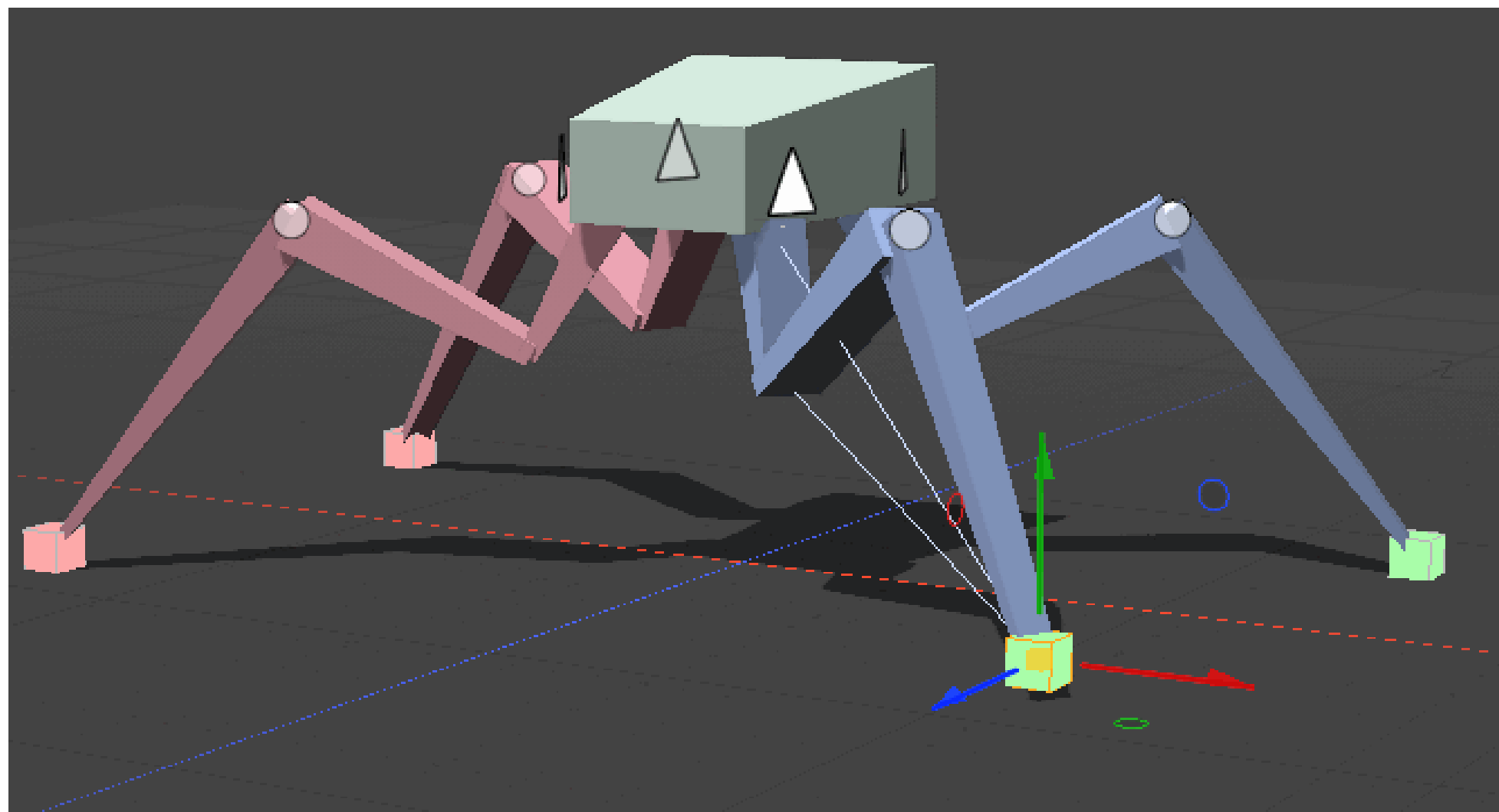


# Introduction to Inverse Kinematics

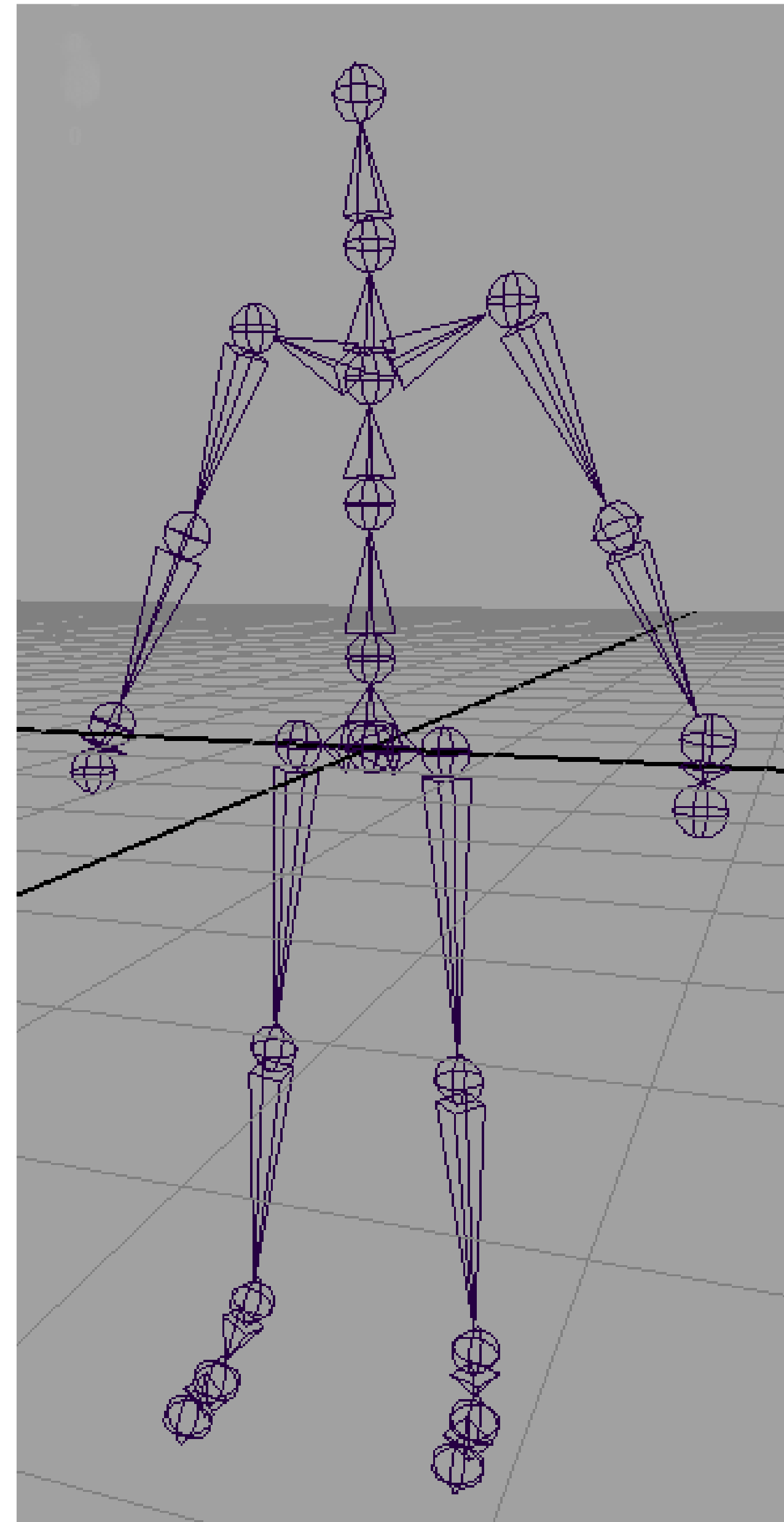


# Learning Objectives

- Learn how to formulate inverse kinematics as a numerical optimization problem
- Understand two of the most common techniques for solving IK, the  $J^t$  (Jacobian transpose) and the  $J^+$  (Jacobian pseudo-inverse) methods
- Learn how to implement and debug gradient-based optimization methods

