### **Ch6: Inverse Kinematics**

Amin Fakhari, Spring 2022

**Inverse Kinematics** 

### **Inverse Kinematics**

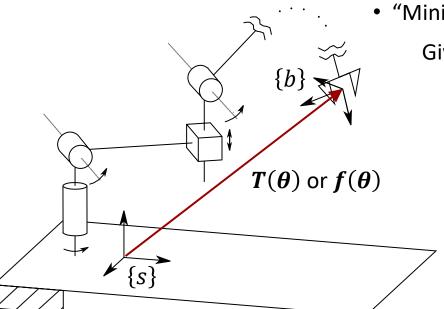


#### **Inverse Kinematics**

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

• "Geometric" inverse kinematics:

Given 
$$T_{sb} = T(\theta) \in SE(3)$$
, 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**

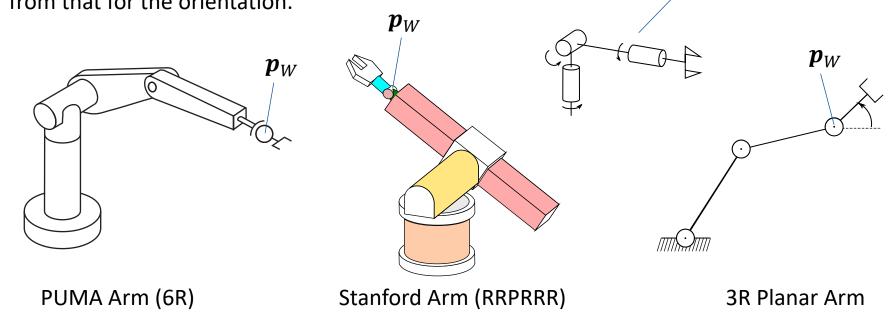
- The equations to solve are in general nonlinear, and thus it is not always possible to find a closed-form solution.
- Multiple solutions may exist.
- Infinite solutions may exist (e.g., in the case of a kinematically redundant manipulator).
- There might be no admissible solutions (e.g., when the given EE pose does not belong to the manipulator dexterous workspace.).
- ➤ Solving Inverse Kinematics Problems:
- Analytic Methods: Finding closed-form solutions using <u>algebraic intuition</u> or <u>geometric intuition</u>.
- Iterative Numerical Methods: When there are no (or it is difficult to find) closed-form solutions.

### **Analytic Inverse Kinematics**



#### **Analytic Inverse Kinematics**

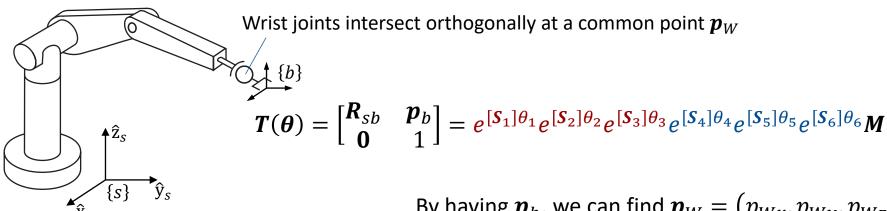
Most of the existing manipulators are typically formed by an **arm** and a **spherical wrist** (i.e., three consecutive revolute joint axes intersect at a common point). Thus, we can <u>decouple</u> the solution for the position (e.g., point  $p_W$  at the intersection of the three revolute axes) from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



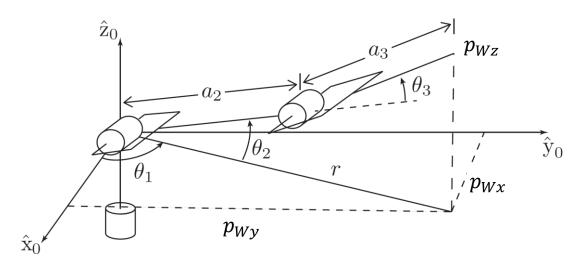
\* Therefore, it is possible to solve the inverse kinematics for the arm separately from the inverse kinematics for the spherical wrist.



