

Ch6: Inverse Kinematics

Inverse Kinematics

Inverse Kinematics

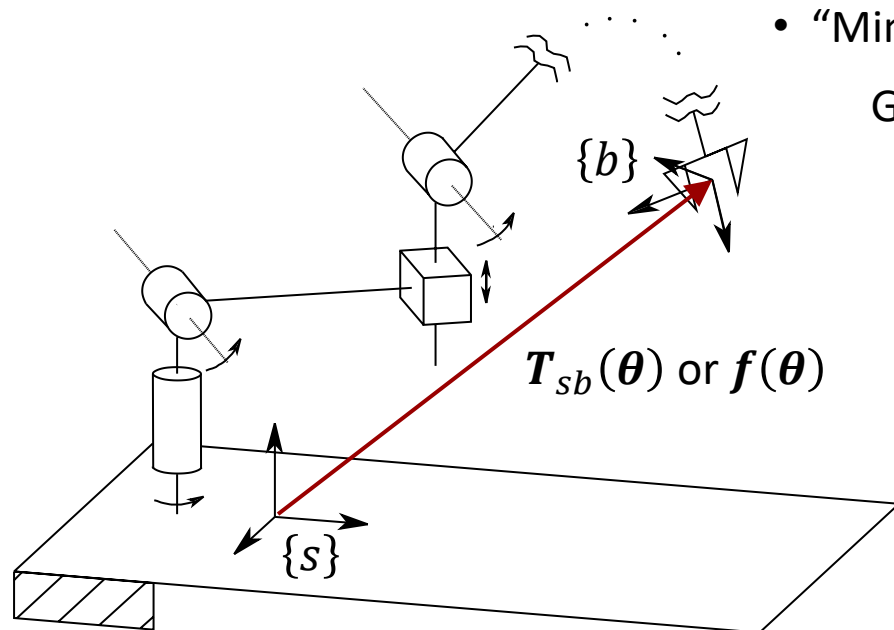
The inverse kinematics of a robot refers to the calculation of the joint coordinates θ from the position and orientation (**pose**) of its end-effector frame.

- “Geometric” inverse kinematics:

Given $T_{sb}(\theta) = T(\theta) \in SE(3)$, Find $\theta \in \mathbb{R}^n$

- “Minimum-Coordinate” inverse kinematics:

Given $x = f(\theta) \in \mathbb{R}^m$, Find $\theta \in \mathbb{R}^n$



Complexities of Inverse Kinematics

- The equations to solve are in general nonlinear, and thus it is not always possible to find a closed-form solution.
- Multiple solutions may exist.
- Infinite solutions may exist (e.g., in the case of a kinematically redundant manipulator).
- There might be no admissible solutions (e.g., when the given EE pose does not belong to the manipulator dexterous workspace.).

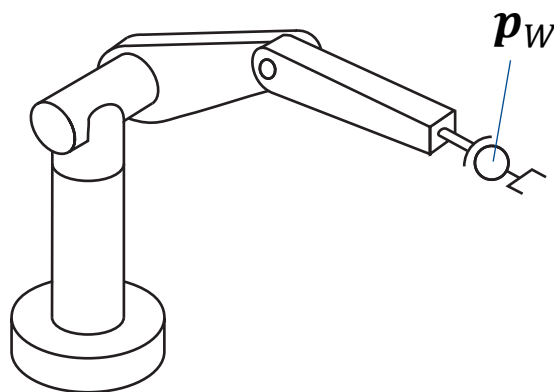
► Solving Inverse Kinematics Problems:

- **Analytic Methods:** Finding closed-form solutions using algebraic intuition or geometric intuition.
- **Iterative Numerical Methods:** When there are no (or it is difficult to find) closed-form solutions.

