 2** with the DoF  $\alpha_1$ ,  $\alpha_2$  and  $l_2$ . Use its FK to derive the related Jacobian matrix  $J$ .



# Problem 2

Forward Kinematics (PS03, problem 2)



$$K(\alpha_1, \alpha_2, l_2) = \begin{pmatrix} x_e \\ y_e \\ 1 \end{pmatrix} = \begin{pmatrix} c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10 \\ s\alpha_1 c\alpha_2 \cdot l_2 + c\alpha_1 s\alpha_2 \cdot l_2 + s\alpha_1 \cdot 10 \\ 1 \end{pmatrix}$$

# Problem 2

Jacobian of  $K(\alpha_1, \alpha_2, l_2)$

$$K(\alpha_1, \alpha_2, l_2) = \begin{pmatrix} c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10 \\ s\alpha_1 c\alpha_2 \cdot l_2 + c\alpha_1 s\alpha_2 \cdot l_2 + s\alpha_1 \cdot 10 \end{pmatrix}$$

$$J = DK(\alpha_1, \alpha_2, l_2) = \begin{pmatrix} \frac{\partial K_x}{\partial \alpha_1} & \frac{\partial K_x}{\partial \alpha_2} & \frac{\partial K_x}{\partial l_2} \\ \frac{\partial K_y}{\partial \alpha_1} & \frac{\partial K_y}{\partial \alpha_2} & \frac{\partial K_y}{\partial l_2} \end{pmatrix}$$

$$= \begin{pmatrix} \frac{c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10}{\partial \alpha_1} & \frac{c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10}{\partial \alpha_2} & \frac{c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10}{\partial l_2} \\ \frac{s\alpha_1 c\alpha_2 \cdot l_2 + c\alpha_1 s\alpha_2 \cdot l_2 + s\alpha_1 \cdot 10}{\partial \alpha_1} & \frac{s\alpha_1 c\alpha_2 \cdot l_2 + c\alpha_1 s\alpha_2 \cdot l_2 + s\alpha_1 \cdot 10}{\partial \alpha_2} & \frac{s\alpha_1 c\alpha_2 \cdot l_2 + c\alpha_1 s\alpha_2 \cdot l_2 + s\alpha_1 \cdot 10}{\partial l_2} \end{pmatrix}$$

$$= \begin{pmatrix} -s\alpha_1 c\alpha_2 \cdot l_2 - c\alpha_1 s\alpha_2 \cdot l_2 - s\alpha_1 \cdot 10 & -c\alpha_1 s\alpha_2 \cdot l_2 - s\alpha_1 c\alpha_2 \cdot l_2 & c\alpha_1 c\alpha_2 - s\alpha_1 s\alpha_2 \\ c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10 & -s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 c\alpha_2 \cdot l_2 & s\alpha_1 c\alpha_2 + c\alpha_1 s\alpha_2 \end{pmatrix}$$

# Problem 3

Take the Jacobian  $J$  from the previous problem 2. Which options do you know to compute the pseudo-inverse  $J^+$  of  $J$ , and when are they applicable?

# Problem 3

- option 1: closed form solution (formula for left or right  $A^+$ )
- option 2: numerical solution using SVD

# Problem 3 (Option 1)

$m \times n$  matrix  $A$ ,  $n \times m$  matrix  $A^+$

- $m > n$ , i.e.,  $A$  is “**tall**” :  
 $A^+ = (A^T A)^{-1} A^T$  aka **left pseudo-inverse**
- $n > m$ , i.e.,  $A$  is “**wide**” :  
 $A^+ = A^T (A A^T)^{-1}$  aka **right pseudo-inverse**

note:  $n = m \Rightarrow A^+ = A^{-1}$

$A$  “tall” :

$$\begin{pmatrix} A \end{pmatrix}$$

$A$  “wide” :

$$\begin{pmatrix} A \end{pmatrix}$$

# Problem 3 (Option 1)

here:  $K(): \mathbb{R}^3 \rightarrow \mathbb{R}^2$ ,  $K(\alpha_1, \alpha_2, l_2) = \begin{pmatrix} x_e \\ y_e \end{pmatrix}$