#### **6R PUMA-Type Arms**

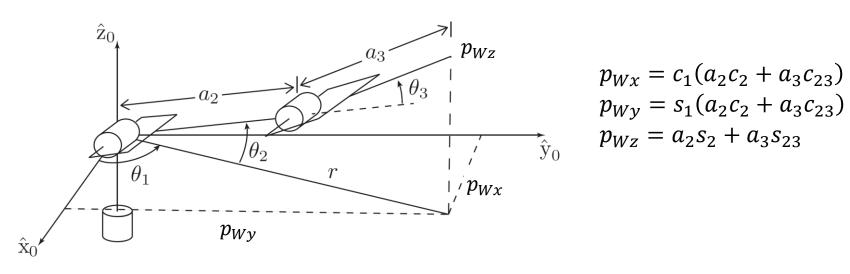


By having  $\boldsymbol{p}_b$ , we can find  $\boldsymbol{p}_W = \left(p_{Wx}, p_{Wy}, p_{Wz}\right)$ 



**Inverse Kinematics** 





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

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

$$c_{3} = \frac{p_{Wx}^{2} + p_{Wy}^{2} + p_{Wz}^{2} - a_{2}^{2} - a_{3}^{2}}{2a_{2}a_{3}}$$

$$c_{3} = \pm \sqrt{1 - c_{3}^{2}}$$

$$\theta_{3} = \operatorname{atan2}(s_{3}, c_{3})$$

$$\theta_{3,II} = -\theta_{3,I}$$



$$p_{Wx}^{2} + p_{Wy}^{2} = (a_{2}c_{2} + a_{3}c_{23})^{2} \longrightarrow a_{2}c_{2} + a_{3}c_{23} = \pm \sqrt{p_{Wx}^{2} + p_{Wy}^{2}}$$

$$p_{Wz} = a_{2}s_{2} + a_{3}s_{23}$$

$$s_{23} = s_{2}c_{3} + s_{3}c_{2}$$

$$c_{23} = c_{2}c_{3} - s_{2}s_{3}$$

$$c_{2} = \frac{\pm \sqrt{p_{Wx}^{2} + p_{Wy}^{2}(a_{2} + a_{3}c_{3}) + p_{Wz}a_{3}s_{3}}}{a_{2}^{2} + a_{3}^{2} + 2a_{2}a_{3}c_{3}}$$

$$c_{2} = \frac{p_{Wz}(a_{2} + a_{3}c_{3}) \mp \sqrt{p_{Wx}^{2} + p_{Wy}^{2}a_{3}s_{3}}}{a_{2}^{2} + a_{3}^{2} + 2a_{2}a_{3}c_{3}}$$

$$\Rightarrow c_{2} = \frac{p_{Wz}(a_{2} + a_{3}c_{3}) \mp \sqrt{p_{Wx}^{2} + p_{Wy}^{2}a_{3}s_{3}}}{a_{2}^{2} + a_{3}^{2} + 2a_{2}a_{3}c_{3}}$$

For each  $\theta_3$ , we have two solutions for  $\theta_2$ :

$$\theta_{3,\mathrm{I}} \to \left(\theta_{2,\mathrm{I}}, \theta_{2,\mathrm{II}}\right)$$
$$\theta_{3,\mathrm{II}} \to \left(\theta_{2,\mathrm{III}}, \theta_{2,\mathrm{IV}}\right)$$