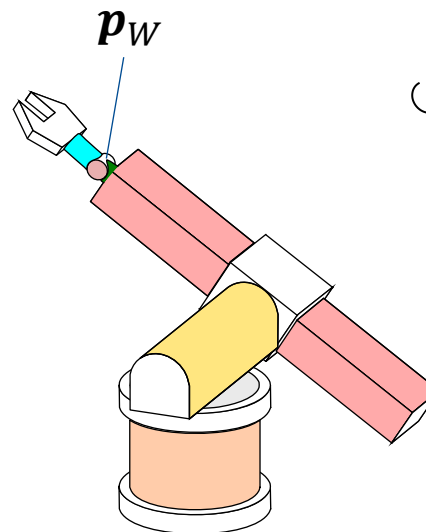
Analytic Inverse Kinematics

Analytic Inverse Kinematics

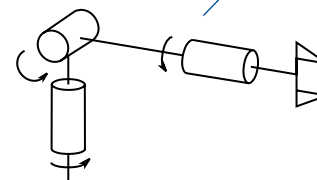
Most of the existing manipulators are typically formed by an **arm** and a **spherical wrist** (i.e., three consecutive revolute joint axes intersect at a common point). Thus, we can decouple the solution for the position (e.g., point p_W at the intersection of the three revolute axes) from that for the orientation.



PUMA Arm (6R)



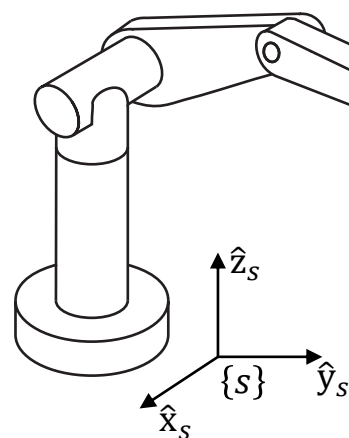
Stanford Arm (RRPRRR)



3R Planar Arm

* Therefore, it is possible to solve the inverse kinematics for the arm separately from the inverse kinematics for the spherical wrist.