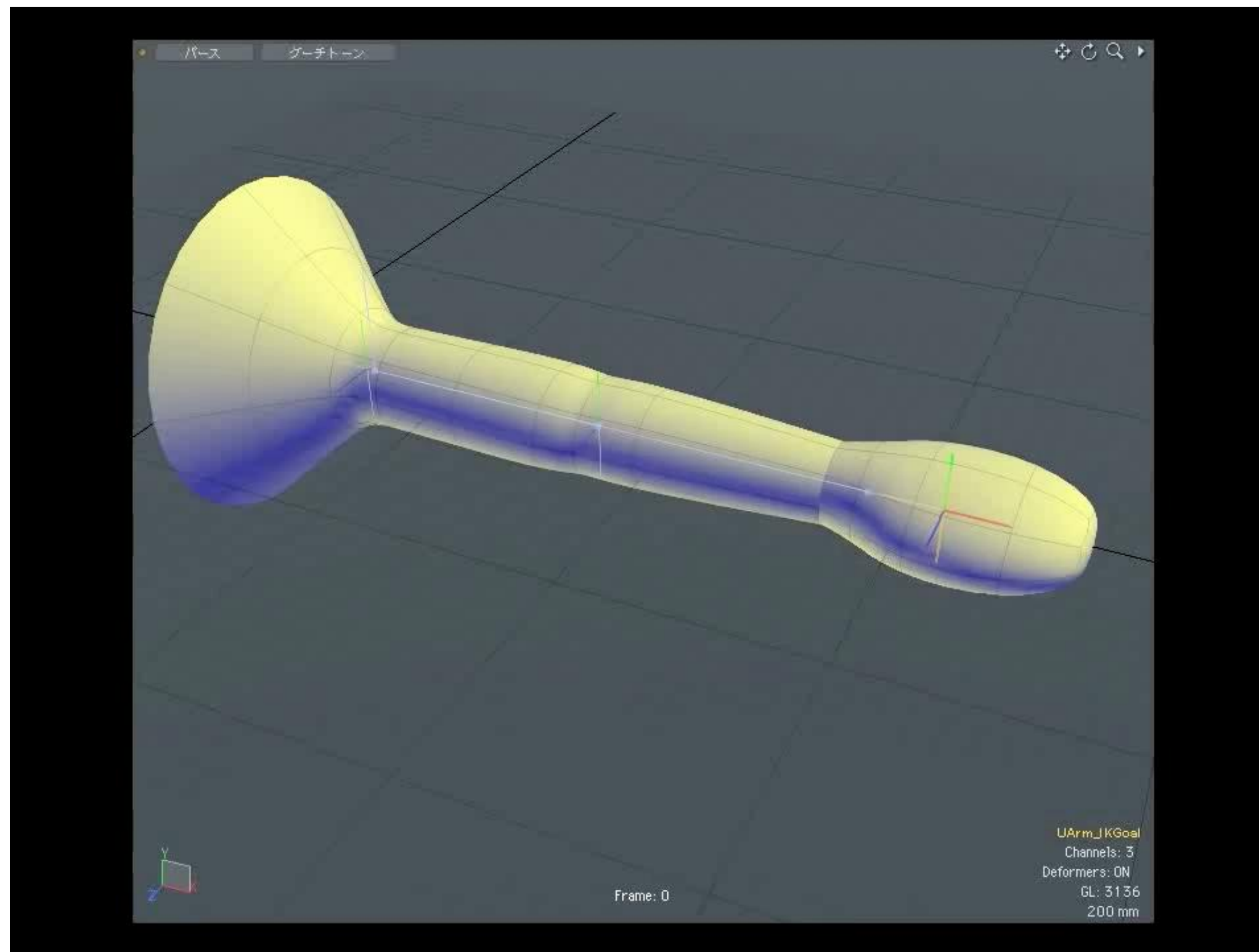
# Forward Kinematics

- **Given joint angles as input, compute skeleton pose**
  - **Bones/links/body parts defined by their geometry (e.g. shape) and a coordinate frame**
  - **Joints define relative rotation between child and parent coordinate frames**
    - **They must store type, position on parent/child frames, rotation axis, etc.**



# Inverse Kinematics (IK)

- Given goal(s) for “end effector” compute joint angles



- Many, many algorithms: analytic formulations for specific cases, cyclic coordinate descent,  $J^T/J^+$  methods, etc