Jacobian  $\mathbf{DK}() = \mathbf{A}$

$\Rightarrow$   $\mathbf{A}$  is a  $2 \times 3$  matrix, i.e., “wide”

$\Rightarrow$  **right pseudo-inverse**  $\mathbf{A}\mathbf{A}^+ = \mathbf{I}$

$$\mathbf{A}^+ = \mathbf{A}^T (\mathbf{A}\mathbf{A}^T)^{-1}$$

# Problem 3 (Option 1)

$$\begin{aligned}
 DK(\alpha_1, \alpha_2, l_2)^+ &= DK(\alpha_1, \alpha_2, l_2)^T (DK(\alpha_1, \alpha_2, l_2) DK(\alpha_1, \alpha_2, l_2)^T)^{-1} \\
 &= \begin{pmatrix} -s\alpha_1 c\alpha_2 l_2 - c\alpha_1 s\alpha_2 l_2 - 10s\alpha_1 & c\alpha_1 c\alpha_2 l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + 10c\alpha_1 \\ -c\alpha_1 s\alpha_2 l_2 - s\alpha_1 c\alpha_2 l_2 & -s\alpha_1 s\alpha_2 l_2 + c\alpha_1 c\alpha_2 l_2 \\ c\alpha_1 c\alpha_2 - s\alpha_1 s\alpha_2 & s\alpha_1 c\alpha_2 + c\alpha_1 s\alpha_2 \end{pmatrix} \\
 &\cdot \begin{pmatrix} (-s\alpha_1 c\alpha_2 l_2 - c\alpha_1 s\alpha_2 l_2 - 10s\alpha_1 & -c\alpha_1 s\alpha_2 l_2 - s\alpha_1 c\alpha_2 l_2 & c\alpha_1 c\alpha_2 - s\alpha_1 s\alpha_2) \\ (c\alpha_1 c\alpha_2 l_2 - s\alpha_1 s\alpha_2 l_2 + 10c\alpha_1 & -s\alpha_1 s\alpha_2 l_2 + c\alpha_1 c\alpha_2 l_2 & s\alpha_1 c\alpha_2 + c\alpha_1 s\alpha_2) \end{pmatrix} \\
 &\cdot \begin{pmatrix} -s\alpha_1 c\alpha_2 l_2 - c\alpha_1 s\alpha_2 l_2 - 10s\alpha_1 & c\alpha_1 c\alpha_2 l_2 - s\alpha_1 s\alpha_2 l_2 + 10c\alpha_1 \\ -c\alpha_1 s\alpha_2 l_2 - s\alpha_1 c\alpha_2 l_2 & -s\alpha_1 s\alpha_2 l_2 + c\alpha_1 c\alpha_2 l_2 \\ c\alpha_1 c\alpha_2 - s\alpha_1 s\alpha_2 & s\alpha_1 c\alpha_2 + c\alpha_1 s\alpha_2 \end{pmatrix}^{-1}
 \end{aligned}$$

may be multiplied out, simplified, inverted, simplified...



# Problem 3 (Option 1)

$$DK(\alpha_1, \alpha_2, l_2)^+ = DK(\alpha_1, \alpha_2, l_2)^T (DK(\alpha_1, \alpha_2, l_2)DK(\alpha_1, \alpha_2, l_2)^T)^{-1}$$

works “always”, i.e., pseudo-inverse is a fct  $DK^+(\alpha_1, \alpha_2, l_2)$

- but tends to be computationally quite complex
- unless some effort spend to derive simpler form  
(multiply matrices out, use trigonometric laws, etc.)