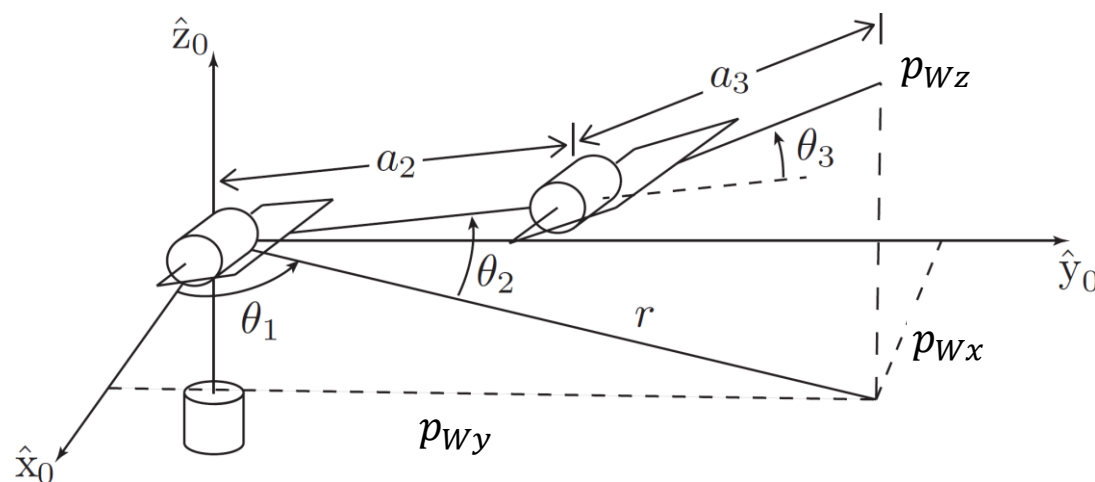
6R PUMA-Type Arms



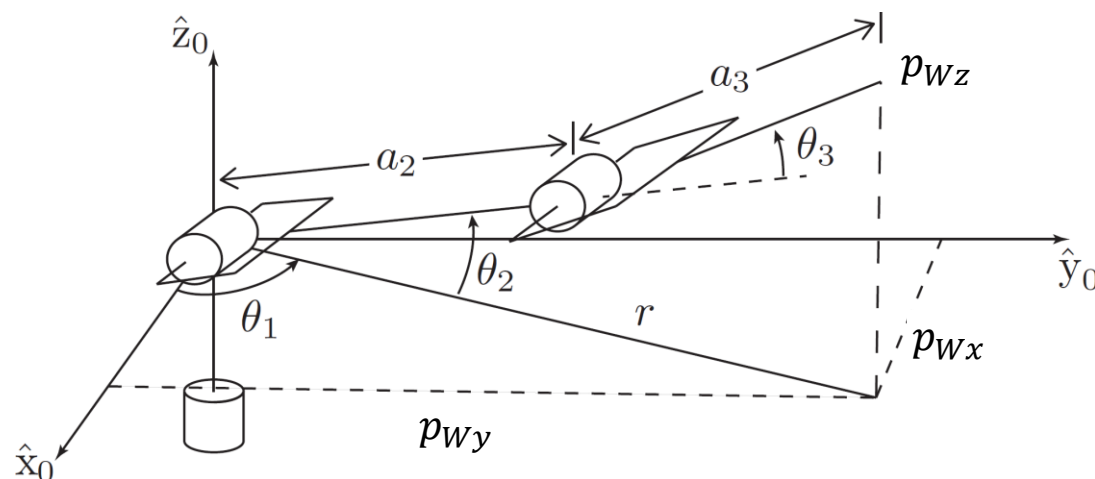
Wrist joints intersect orthogonally at a common point \mathbf{p}_W

$$\mathbf{T}(\boldsymbol{\theta}) = \begin{bmatrix} \mathbf{R}_{sb} & \mathbf{p}_b \\ \mathbf{0} & 1 \end{bmatrix} = e^{[s_1]\theta_1} e^{[s_2]\theta_2} e^{[s_3]\theta_3} e^{[s_4]\theta_4} e^{[s_5]\theta_5} e^{[s_6]\theta_6} \mathbf{M}$$

By having \mathbf{p}_b , we can find $\mathbf{p}_W = (p_{Wx}, p_{Wy}, p_{Wz})$



6R PUMA-Type Arms (cont.)



$$p_{Wx} = c_1(a_2c_2 + a_3c_{23})$$

$$p_{Wy} = s_1(a_2c_2 + a_3c_{23})$$

$$p_{Wz} = a_2s_2 + a_3s_{23}$$

❖ Inverse position problem of finding $(\theta_1, \theta_2, \theta_3)$ using algebraic intuition:

$$p_{Wx}^2 + p_{Wy}^2 + p_{Wz}^2 = a_2^2 + a_3^2 + 2a_2a_3c_3$$

$$c_3 = \frac{p_{Wx}^2 + p_{Wy}^2 + p_{Wz}^2 - a_2^2 - a_3^2}{2a_2a_3}$$

$$s_3 = \pm \sqrt{1 - c_3^2}$$



$$\theta_3 = \text{atan2}(s_3, c_3)$$



$$\begin{aligned} \theta_{3,I} &\in [-\pi, \pi] \\ \theta_{3,II} &= -\theta_{3,I} \end{aligned}$$

6R PUMA-Type Arms (cont.)

$$p_{Wx}^2 + p_{Wy}^2 = (a_2 c_2 + a_3 c_{23})^2 \longrightarrow \left. \begin{aligned} a_2 c_2 + a_3 c_{23} &= \pm \sqrt{p_{Wx}^2 + p_{Wy}^2} \\ p_{Wz} &= a_2 s_2 + a_3 s_{23} \\ s_{23} &= s_2 c_3 + s_3 c_2 \\ c_{23} &= c_2 c_3 - s_2 s_3 \end{aligned} \right\} \Rightarrow$$

$$c_2 = \frac{\pm \sqrt{p_{Wx}^2 + p_{Wy}^2} (a_2 + a_3 c_3) + p_{Wz} a_3 s_3}{a_2^2 + a_3^2 + 2a_2 a_3 c_3}$$

$$s_2 = \frac{p_{Wz} (a_2 + a_3 c_3) \mp \sqrt{p_{Wx}^2 + p_{Wy}^2} a_3 s_3}{a_2^2 + a_3^2 + 2a_2 a_3 c_3}$$

$$\Rightarrow \theta_2 = \text{atan2}(s_2, c_2)$$

For each θ_3 , we have two solutions for θ_2 :

$$\theta_{3,I} \rightarrow (\theta_{2,I}, \theta_{2,II})$$

$$\theta_{3,II} \rightarrow (\theta_{2,III}, \theta_{2,IV})$$

6R PUMA-Type Arms (cont.)

