# Mathematics for Inverse Kinematics

15-464: Technical Animation

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#### Overview

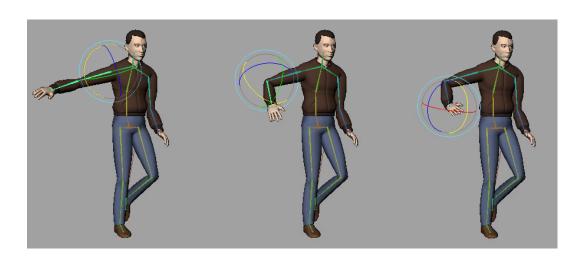
- Kinematics
- Forward Kinematics and Inverse Kinematics
- Jabobian
- Pseudoinverse of the Jacobian
- Assignment 2

## Vocabulary of Kinematics

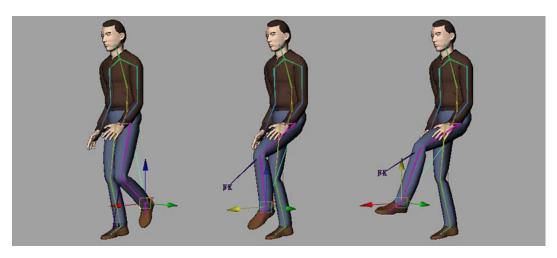
- Kinematics is the study of how things move, it describes the motion of a hierarchical skeleton structure.
- Base and End Effector.



# FK vs. IK



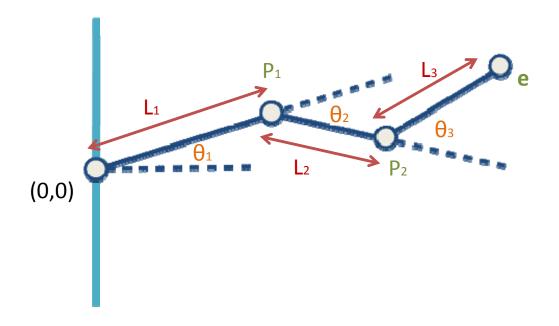
**Forward Kinematics** 



**Inverse Kinematics** 

# Forward Kinematics

 The process of computing world space geometric description based on joint DOF values.



#### **Forward Kinematics**

We have joint DOF values:

$$\mathbf{\Theta} = [\mathbf{\Theta}_1 \ \mathbf{\Theta}_2 \ \cdots \ \mathbf{\Theta}_M]$$

 We want the end effector description in world space (N=3 in our case):

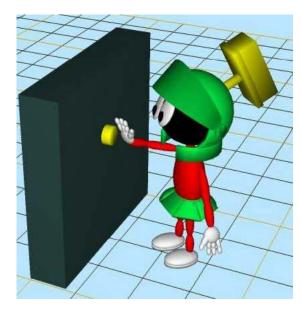
$$\mathbf{e} = [\mathbf{e}_1 \ \mathbf{e}_2 \ \cdots \ \mathbf{e}_N]$$

• FK gives us:

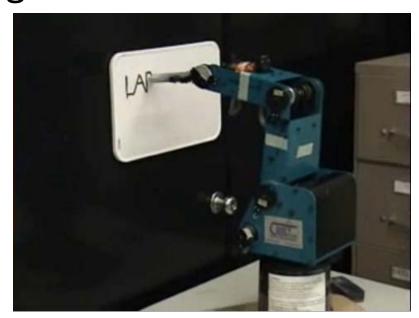
$$e = f(\theta)$$

#### But Sometimes We Want the Opposite

 We want to know how the upper joints of the hierarchy would rotate if we want the end effector to reach some goal.



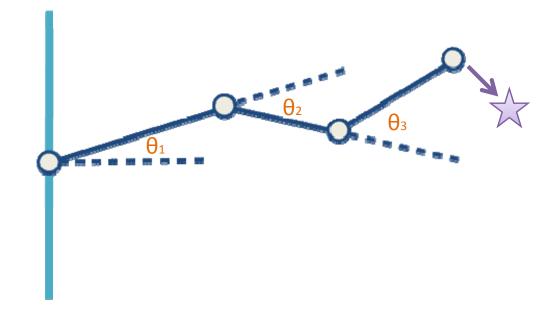
Animations



Robotics

#### **Inverse Kinematics**

 The goal of inverse kinematics is to compute the vector of joint DOFs that will cause the end effector to reach some desired goal state



#### **Inverse Kinematics**

• We have:

$$\mathbf{e} = [\mathbf{e}_1 \ \mathbf{e}_2 \ \cdots \ \mathbf{e}_N]$$

• And we want:

$$\mathbf{\Theta} = [\mathbf{\Theta}_1 \ \mathbf{\Theta}_2 \ \cdots \ \mathbf{\Theta}_M]$$