# Problem 3 (Option 1)

option 1 with concrete values: step 1, the concrete Jacobian

$$DK(90^\circ, 0^\circ, 8)$$

$$\begin{aligned} &= \begin{pmatrix} -s90^\circ c0^\circ \cdot 8 - c90^\circ s0^\circ \cdot 8 - s90^\circ \cdot 10 & -c90^\circ s0^\circ \cdot 8 - s90^\circ c0^\circ \cdot 8 & c90^\circ c0^\circ - s90^\circ s0^\circ \\ c90^\circ c0^\circ \cdot 8 - s90^\circ s0^\circ \cdot 8 + c90^\circ \cdot 10 & -s90^\circ s0^\circ \cdot 8 + c90^\circ c0^\circ \cdot 8 & s90^\circ c0^\circ + c90^\circ s0^\circ \end{pmatrix} \\ &= \begin{pmatrix} -8 - 0 - 10 & 0 - 8 & 0 - 0 \\ 0 - 0 + 0 & 0 + 0 & 1 + 0 \end{pmatrix} \\ &= \begin{pmatrix} -18 & -8 & 0 \\ 0 & 0 & 1 \end{pmatrix} \end{aligned}$$

as mentioned,  $DK() = A$  is “wide”  $\Rightarrow A^+ = A^T(AA^T)^{-1}$

# Problem 3 (Option 1)

$$DK(90^\circ, 0^\circ, 8) = \begin{pmatrix} -18 & -8 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\text{wide } A \Rightarrow A^+ = A^T (AA^T)^{-1}$$

$$\begin{aligned} DK^+(90^\circ, 0^\circ, 8) &= \begin{pmatrix} -18 & 0 \\ -8 & 0 \\ 0 & 1 \end{pmatrix} \left( \begin{pmatrix} -18 & -8 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} -18 & 0 \\ -8 & 0 \\ 0 & 1 \end{pmatrix} \right)^{-1} \\ &= \begin{pmatrix} -18 & 0 \\ -8 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 388 & 0 \\ 0 & 1 \end{pmatrix}^{-1} = \begin{pmatrix} -18 & 0 \\ -8 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \frac{1}{388} & 0 \\ 0 & 1 \end{pmatrix} \\ &= \begin{pmatrix} -0.04644 & 0 \\ -0.02064 & 0 \\ 0 & 1 \end{pmatrix} \end{aligned}$$

closed form pseudo-inverse of the Jacobian  
with concrete values for input DoF

# Problem 3 (Option 2)

## singular value decomposition (SVD)

can only be used when fixed DoF values are given

$$DK(\alpha_1, \alpha_2, l_2) = \begin{pmatrix} -18 & -8 & 0 \\ 0 & 0 & 1 \end{pmatrix} = UWV^T$$

$$U = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, W = \begin{pmatrix} 19.698 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, V^T = \begin{pmatrix} -0.914 & -0.406 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

# Problem 3 (Option 2)

$$U = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, W = \begin{pmatrix} 19.698 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, V^T = \begin{pmatrix} -0.914 & -0.406 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

$$DK^+(90^\circ, 0^\circ, 8) = VW^+U^T$$

$$= \begin{pmatrix} -0.914 & -0.406 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 19.698 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}^+ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}^T$$

$$= \begin{pmatrix} -0.914 & 0 & 0 \\ -0.406 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0.05077 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} -0.04640 & 0 \\ -0.02061 & 0 \\ 0 & 1 \end{pmatrix}$$

# Problem 3 (Option 2)

note:  $DK^+(90^\circ, 0^\circ, 8) = VW^+U^T$  with  $W^+$

- transpose of  $W$ , i.e.,  $W_{ij}^+ = W_{ji}$ , and
- reciprocals of non-Zero diagonal entries, i.e.,  $W_{ii}^+ = 1/W_{ii}$  if  $W_{ii} \neq 0$

$$W^+ = \begin{pmatrix} 19.698 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}^+ = \begin{pmatrix} \frac{1}{19.698} & 0 & 0 \\ 0 & \frac{1}{1} & 0 \end{pmatrix}^T = \begin{pmatrix} 0.05077 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix}$$

# Problem 3: Pseudo-Inverse

option 1: closed form

$$A^+ = A^T (AA^T)^{-1}$$
$$DK^+(q) = \begin{pmatrix} -0.04644 & 0 \\ -0.02064 & 0 \\ 0 & 1 \end{pmatrix}$$

works “always”, i.e.,  
pseudo-inverse can be derived as  
fct  $DK^+(q)$

option 2: SVD

$$A^+ = VS^+U^T$$
$$DK^+(q) = \begin{pmatrix} -0.04640 & 0 \\ -0.02061 & 0 \\ 0 & 1 \end{pmatrix}$$

can be used when fixed DoF  
values are given

(note: both examples done with some rounding in the calculations)

# Problem 4

Take the arm, its FK, and the related pseudo-inverse of the Jacobian  $J^+$  from the previous problems 2 and 3.