$$p_{Wx} = c_1(a_2c_2 + a_3c_{23})$$

$$p_{Wy} = s_1(a_2c_2 + a_3c_{23})$$

$$a_2c_2 + a_3c_{23} = \pm \sqrt{p_{Wx}^2 + p_{Wy}^2}$$



$$p_{Wx} = \pm c_1 \sqrt{p_{Wx}^2 + p_{Wy}^2}$$

$$p_{Wy} = \pm s_1 \sqrt{p_{Wx}^2 + p_{Wy}^2}$$



$$\theta_{1,I} = \text{atan2}(p_{Wy}, p_{Wx})$$

$$\theta_{1,II} = \text{atan2}(-p_{Wy}, -p_{Wx})$$

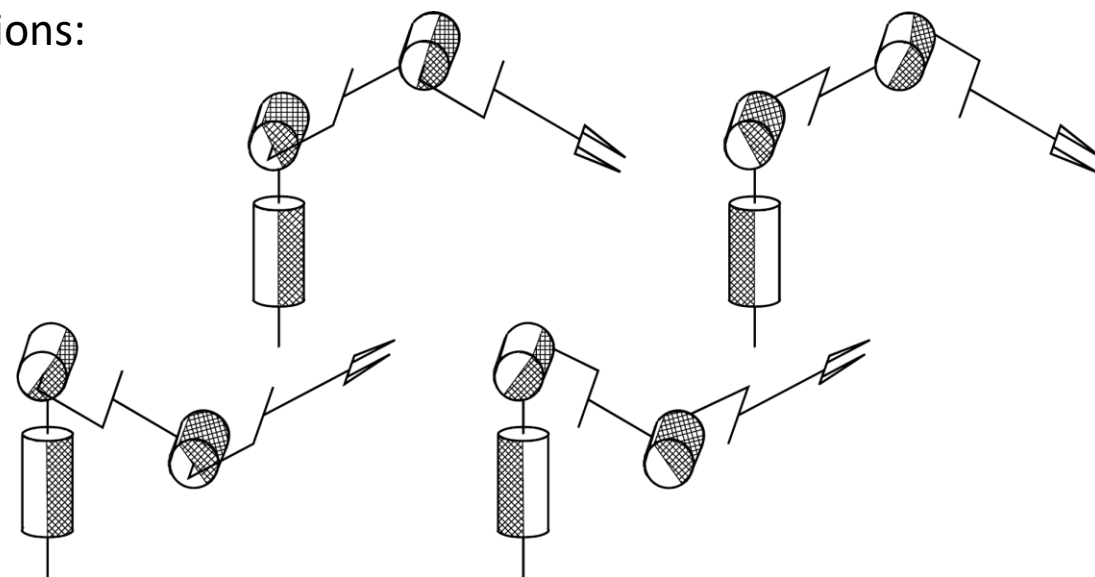
Thus, in total, there exist four solutions:

$$(\theta_{1,I}, \theta_{2,I}, \theta_{3,I})$$

$$(\theta_{1,I}, \theta_{2,III}, \theta_{3,II})$$

$$(\theta_{1,II}, \theta_{2,II}, \theta_{3,I})$$

$$(\theta_{1,II}, \theta_{2,IV}, \theta_{3,II})$$



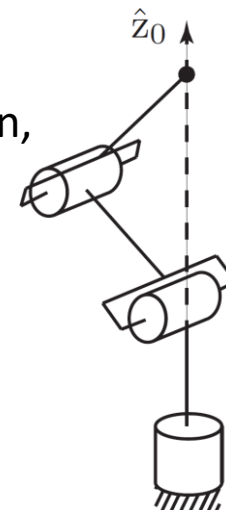
6R PUMA-Type Arms (cont.)

