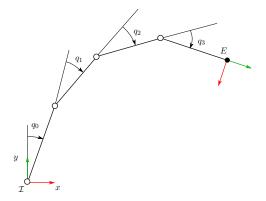
### Inverse Kinematics

August 27, 2019

- Introduction
- Forward and Inverse Kinematics
- Forward Differential Kinematics
- Pseudo Inverse Matrices
- Inverse Differential Kinematics
- Numerical Approach To Solve Inverse Kinematics Problems

#### What is kinematics?

- Greek:  $\kappa \iota \nu \eta \mu \alpha$ , kinema, movement
- Movement of a body can be described by positions, velocities, accelerations
- Kinematics does not describe what causes the movement (i.e. no forces and torques)



Generalized Coordinates

$$q = (q_1, q_2, \dots, q_n)^T$$

Endeffector Position and Orientation (Cartesian Coordinates)

$$X_E = (x, y, z, \theta_x, \theta_y, \theta_z)^T$$



### Forward Kinematics $X_E = f_F(q)$

- Straightforward
- $\bullet$  Use to obtain endeffector position and orientation given all joint angles q

### Inverse Kinematics $q = f_I(X_E)$

- ullet Hard or even impossible to obtain the function  $f_I$
- If closed-form expression can be found, very fast
- Multiple solutions may be obtained

## Forward Differential Kinematics $\dot{X}_E = J_{e0}(q)\dot{q}$

- Linear map:  $J_{e0}(q): \mathbb{R}^n \to \mathbb{R}^6$
- Linearization of robot configuration around a given joint configuration

## Geometric Endeffector Jacobian $J_{e0}(q) \in \mathbb{R}^{6 \times n}$

- Structure is only a function of the robot geometry
- In general  $J_{e0}^{-1}(q)$  is not defined (i.e.  $n \neq 6$ ), hence we need a *Pseudo Inverse*

# Inverse Differential Kinematics $\dot{q}=J_{e0}^{+}(q)\dot{X}_{E}$

- Linear map:  $J_{e0}^+(q): \mathbb{R}^6 \to \mathbb{R}^n$
- Note:  $J_{e0}^+(q)$  is called the *Pseudo Inverse* of  $J_{e0}(q)$



Consider the system

$$Ax = b$$

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where  $A \in \mathbb{R}^{m \times n}$ ,  $x \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$  with  $m \neq n$ 

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- **1** Define  $A^+ := (A^T A)^{-1} A^T$

### Damped Least-Squares Solution (DLS)

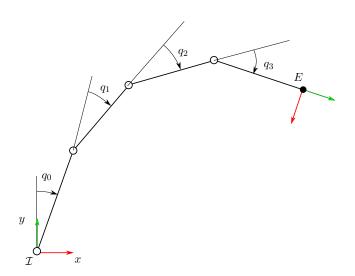
- Moore-Penrose inverse can become singular (i.e.  $A^TA$  is not invertible)
- Add small positive values to diagonal entries of  $A^TA$ , then matrix will *always* be invertible
- Define  $A^+ := (A^T A + \lambda \mathbb{I})^{-1} A^T$
- Note that the introduced error becomes larger as  $\lambda$  becomes larger.

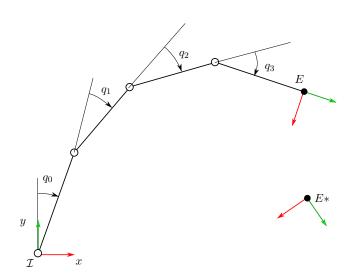
#### **Iterative Inverse Kinematics**

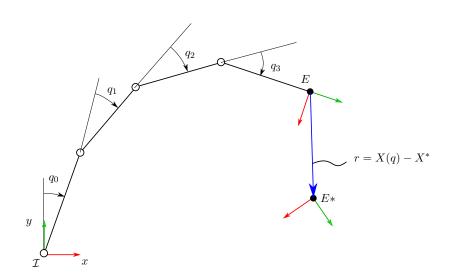
Given a desired endeffector position and orientation  $X_E^*$ , what are the corresponding joint angles  $q^*$ ?

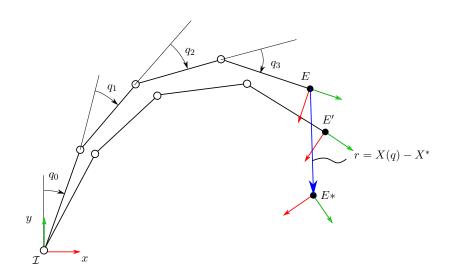
- **1** Initialize joints to some arbitrary, nonsingular position  $q^{(0)}$
- ② Compute the position and orientation of endeffector  $X_E(q^{(0)})$  using simple Forward Kinematics
- **3** Compute pose difference  $\Delta X_E = X_E^* X_E(q^{(0)})$
- lacktriangle Interpret  $\Delta X_E$  as velocity vector pointing to the goal pose
- Use Inverse Differential Kinematics to compute joint velocities that move the endeffector towards goal pose:  $\dot{q}^{(0)} = J_{e0}^+(q^{(0)})\Delta X_E$
- **1** Apply a scaled version of  $\dot{q}^{(0)}$  to the joints and update the robot pose, i.e.  $q^{(1)} = q^{(0)} + k\dot{q}^{(0)}$  where  $k \in \mathbb{R}^+$
- Go to step 2, use updated joint configuration  $q^{(1)}$

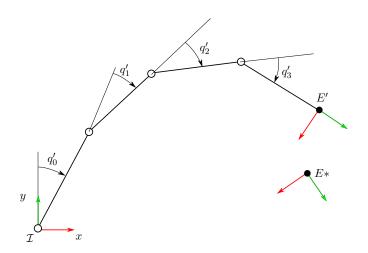


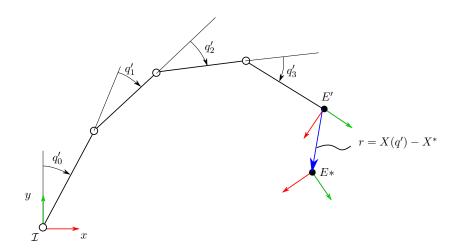


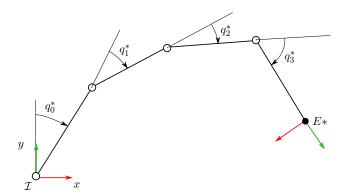












#### **Iterative Inverse Kinematics**

What can go wrong?

- Initialize in singular position
- ullet If  $k\dot{q}^{(i)}$  is large, the linearization may no longer be a good approximation of the robots configuration
- No axis-limits are considered
- ullet Moore-Penrose inverse may become singular o use DLS