Given the goal position  $p_t = (5, 10)^T$  and the starting DoF values  $\alpha_1(0) = 90^\circ$ ,  $\alpha_2(0) = 0^\circ$ ,  $l_2(0) = 8$ .

Formulate the numerical IK with

- a) Newton's method
- b) Gradient descent



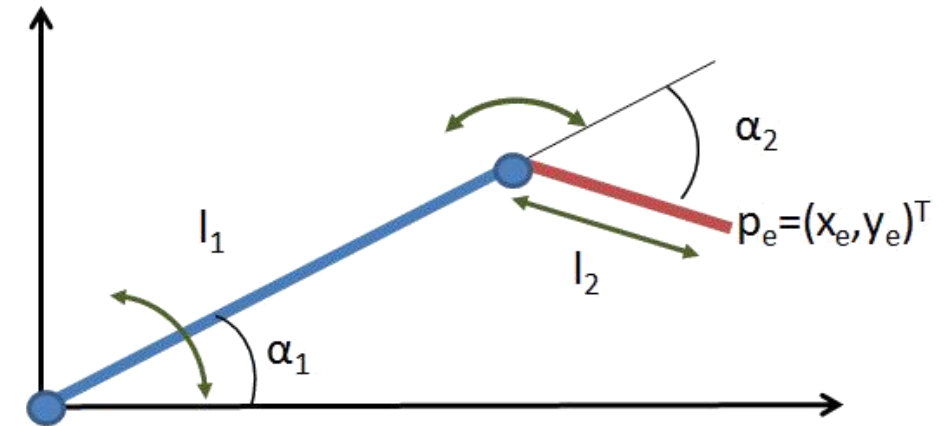
# Problem 4

- a) Newton's method:  $q_{k+1} = q_k + \alpha J(q_k)^+ (p_t - K(q_k))$   
b) Gradient descent:  $q_{k+1} = q_k + \alpha J(q_k)^T (p_t - K(q_k))$

note: yet an other “clear from context”

- dof, i.e.,  $q$ , in kinematics as a column vector
- $q$  as “input” to fct  $K()$  as tuple, i.e., like row vector

$$q = (\alpha_1, \alpha_2, l_2)^T$$



$$K(q) = K(\alpha_1, \alpha_2, l_2) = \begin{pmatrix} x_e \\ y_e \\ 1 \end{pmatrix} = \begin{pmatrix} c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10 \\ s\alpha_1 c\alpha_2 \cdot l_2 + c\alpha_1 s\alpha_2 \cdot l_2 + s\alpha_1 \cdot 10 \\ 1 \end{pmatrix}$$

# Problem 4

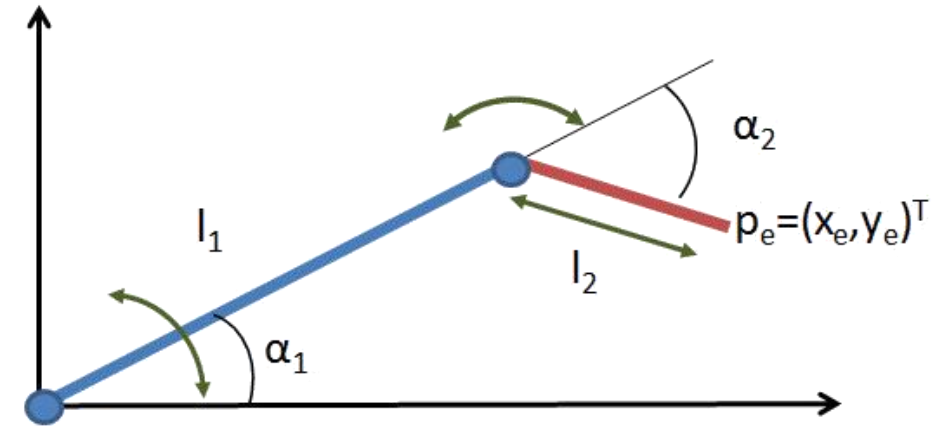
- a) Newton's method:  $q_{k+1} = q_k + \alpha J(q_k)^+ (p_t - K(q_k))$   
 b) Gradient descent:  $q_{k+1} = q_k + \alpha J(q_k)^T (p_t - K(q_k))$