Note: When $p_{Wx} = p_{Wy} = 0$, the arm is in a kinematically singular configuration, and there are infinitely many possible solutions for θ_1 .

❖ Inverse orientation problem of finding $(\theta_4, \theta_5, \theta_6)$:

$$e^{[S_4]\theta_4} e^{[S_5]\theta_5} e^{[S_6]\theta_6} = e^{-[S_3]\theta_3} e^{-[S_2]\theta_2} e^{-[S_1]\theta_1} \mathbf{T}(\boldsymbol{\theta}) \mathbf{M}^{-1} = \mathbf{T}'$$

known

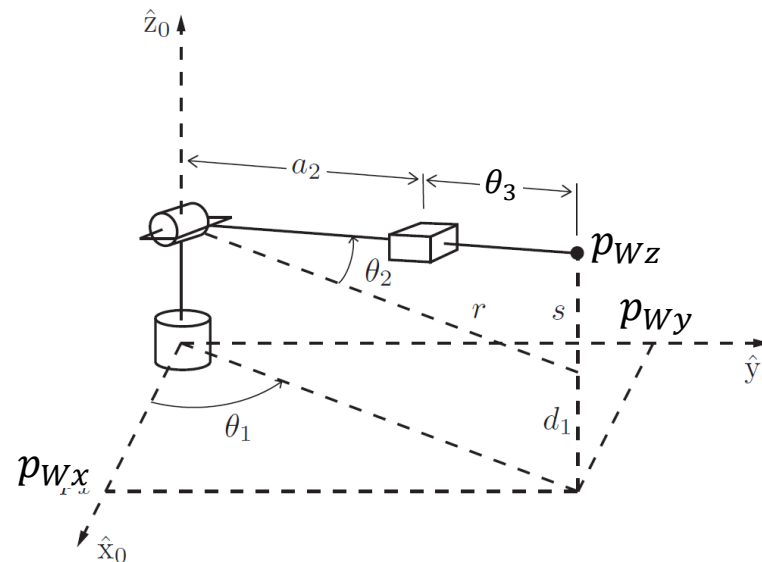
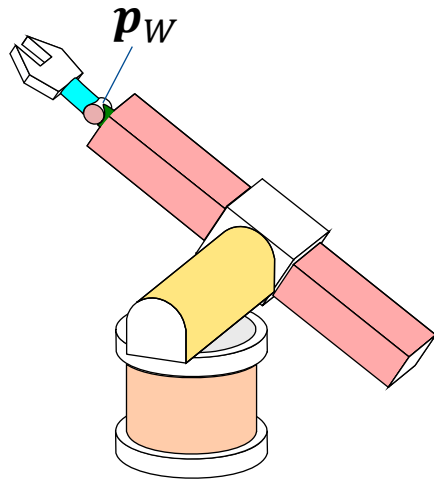


Assume that the joint axes (S_4, S_5, S_6) of the spherical wrist are aligned in the $(\hat{z}_s, \hat{y}_s, \hat{x}_s)$ directions, respectively:

$$\begin{aligned} S_{\omega_4} &= (0, 0, 1) \\ S_{\omega_5} &= (0, 1, 0) \\ S_{\omega_6} &= (1, 0, 0) \end{aligned} \quad \Rightarrow \quad \text{Rot}(\hat{z}, \theta_4) \text{Rot}(\hat{y}, \theta_5) \text{Rot}(\hat{x}, \theta_6) = \mathbf{R}' \quad \Rightarrow \quad \begin{aligned} &\text{This corresponds to} \\ &\text{the ZYX Euler angles.} \end{aligned}$$

$$\mathbf{T}' = (\mathbf{R}', \mathbf{p}') \quad \Rightarrow \quad (\theta_4, \theta_5, \theta_6)$$