# Typical applications of inverse kinematics

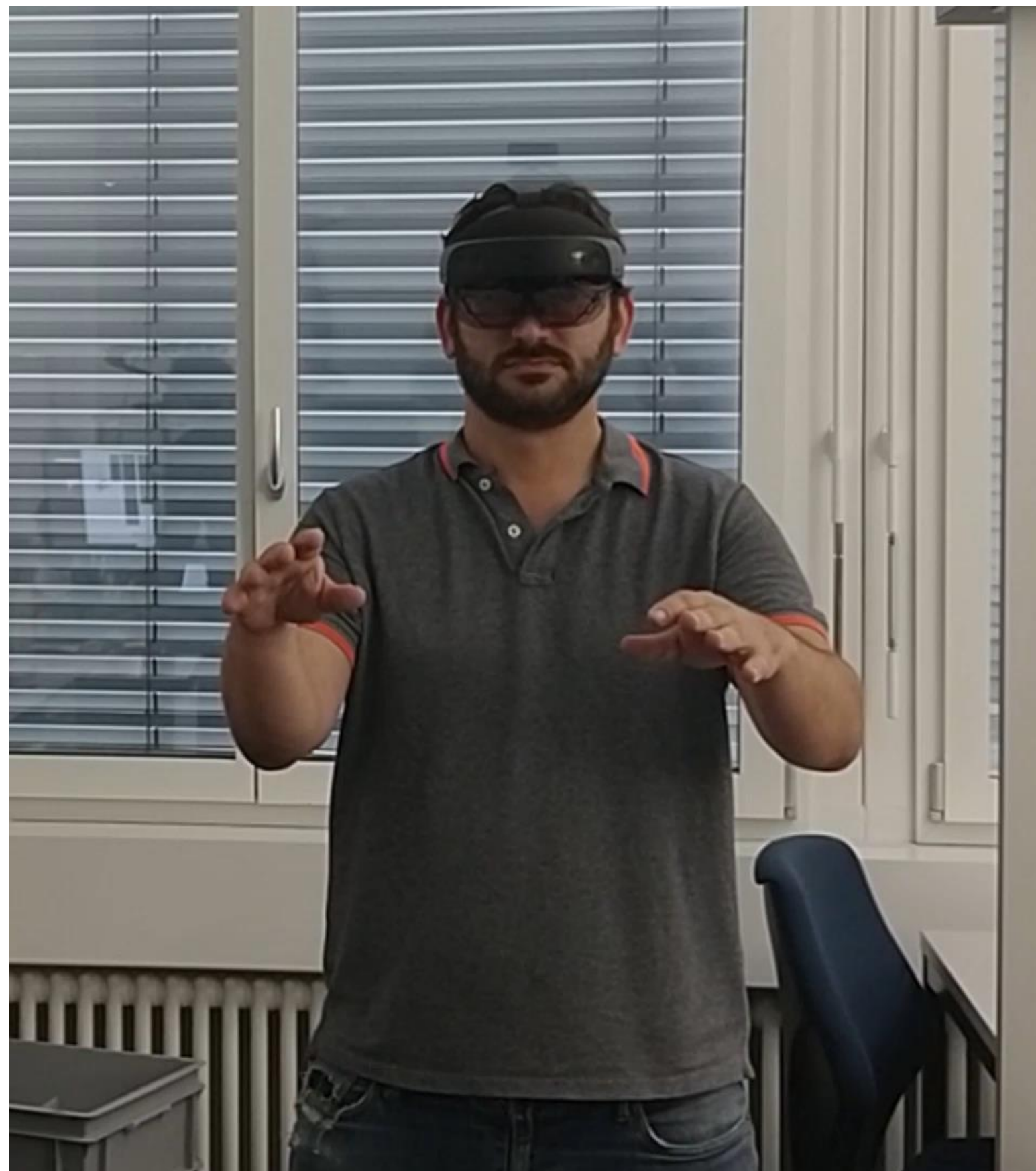
- Intuitive way of creating motions via high-level goals
- Full body motion capture
- Motion retargeting and teleoperation