see problem 2

$$K(q) = \begin{pmatrix} c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10 \\ s\alpha_1 c\alpha_2 \cdot l_2 + c\alpha_1 s\alpha_2 \cdot l_2 + s\alpha_1 \cdot 10 \end{pmatrix}$$

$$J = DK(q)$$

$$= \begin{pmatrix} -s\alpha_1 c\alpha_2 \cdot l_2 - c\alpha_1 s\alpha_2 \cdot l_2 - s\alpha_1 \cdot 10 & -c\alpha_1 s\alpha_2 \cdot l_2 - s\alpha_1 c\alpha_2 \cdot l_2 & c\alpha_1 c\alpha_2 - s\alpha_1 s\alpha_2 \\ c\alpha_1 c\alpha_2 \cdot l_2 - s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 \cdot 10 & -s\alpha_1 s\alpha_2 \cdot l_2 + c\alpha_1 c\alpha_2 \cdot l_2 & s\alpha_1 c\alpha_2 + c\alpha_1 s\alpha_2 \end{pmatrix}$$



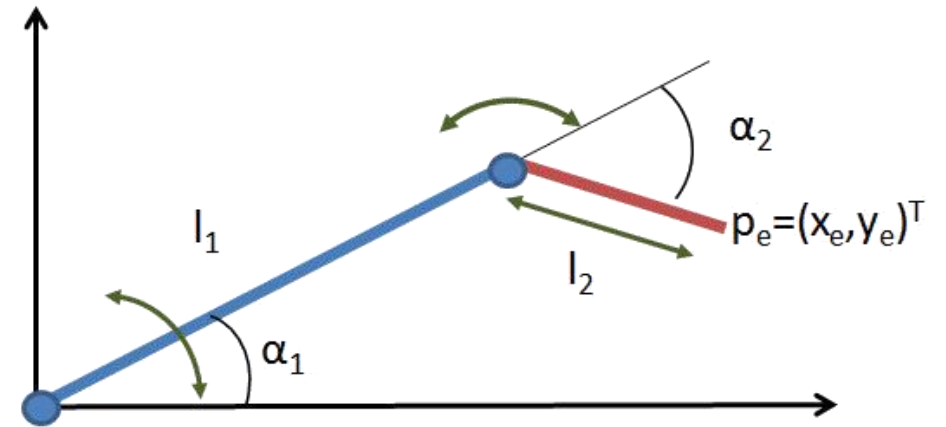
# Problem 4: Newton-Raphson

note: here () as notation for the time-steps instead of subscripts

$$\begin{aligned} q(k+1) &= q(k) + \alpha J(q(k))^+ (p_t - K(q(k))) \\ &= q(k) + \alpha \Delta q \text{ with } \Delta q = J(q(k))^+ (p_t - K(q(k))) \end{aligned}$$

start:  $k = 0$

$$\begin{aligned} q(0) &= (\alpha_1(0), \alpha_2(0), l_2(0))^T \\ &= (90^\circ, 0^\circ, 8)^T \end{aligned}$$