Stanford-Type Arms



$$r^2 = p_{Wx}^2 + p_{Wy}^2$$

$$s = p_{Wz} - d_1$$

❖ Inverse position problem of finding $(\theta_1, \theta_2, \theta_3)$ using geometric intuition:

$$\text{If } p_{Wx}, p_{Wy} \neq 0: \begin{cases} \theta_1 = \text{atan2}(p_{Wy}, p_{Wx}) \\ \theta_2 = \text{atan2}(s, r) \end{cases}, \quad \begin{cases} \theta_1 = \pi + \text{atan2}(p_{Wy}, p_{Wx}) \\ \theta_2 = \pi - \text{atan2}(s, r) \end{cases}$$

$$(\theta_3 + a_2)^2 = r^2 + s^2 \quad \longrightarrow \quad \theta_3 = \sqrt{r^2 + s^2} = \sqrt{p_{Wx}^2 + p_{Wy}^2 + (p_{Wz} - d_1)^2} - a_2$$

\Rightarrow Thus, there are 2 solutions to the inverse kinematics problem.

❖ Inverse orientation problem of finding $(\theta_4, \theta_5, \theta_6)$ is similar to PUMA.

Numerical Inverse Kinematics

Newton–Raphson Method

Newton–Raphson Method is an iterative method for numerically finding the roots of a nonlinear equation $f(\theta) = 0$ where $f: \mathbb{R} \rightarrow \mathbb{R}$ is differentiable.

If θ^0 is an initial guess for the solution, Taylor expansion of $f(\theta)$ at θ^0 is

$$f(\theta) = f(\theta^0) + \frac{\partial f}{\partial \theta}(\theta^0)(\theta - \theta^0) + \text{higher-order terms (h.o.t)} \xrightarrow{f(\theta) = 0} \theta = \theta^0 - \left(\frac{\partial f}{\partial \theta}(\theta^0) \right)^{-1} f(\theta^0)$$

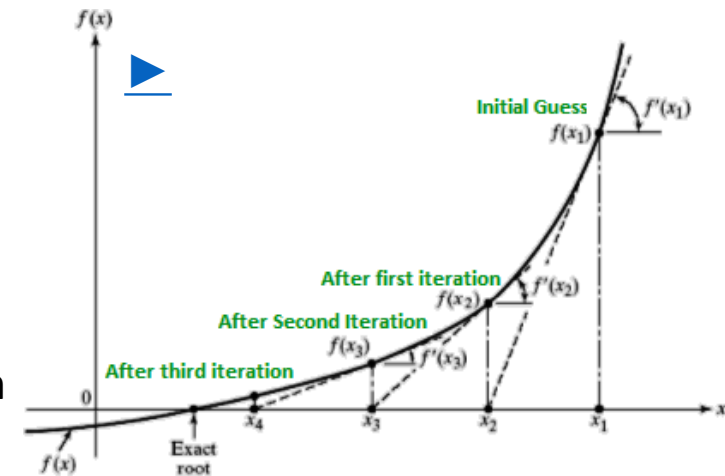
Using θ as the new guess for the solution and repeating:

$$\theta^{k+1} = \theta^k - \left(\frac{\partial f}{\partial \theta}(\theta^k) \right)^{-1} f(\theta^k)$$

The iteration is repeated until some stopping criterion is satisfied:

$$\frac{|f(\theta^k) - f(\theta^{k+1})|}{|f(\theta^k)|} \leq \epsilon$$

ϵ : a given threshold value



Inverse Kinematics Based on Newton–Raphson Method (Minimum-Coordinate IK)

- ❖ If the desired end-effector configuration represented by the minimum number of coordinates, i.e., $\mathbf{x}_d = \mathbf{f}(\boldsymbol{\theta}) \in \mathbb{R}^m$, $\boldsymbol{\theta} \in \mathbb{R}^n$, then the goal is to find joint coordinates $\boldsymbol{\theta}_d$ such that

$$\mathbf{x}_d - \mathbf{f}(\boldsymbol{\theta}_d) = \mathbf{0} \quad (\text{Assumption: } \mathbf{f} \text{ is differentiable})$$