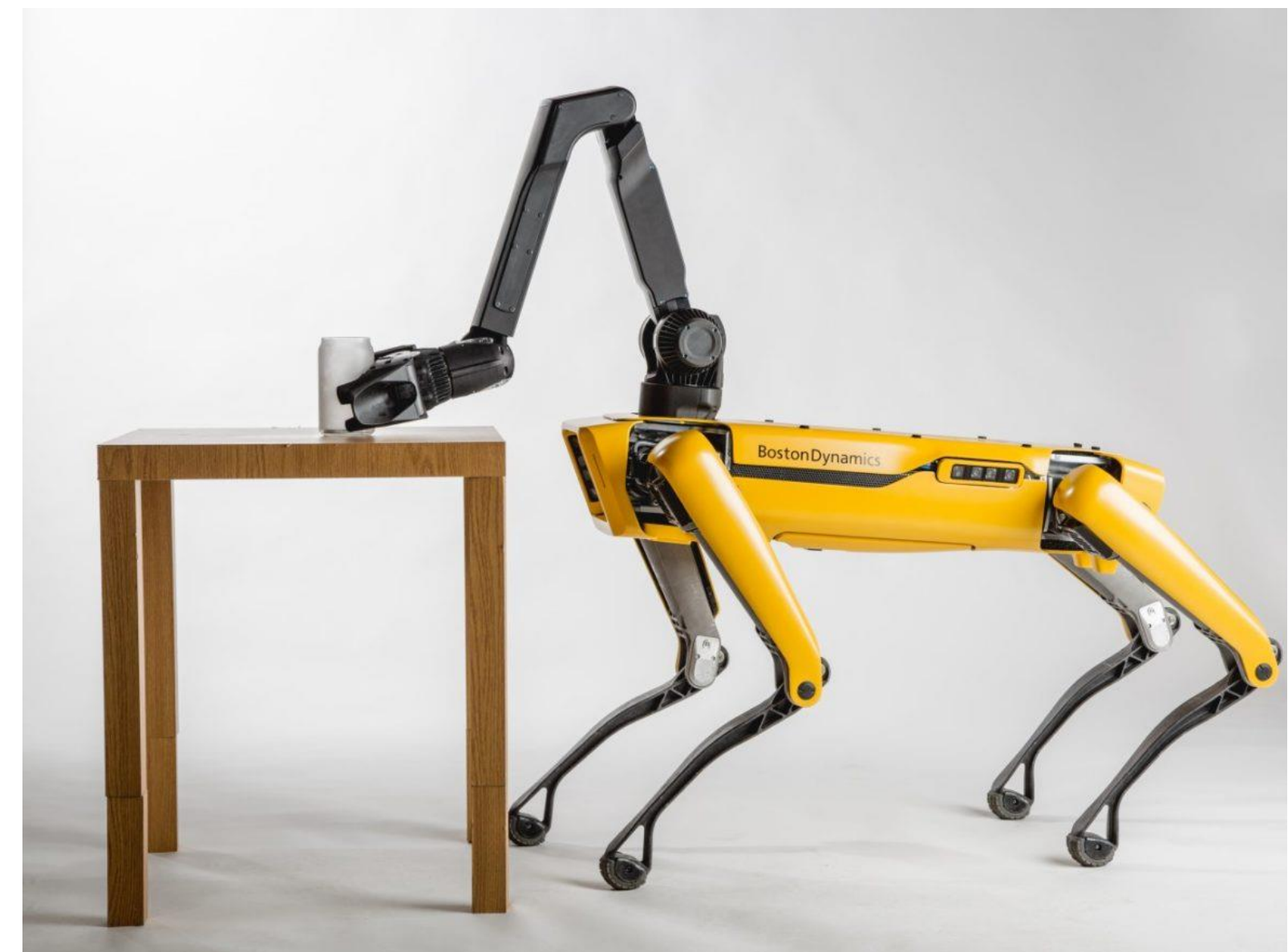
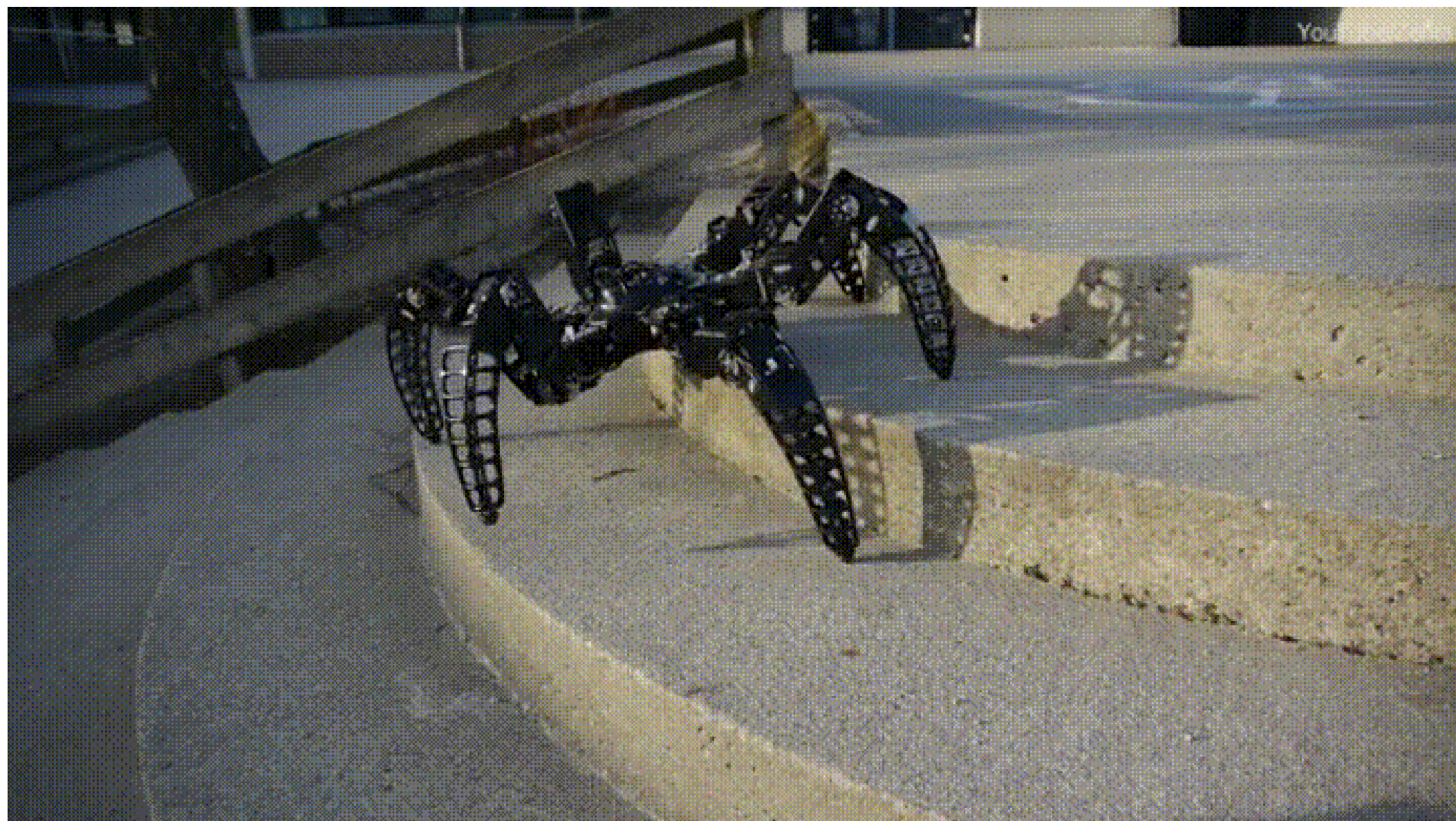
# Typical applications of inverse kinematics