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

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

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

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

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

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

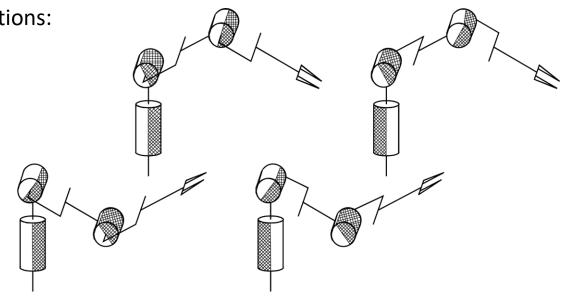
Thus, in total, there exist four solutions:

$$\begin{pmatrix} \theta_{1,\mathrm{I}}, \theta_{2,\mathrm{I}}, \theta_{3,\mathrm{I}} \end{pmatrix}$$

$$\begin{pmatrix} \theta_{1,\mathrm{I}}, \theta_{2,\mathrm{III}}, \theta_{3,\mathrm{II}} \end{pmatrix}$$

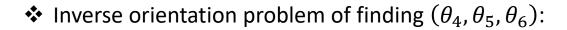
$$\begin{pmatrix} \theta_{1,\mathrm{II}}, \theta_{2,\mathrm{II}}, \theta_{3,\mathrm{I}} \end{pmatrix}$$

$$\begin{pmatrix} \theta_{1,\mathrm{II}}, \theta_{2,\mathrm{IV}}, \theta_{3,\mathrm{II}} \end{pmatrix}$$

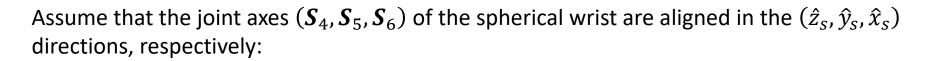




**Note**: When  $p_{Wx} = p_{Wy} = 0$ , the arm is in a kinematically singular configuration, and there are infinitely many possible solutions for  $\theta_1$ .



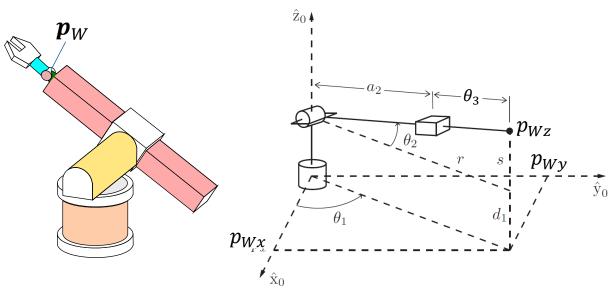
$$e^{[S_4]\theta_4}e^{[S_5]\theta_5}e^{[S_6]\theta_6} = e^{-[S_3]\theta_3}e^{-[S_2]\theta_2}e^{-[S_1]\theta_1}T(\theta)M^{-1} = T'$$
known



$$egin{aligned} & m{S}_{\omega_4} = (0,0,1) \\ & m{S}_{\omega_5} = (0,1,0) \end{aligned} & egin{aligned} & \operatorname{Rot}(\hat{\mathbf{z}}, heta_4) \operatorname{Rot}(\hat{\mathbf{y}}, heta_5) \operatorname{Rot}(\hat{\mathbf{x}}, heta_6) = m{R}' \end{aligned} & egin{aligned} & \text{This corresponds to} \\ & \text{the ZYX Euler angles.} \end{aligned} \\ & m{S}_{\omega_6} = (1,0,0) \end{aligned} & m{T}' = (m{R}', m{p}') \end{aligned} & egin{aligned} & m{C} & (\theta_4, \theta_5, \theta_6) \end{aligned}$$



#### **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:

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

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

- ⇒ 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**

**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  $f: \mathbb{R} \to \mathbb{R}$  is <u>differentiable</u>.

After first iteration After Second Iteration If  $\theta^0$  is an initial guess for the solution, Taylor expansion of  $f(\theta)$  at  $\theta^0$  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  $\theta$  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:

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

 $\epsilon$ : 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.,  $x_d = f(\theta) \in \mathbb{R}^m$ ,  $\theta \in \mathbb{R}^n$ , then the goal is to find joint coordinates  $\theta_d$  such that