- We use a method similar to the Newton–Raphson method for nonlinear root-finding:
Given an initial guess $\boldsymbol{\theta}^0$ which is “close to” a solution $\boldsymbol{\theta}_d$, and using the Taylor expansion:

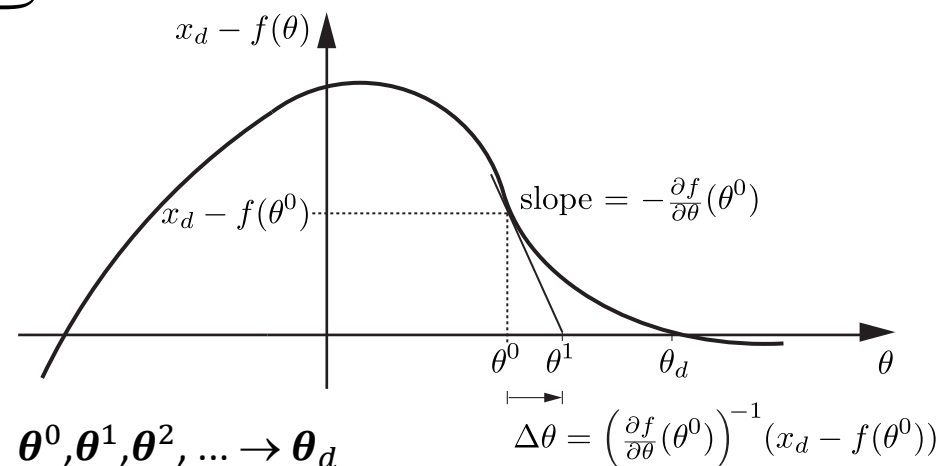
$$\mathbf{x}_d = \mathbf{f}(\boldsymbol{\theta}_d) = \mathbf{f}(\boldsymbol{\theta}^0) + \underbrace{\frac{\partial \mathbf{f}}{\partial \boldsymbol{\theta}} \bigg|_{\boldsymbol{\theta}^0}}_{\mathbf{J}(\boldsymbol{\theta}^0) \in \mathbb{R}^{m \times n}} \underbrace{(\boldsymbol{\theta}_d - \boldsymbol{\theta}^0)}_{\Delta \boldsymbol{\theta}} + \text{h.o.t.}$$

Analytical Jacobian at $\boldsymbol{\theta}^0$

Approximately: $\mathbf{J}(\boldsymbol{\theta}^0) \Delta \boldsymbol{\theta} = \mathbf{x}_d - \mathbf{f}(\boldsymbol{\theta}^0)$
(h.o.t. = 0)

- * If \mathbf{J} is square ($m = n$) and invertible:

$$\Delta \boldsymbol{\theta} = \mathbf{J}^{-1}(\boldsymbol{\theta}^0)(\mathbf{x}_d - \mathbf{f}(\boldsymbol{\theta}^0))$$



Inverse Kinematics Based on Newton–Raphson Method (Minimum Coordinate IK)

* If J is not invertible, either because it is not square or because it is singular,:

$$\Delta\theta = J^{\dagger}(\theta^0)(x_d - f(\theta^0)) \quad J^{\dagger}: \text{Moore–Penrose pseudoinverse}$$

* If J is full rank ($\text{rank}(A) = \min(m, n)$), i.e., the robot is not at a singularity:

- If $n > m$ (the robot has more joints n than end-effector coordinates m):

$$J^{\dagger} = J^T(JJ^T)^{-1}$$

- If $n < m$ (the robot has fewer joints n than end-effector coordinates m or it is at a singularity).

$$J^{\dagger} = (J^T J)^{-1} J^T$$

Note: If there are multiple inverse kinematics solutions, the iterative process tends to converge to the solution that is “closest” to the initial guess θ^0 .