# Problem 4: Newton-Raphson

$$q(0) = (90^\circ, 0^\circ, 8)^T$$

$$q(1) = q(0) + \alpha \Delta q \text{ with } \Delta q = J(q(0))^+ (\boxed{p_t} - \boxed{K(q(0))})$$

hand target:  $\boxed{p_t} = \begin{pmatrix} 5 \\ 10 \end{pmatrix}$

hand at  $k = 0$ :

$$\boxed{K(q(0))} = K(90^\circ, 0^\circ, 8)$$

$$= \begin{pmatrix} c(90^\circ)c(0^\circ) \cdot 8 - s(90^\circ)s(0^\circ) \cdot 8 + c(90^\circ) \cdot 10 \\ s(90^\circ)c(0^\circ) \cdot 8 + c(90^\circ)s(0^\circ) \cdot 8 + s(90^\circ) \cdot 10 \end{pmatrix}$$

$$= \begin{pmatrix} 0 - 0 + 0 \\ 8 + 0 + 10 \end{pmatrix} = \boxed{\begin{pmatrix} 0 \\ 18 \end{pmatrix}}$$

# Problem 4: Newton-Raphson

$$q(0) = (90^\circ, 0^\circ, 8)^T$$

$$q(1) = q(0) + \alpha \Delta q \text{ with}$$

$$\Delta q = J(q(0))^+ \left( \begin{pmatrix} 5 \\ 10 \end{pmatrix} - \begin{pmatrix} 0 \\ 18 \end{pmatrix} \right) = J(q(0))^+ \begin{pmatrix} 5 \\ -8 \end{pmatrix}$$

pseudo-inverse of the Jacobian at  $k = 0$ :

$$J(q(0))^+ = DK^+(q(0)) = \begin{pmatrix} -0.04644 & 0 \\ -0.02064 & 0 \\ 0 & 1 \end{pmatrix} \quad \text{problem 3}$$

# Problem 4: Newton-Raphson

$$q(0) = (90^\circ, 0^\circ, 8)^T$$

$$q(1) = q(0) + \alpha \Delta q \text{ with } J(q(0))^+ (p_t - K(q(0)))$$

$$\Delta q = \begin{pmatrix} -0.04644 & 0 \\ -0.02064 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 5 \\ -8 \end{pmatrix} = \begin{pmatrix} -0.23220 \\ -0.10320 \\ 8 \end{pmatrix}$$

note: the  
angular changes  
are in radians

$$\text{hence, } \Delta q = \begin{pmatrix} -0.23220 \\ -0.10320 \\ 8 \end{pmatrix} = \begin{pmatrix} -0.23220/\pi \cdot 180^\circ \\ -0.10320/\pi \cdot 180^\circ \\ 8 \end{pmatrix} = \begin{pmatrix} -13.30^\circ \\ -5.91^\circ \\ 8 \end{pmatrix}$$

# Problem 4: Newton-Raphson

$$q(0) = (90^{\circ}, 0^{\circ}, 8)$$

$$q(1) = q(0) + \alpha \Delta q \text{ with } J(q(0))^+ \left( p_t - K(q(0)) \right)$$

e.g.,  $\alpha = 0.1$ :

$$\begin{aligned} q(1) &= \begin{pmatrix} 90^{\circ} \\ 0^{\circ} \\ 8 \end{pmatrix} + 0.1 \cdot \begin{pmatrix} -13.30^{\circ} \\ -5.91^{\circ} \\ 8 \end{pmatrix} = \begin{pmatrix} 90^{\circ} \\ 0^{\circ} \\ 8 \end{pmatrix} + \begin{pmatrix} -1.330^{\circ} \\ -0.591^{\circ} \\ 0.8 \end{pmatrix} \\ &= (88.670^{\circ} \quad 359.319^{\circ} \quad 8.8)^T \end{aligned}$$

# Problem 4: Newton-Raphson

keep on iterating:

- compute forward kinematics  $K(q(1))$  of  $q(1)$
- next Jacobian  $J(q(1))$  at  $q(1)$
- next pseudo-inverse  $J^+(q(1))$
- get  $q(2) = q(1) + \alpha J(q(1))^+ (p_t - K(q(1)))$
- and so on...