- Intuitive way of creating motions via high-level goals
- Full body motion capture
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# Typical applications of inverse kinematics

- Intuitive way of creating motions via high-level goals
- Full body motion capture
- Motion retargeting and teleoperation
- Starting point for more sophisticated control methods



# **Inverse Kinematics**

**A numerical optimization approach**

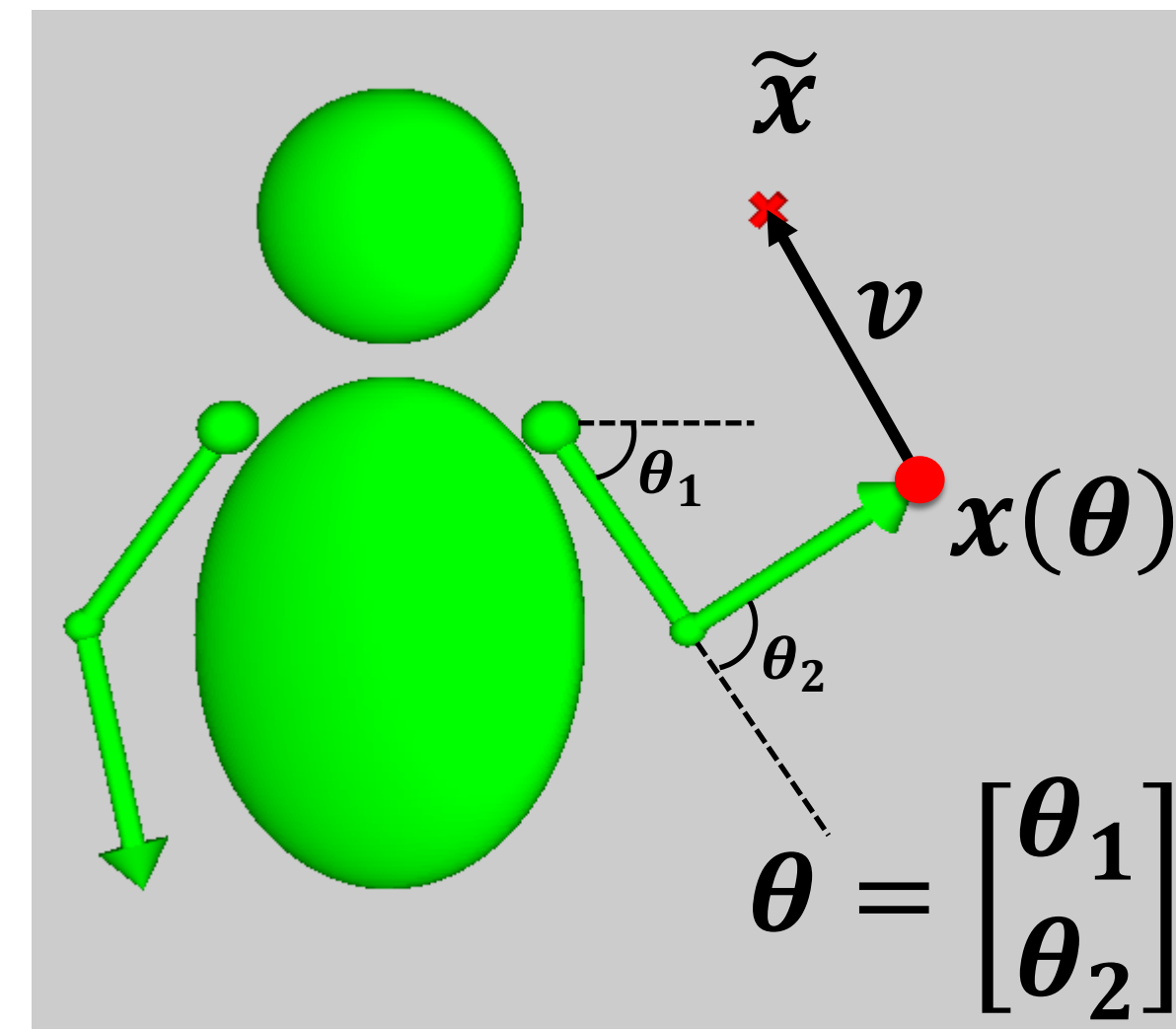


# Basic idea

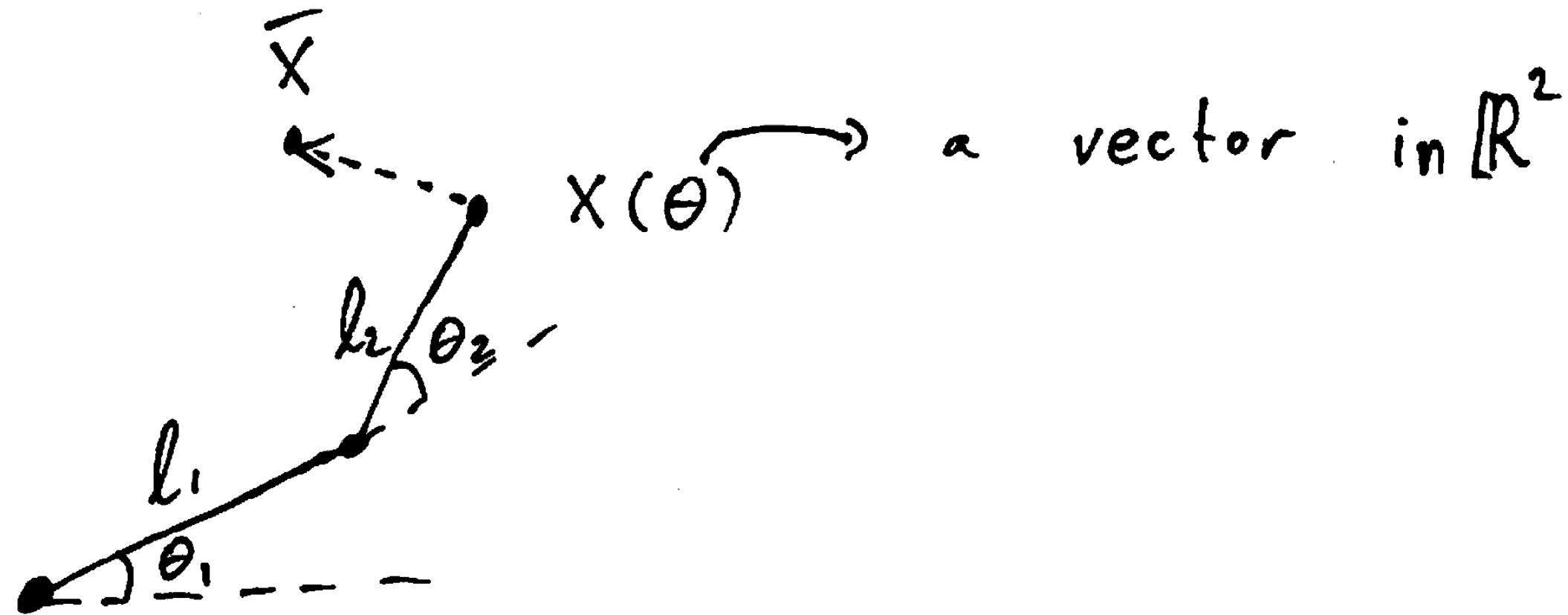
- Write down distance between final point and “target” and set up an objective/energy/error/loss function

$$E(\theta) = \frac{1}{2} (x(\theta) - \tilde{x})^T (x(\theta) - \tilde{x})$$

- Many possible extensions
  - Joint limits, pose regularizers, objectives on end effector orientation, etc.
- compute derivatives with respect to joint angles
- apply gradient-based methods to find minimum of objective



# Forward kinematics: computing $x(\theta)$



$$x(\theta) = \begin{bmatrix} l_1 \cos \theta_1 + l_2 \cos(\theta_1 + \theta_2) \\ l_1 \sin \theta_1 + l_2 \sin(\theta_1 + \theta_2) \end{bmatrix}$$