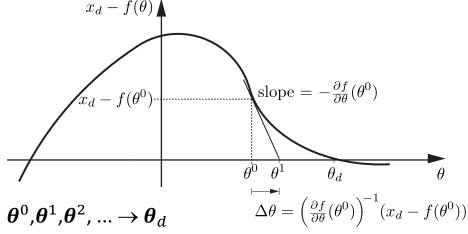
$$x_d - f(\theta_d) = 0$$
 (Assumption:  $f$  is differentiable)

• We use a method similar to the Newton–Raphson method for nonlinear root-finding: Given an initial guess  $\theta^0$  which is "close to" a solution  $\theta_d$ , and using the Taylor expansion:

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

\* If J is square (m = n) and invertible:

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



# 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,:

$$\Delta \boldsymbol{\theta} = \boldsymbol{J}^{\dagger}(\boldsymbol{\theta}^{0}) \big( \boldsymbol{x}_{d} - \boldsymbol{f}(\boldsymbol{\theta}^{0}) \big)$$
  $\boldsymbol{J}^{\dagger}$ : Moore–Penrose pseudoinverse

- \* If J is full rank (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):

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

- If n < m (the robot has fewer joints n than end-effector coordinates m or it is at a singularity).  $\boldsymbol{J}^\dagger = \left(\boldsymbol{J}^\mathrm{T}\boldsymbol{J}\right)^{-1}\boldsymbol{J}^\mathrm{T}$ 

**Note**: If there are multiple inverse kinematics solutions, the iterative process tends to converge to the solution that is "closest" to the initial guess  $\theta^0$ .



#### **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  $\boldsymbol{\theta}^{i+1} = \boldsymbol{\theta}^i + \boldsymbol{J}^{\dagger}(\boldsymbol{\theta}^i)\boldsymbol{e}$ .
  - 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</pre>
```

```
x_d - f(\theta^0)
\theta^0 \quad \theta^1 \quad \theta_d \quad \theta
\Delta \theta = \left(\frac{\partial f}{\partial \theta}(\theta^0)\right)^{-1} (x_d - f(\theta^0))
```

 $x_d - f(\theta)$ 



# 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  $\theta_d$  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  $[\mathcal{V}_b] = \log (\mathbf{T}_{bd}(\theta^i)) = \log(\mathbf{T}_{sb}^{-1}(\theta^i)\mathbf{T}_{sd})$ . While  $\|\boldsymbol{\omega}_b\| > \epsilon_\omega$  or  $\|\boldsymbol{v}_b\| > \epsilon_v$  for some small  $\epsilon_\omega$ ,  $\epsilon_v \in \mathbb{R}$ : (in Body Frame)
  - Set  $\boldsymbol{\theta}^{i+1} = \boldsymbol{\theta}^i + \boldsymbol{J}_b^{\dagger}(\boldsymbol{\theta}^i)\boldsymbol{\mathcal{V}}_b$ .
  - 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  $[\mathcal{V}_s] = [\mathrm{Ad}_{T_{sb}}] \log (T_{bd}(\theta^i)) = [\mathrm{Ad}_{T_{sb}}] \log (T_{sb}^{-1}(\theta^i)T_{sd})$ . While  $\|\boldsymbol{\omega}_s\| > \epsilon_{\omega}$  or  $\|\boldsymbol{v}_s\| > \epsilon_v$  for some small  $\epsilon_{\omega}$ ,  $\epsilon_v \in \mathbb{R}$ : (in Space Frame)
  - Set  $\boldsymbol{\theta}^{i+1} = \boldsymbol{\theta}^i + \boldsymbol{J}_s^{\dagger}(\boldsymbol{\theta}^i)\boldsymbol{\mathcal{V}}_s$ .
  - Increment i.

( $\mathcal{V}_{s}$  or  $\mathcal{V}_{b}$  is the twist that takes  $\mathbf{\textit{T}}_{sb}$  to  $\mathbf{\textit{T}}_{sd}$  in 1s)