until small error to target, i.e.,  $|p_t - K(q(n))| < \varepsilon$



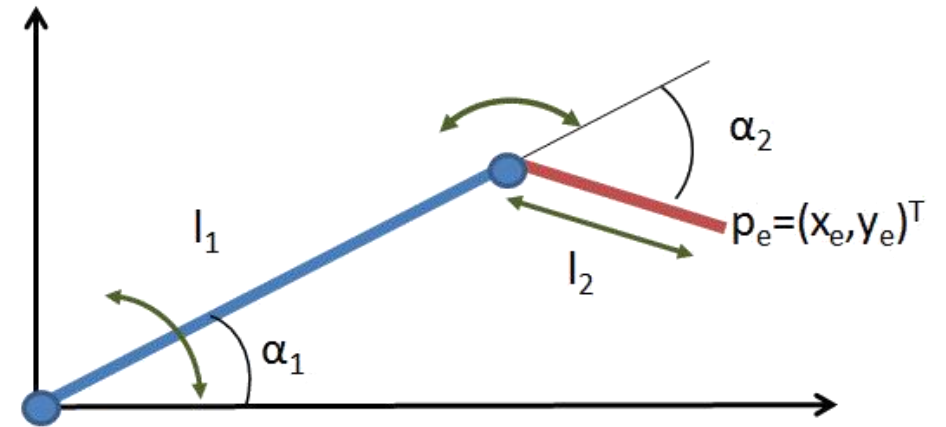
# Problem 4: Gradient Descent

note: also here () for the time-steps

$$\begin{aligned} q(k+1) &= q(k) + \alpha J(q(k))^T (p_t - K(q(k))) \\ &= q(k) + \alpha \Delta q \text{ with } \Delta q = J(q(k))^T (p_t - K(q(k))) \end{aligned}$$

start:  $k = 0$

$$\begin{aligned} q(0) &= (\alpha_1(0), \alpha_2(0), l_2(0))^T \\ &= (90^\circ, 0^\circ, 8)^T \end{aligned}$$



# Problem 4: Gradient Descent

$$q(0) = (90^\circ, 0^\circ, 8)^T$$

$$q(1) = q(0) + \alpha \Delta q \text{ with } \Delta q = J(q(0))^+ (\boxed{p_t} - \boxed{K(q(0))})$$

this part is  
exactly like  
Newton

hand target:

$$\boxed{p_t} = \begin{pmatrix} 5 \\ 10 \end{pmatrix}$$

hand at  $k = 0$ :

$$\boxed{K(q(0))} = K(90^\circ, 0^\circ, 8)$$

$$= \begin{pmatrix} c(90^\circ)c(0^\circ) \cdot 8 - s(90^\circ)s(0^\circ) \cdot 8 + c(90^\circ) \cdot 10 \\ s(90^\circ)c(0^\circ) \cdot 8 + c(90^\circ)s(0^\circ) \cdot 8 + s(90^\circ) \cdot 10 \end{pmatrix}$$

$$= \begin{pmatrix} 0 - 0 + 0 \\ 8 + 0 + 10 \end{pmatrix} = \boxed{\begin{pmatrix} 0 \\ 18 \end{pmatrix}}$$

# Problem 4: Gradient Descent

$$q(0) = (90^\circ, 0^\circ, 8)^T$$

$$q(1) = q(0) + \alpha \Delta q \text{ with}$$

$$\Delta q = J(q(0))^T \left( \begin{pmatrix} 5 \\ 10 \end{pmatrix} - \begin{pmatrix} 0 \\ 18 \end{pmatrix} \right) = J(q(0))^T \begin{pmatrix} 5 \\ -8 \end{pmatrix}$$

**transpose** of the Jacobian at  $k = 0$ :

Newton needs  
pseudo-inverse

$$J(q(0))^T = DK^T(q) = \begin{pmatrix} -18 & 0 \\ -8 & 0 \\ 0 & 1 \end{pmatrix}$$