# Inverse kinematics: derivatives

$$x(\theta) = \begin{bmatrix} l_1 \cos \theta_1 + l_2 \cos(\theta_1 + \theta_2) \\ l_1 \sin \theta_1 + l_2 \sin(\theta_1 + \theta_2) \end{bmatrix}$$

The energy function to minimize:

$$E(\theta) = \frac{1}{2} (x(\theta) - \bar{x})^T \underbrace{(x(\theta) - \bar{x})}_v$$

$$\nabla E = \frac{dE}{d\theta} = \underbrace{\frac{dx}{d\theta}}_{m \times n} \cdot \underbrace{\frac{dE}{dx}}_{n \times 1} = \frac{dx^T}{d\theta} \cdot v = J^T v$$

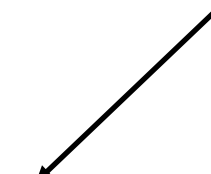
$$\frac{dx}{d\theta} = \begin{bmatrix} \frac{dx_x}{d\theta_1} & \frac{dx_x}{d\theta_2} \\ \frac{dx_y}{d\theta_1} & \frac{dx_y}{d\theta_2} \end{bmatrix} = \begin{bmatrix} \frac{dx}{d\theta_1} & \frac{dx}{d\theta_2} \end{bmatrix} = J$$

# Inverse kinematics: derivatives

$$x(\theta) = \begin{bmatrix} l_1 \cos \theta_1 + l_2 \cos(\theta_1 + \theta_2) \\ l_1 \sin \theta_1 + l_2 \sin(\theta_1 + \theta_2) \end{bmatrix}$$

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$$\nabla_{\theta}^2 E = J^T J + \frac{dJ}{d\theta} \cdot v$$

These are the basic ingredients you need to get started with IK