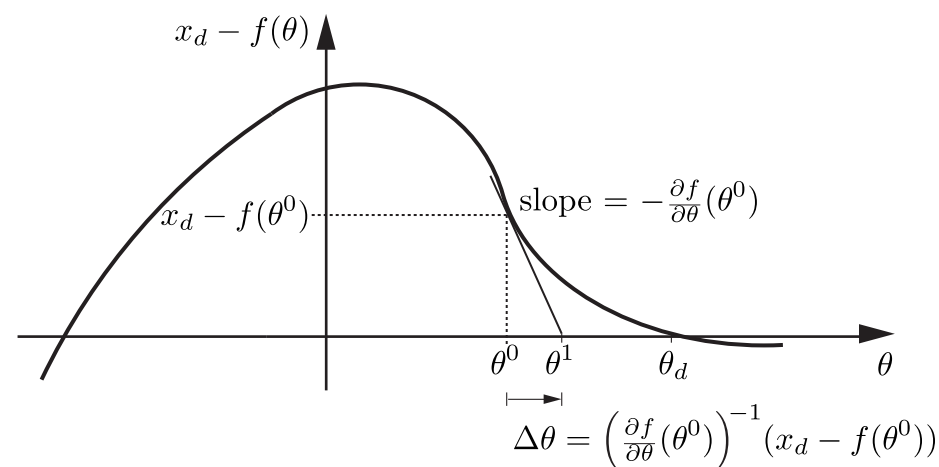
Algorithm for Minimum Coordinate Representation

a) Initialization: Given $x_d \in \mathbb{R}^m$ and an initial guess $\theta^0 \in \mathbb{R}^n$, set $i = 0$.

b) Iteration: Set $e = x_d - f(\theta^i)$. While $\|e\| > \epsilon$ for some small $\epsilon \in \mathbb{R}$:

- Set $\theta^{i+1} = \theta^i + J^\dagger(\theta^i)e$.
- Increment i .

```
max_iterations = 20;  
i = 0;  
Theta = Theta_0;  
e = X_d - FK(Theta);  
while norm(e) > epsilon && i < max_iterations  
    Theta = Theta + pinv(J(Theta)) * e;  
    i = i + 1;  
    e = X_d - FK(Theta);  
end
```



Inverse Kinematics Based on Newton–Raphson Method (Geometric IK)

- ❖ If the desired end-effector configuration represented as the Transformation Matrix, i.e., $T_{sd} = T(\theta) \in SE(3)$, $\theta \in \mathbb{R}^n$, then the goal is to find joint coordinates θ_d such that

$$T_{sd} - T(\theta_d) = \mathbf{0}$$

Algorithm for Matrix Transformation Representation:

- a) Initialization:** Given $T_{sd} \in SE(3)$ and an initial guess $\theta^0 \in \mathbb{R}^n$, set $i = 0$.
- b) Iteration:** Set $[\mathcal{V}_b] = \log(T_{bd}(\theta^i)) = \log(T_{sb}^{-1}(\theta^i)T_{sd})$. While $\|\omega_b\| > \epsilon_\omega$ or $\|\mathcal{V}_b\| > \epsilon_v$ for some small $\epsilon_\omega, \epsilon_v \in \mathbb{R}$:
(in Body Frame)
- Set $\theta^{i+1} = \theta^i + J_b^\dagger(\theta^i)\mathcal{V}_b$.
 - Increment i .
-
- a) Initialization:** Given $T_{sd} \in SE(3)$ and an initial guess $\theta^0 \in \mathbb{R}^n$, set $i = 0$.
- b) Iteration:** Set $[\mathcal{V}_s] = [\text{Ad}_{T_{sb}}] \log(T_{bd}(\theta^i)) = [\text{Ad}_{T_{sb}}] \log(T_{sb}^{-1}(\theta^i)T_{sd})$. While $\|\omega_s\| > \epsilon_\omega$ or $\|\mathcal{V}_s\| > \epsilon_v$ for some small $\epsilon_\omega, \epsilon_v \in \mathbb{R}$:
(in Space Frame)
- Set $\theta^{i+1} = \theta^i + J_s^\dagger(\theta^i)\mathcal{V}_s$.
 - Increment i .
- (\mathcal{V}_s or \mathcal{V}_b is the twist that takes T_{sb} to T_{sd} in 1s)