# Problem 4: Gradient Descent

$$q(0) = (90^{\circ}, 0^{\circ}, 8)$$

$$q(1) = q(0) + \alpha \Delta q \text{ with } J(q(0))^T (p_t - K(q(0)))$$

e.g.,  $\alpha = 0.1$ :

$$\begin{aligned} q(1) &= \begin{pmatrix} 90^{\circ} \\ 0^{\circ} \\ 8 \end{pmatrix} + 0.01 \cdot \begin{pmatrix} -18 & 0 \\ -8 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 5 \\ -8 \end{pmatrix} = \begin{pmatrix} 90^{\circ} \\ 0^{\circ} \\ 8 \end{pmatrix} + \begin{pmatrix} -0.9 \cdot \frac{180^{\circ}}{\pi} \\ -0.4 \cdot \frac{180^{\circ}}{\pi} \\ -0.08 \end{pmatrix} \\ &= \begin{pmatrix} 90^{\circ} \\ 0^{\circ} \\ 8 \end{pmatrix} + \begin{pmatrix} -51.57^{\circ} \\ -22.92^{\circ} \\ -0.08 \end{pmatrix} = \begin{pmatrix} 38.43^{\circ} \\ 337.08^{\circ} \\ 7.92 \end{pmatrix} \end{aligned}$$

# Problem 4: Gradient Descent

keep on iterating:

- compute forward kinematics  $K(q(1))$  of  $q(1)$
- next Jacobian  $J(q(1))$  at  $q(1)$
- next transpose  $J^T(q(1))$
- get  $q(2) = q(1) + \alpha J(q(1))^T (p_t - K(q(1)))$
- and so on...

until small error to target, i.e.,  $|p_t - K(q(n))| < \varepsilon$

# Problem 4

IK with

a) Newton's method:  $q_{k+1} = q_k + \alpha J(q_k)^+ (p_t - K(q_k))$

b) Gradient descent:  $q_{k+1} = q_k + \alpha J(q_k)^T (p_t - K(q_k))$

in general,

a) Newton's method:  $x_{k+1} = x_k - \alpha J_F(x_k)^+ F(x_k)$

b) Gradient descent:  $x_{k+1} = x_k - \alpha \nabla F(x_k)$  with  $\nabla F(x_k) = J_F(x_k)^T$

***why once “+” and once “-”?***

# Problem 4

general:  $x_{k+1} = x_k - \alpha J_F(x_k)^{+,T} F(x_k)$

IK:  $q_{k+1} = q_k + \alpha J(q_k)^{+,T} (p_t - K(q_k))$

$$F(x_k) \leftrightarrow F(q_k) = (p_t - K(q_k))$$

$$J_F(x_k) \leftrightarrow J_F(q_k) = DF(q_k)$$

$$\begin{aligned} &= \frac{\delta(p_t - K(q_k))}{\delta q_k} \quad (p_t \text{ is constant}) \\ &= -J_K(q_k) \end{aligned}$$

# Problem 4

IK with

a) Newton's method:  $q_{k+1} = q_k + \alpha J(q_k)^+ (p_t - K(q_k))$

b) Gradient descent:  $q_{k+1} = q_k + \alpha J(q_k)^T (p_t - K(q_k))$

in general,

a) Newton's method:  $x_{k+1} = x_k - \alpha J_F(x_k)^+ F(x_k)$

b) Gradient descent:  $x_{k+1} = x_k - \alpha \nabla F(x_k)$  with  $\nabla F(x_k) = J_F(x_k)^T$

***which is better?***



# Problem 4

just as a very rough guideline

- **Newton Raphson**

$$\mathbf{x}_{k+1} = \mathbf{x}_k - \alpha \mathbf{J}_F(\mathbf{x}_k)^+ F(\mathbf{x}_k)$$

- finds the **root** of a multivariate, **vector-valued function**
- i.e.,  $F(\hat{\mathbf{x}}) = \mathbf{0}$  for  $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$
- it is more complex per step (needs pseudo-inverse) but it can converge faster

- **gradient descent**

$$\mathbf{x}_{k+1} = \mathbf{x}_k - \alpha \nabla F(\mathbf{x}_k)$$

- finds the **minimum** of a multivariate, **real-valued function**
- i.e.,  $\hat{\mathbf{x}} = \min_x F(\mathbf{x})$  for  $F : \mathbb{R}^n \rightarrow \mathbb{R}$
- it is simpler (just needs  $J$  transpose) but smaller steps towards the minimum

note: and there are also other methods for numerical optimization... (and for IK)