• We need:

$$\theta = f^{-1}(e)$$

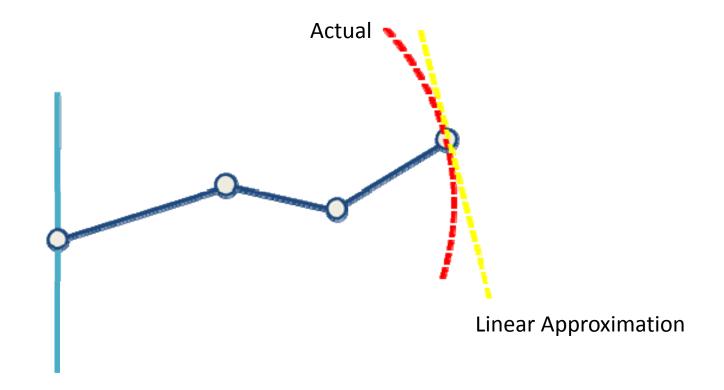
#### Inverse Kinematics Issues

- While FK is relatively easy to evaluate.
- IK is more challenging: several possible solutions, or sometimes maybe no solutions.
- Require Complex and Expensive computations to find a solution.

### **IK Solutions**

- Jacobian
- Cyclic Coordinate Descent (CCD)
- Required to implement in Assignment 2

What is Jacobian? A linear approximation to f()



- Matrix of partial derivatives of entire system.
- Defines how the end effector e changes relative to instantaneous changes in the system.

$$J = \frac{d\mathbf{e}}{d\mathbf{\theta}} \qquad \qquad d\mathbf{e} = Jd\mathbf{\theta}$$

$$\mathbf{e} = \begin{bmatrix} \mathbf{e}_{\mathbf{x}} & \mathbf{e}_{\mathbf{y}} & \mathbf{e}_{\mathbf{z}} \end{bmatrix}^{\mathrm{T}}$$
$$\mathbf{\theta} = \begin{bmatrix} \mathbf{\theta}_{1} & \mathbf{\theta}_{2} & \cdots & \mathbf{\theta}_{M} \end{bmatrix}^{\mathrm{T}}$$

$$J = \begin{bmatrix} \frac{\partial e_x}{\partial \theta_1} & \frac{\partial e_x}{\partial \theta_2} & \dots & \frac{\partial e_x}{\partial \theta_M} \\ \frac{\partial e_y}{\partial \theta_1} & \frac{\partial e_y}{\partial \theta_2} & \dots & \frac{\partial e_y}{\partial \theta_M} \\ \frac{\partial e_z}{\partial \theta_1} & \frac{\partial e_z}{\partial \theta_2} & \dots & \frac{\partial e_z}{\partial \theta_M} \end{bmatrix}$$

Recall that

$$\theta = f^{-1}(e)$$

$$d\theta = J^{-1}de$$

$$de = Jd\theta$$

#### Problems

How to compute J?

Numerically (Required)

Analytically (Extra Credit)

• How to invert J?

Pseudoinverse of Jacobian (Required)

Cheat by using transpose (Too easy, we don't do that)

## Computing the Jacobian Numerically

Let's examine one column of the Jacobian
 Matirx
 Δο Γδο δο 1<sup>T</sup>

$$\frac{\partial \mathbf{e}}{\partial \mathbf{\theta}_{i}} = \left[ \frac{\partial \mathbf{e}_{x}}{\partial \mathbf{\theta}_{i}} \frac{\partial \mathbf{e}_{y}}{\partial \mathbf{\theta}_{i}} \frac{\partial \mathbf{e}_{z}}{\partial \mathbf{\theta}_{i}} \right]^{T}$$

- We can add a small  $\Delta\theta$  to  $\theta_i$
- Then we can calculate how the end effector moves:  $\Delta \mathbf{e} = \mathbf{e'} \mathbf{e}$
- Now we have:

$$\frac{\partial \mathbf{e}}{\partial \theta_1} \approx \frac{\Delta \mathbf{e}}{\Delta \theta} = \left[ \frac{\Delta \mathbf{e}_x}{\Delta \theta} \, \frac{\Delta \mathbf{e}_y}{\Delta \theta} \, \frac{\Delta \mathbf{e}_z}{\Delta \theta} \right]^T$$

### Computing the Jacobian Numerically

$$\frac{\partial \mathbf{e}}{\partial \theta_1} \approx \frac{\Delta \mathbf{e}}{\Delta \theta} = \left[ \frac{\Delta \mathbf{e}_x}{\Delta \theta} \, \frac{\Delta \mathbf{e}_y}{\Delta \theta} \, \frac{\Delta \mathbf{e}_z}{\Delta \theta} \right]^T$$

 We can use this method to fill the Jacobian Matirx!

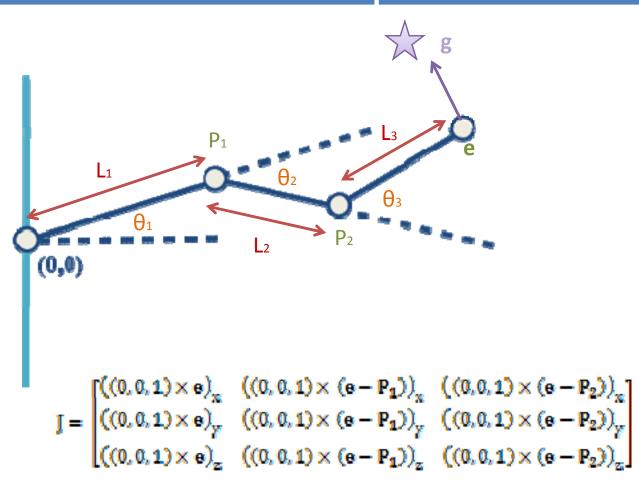
### Computing the Jacobian Analytically

 For a rotational joint, the linear change in the end effector is the cross product of the axis of revolution and a vector from the joint to the end effector.

$$\frac{\partial \mathbf{e}}{\partial \mathbf{\theta}_{i}} = \left[ \frac{\partial \mathbf{e}_{x}}{\partial \mathbf{\theta}_{i}} \frac{\partial \mathbf{e}_{y}}{\partial \mathbf{\theta}_{i}} \frac{\partial \mathbf{e}_{z}}{\partial \mathbf{\theta}_{i}} \right]^{T} = \left( \mathbf{a}_{i}' \times (\mathbf{e} - \mathbf{r}_{i}') \right)$$

 Important to make sure all the coordinate values are in the same coordinate system. (Hard to get right.)

# Computing the Jacobian Analytically — A 2D example



## Inverting the Jacobian

- No guarantee it is invertible
  - Typically not a square matrix.
  - Singularities.
  - Even it's invertible, as the pose vector changes,
     the properties of the matrix will change.

# Inverting the Jacobian— Pseudo Inverse

 We can try using the pseudo inverse to find a matrix that effectively inverts a non-square matrix:

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

# Inverting the Jacobian— Pseudo Inverse

$$d\mathbf{e} = \mathbf{J} \cdot d\mathbf{\theta}$$

$$J^{T} \cdot d\mathbf{e} = J^{T} \mathbf{J} \cdot d\mathbf{\theta}$$

$$(J^{T} \mathbf{J})^{-1} J^{T} \cdot d\mathbf{e} = (J^{T} \mathbf{J})^{-1} (J^{T} \mathbf{J}) \cdot d\mathbf{\theta}$$

$$(J^{T} \mathbf{J})^{-1} J^{T} \cdot d\mathbf{e} = d\mathbf{\theta}$$

$$J^{+} \cdot d\mathbf{e} = d\Delta\mathbf{\theta}$$

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

# Inverting the Jacobian— Jacobian Transpose

- Another technique is just to use the transpose of the Jacobian matrix.
- The Jacobian is already an approximation to f()—Cheat more
- It is much faster.
- But if you prefers quality over performance, the pseudo inverse method would be better.

# Solving IK—Incremental Changes

- FK is nonlinear
- Implies that the Jacobian can only be used as an approximation that is valid near the current configuration
- So we must Repeat the process of computing a Jacobian and then taking a small step towards the goal until we get close enough

# Solving IK—Algorithm of the Jacobian Method

```
while (e is too far from g){ compute the Jacobian matrix J compute the pseudoinverse of the Jacobian matrix— J^+ compute change in joint DOFs: \Delta \theta = J^+ \cdot \Delta e apply the change to DOFs, move a small step of \alpha \Delta \theta: \theta = \theta + \alpha \Delta \theta}
```

# Cyclic Coordinate Descent (CCD)

- Much easier than the Jacobian Method
- Read two articles "Oh My God, I inverted Kine!" and "Making Kine More Flexible"
- Will be talked about next time

## Assignment 2

- Will be out later today.
- Require to implement the Jacobian Method and CCD.
- Load ASF file into Maya and add IK to the skeleton.
- Start early! More challenging than Assignment
   1.

## Assignment 1 is Due Tonight

- Handin videos and presentation materials
- Presentation on Thursday!

# Questions?

Thank you and See you Thursday!