#### Forward and Inverse Kinematics

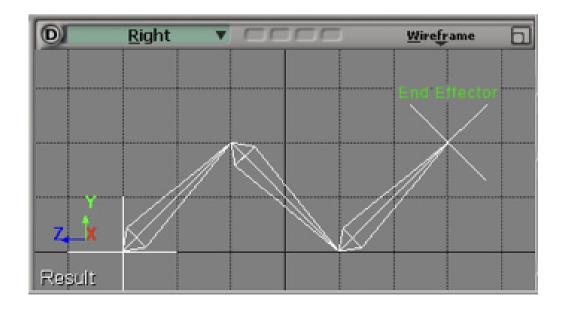
#### CS475 / 675, Fall 2016

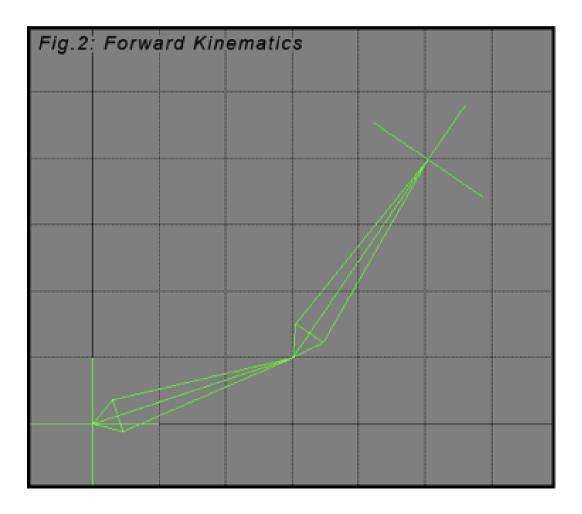
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Slides from Tiffany Barnes, Bill Baxter, Serafim Batzoglou and Jean-Claude Latombe

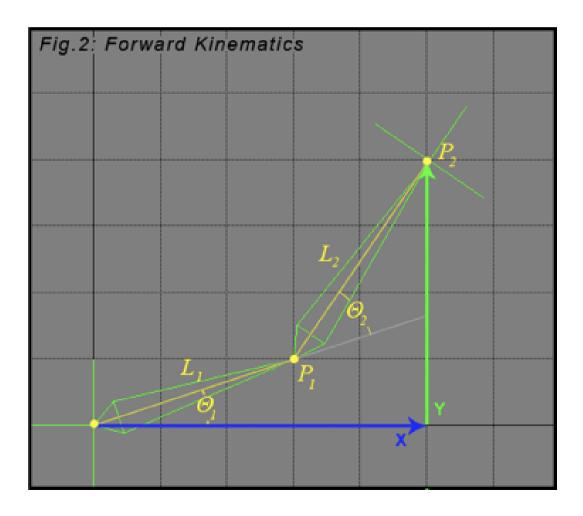
#### Forward Kinematics

- Incrementally manipulating each component of a flexible, jointed object to achieve an overall desired pose
- Mathematically: find the position of the end effector, given the angles of the joints and the length of each articulated segment

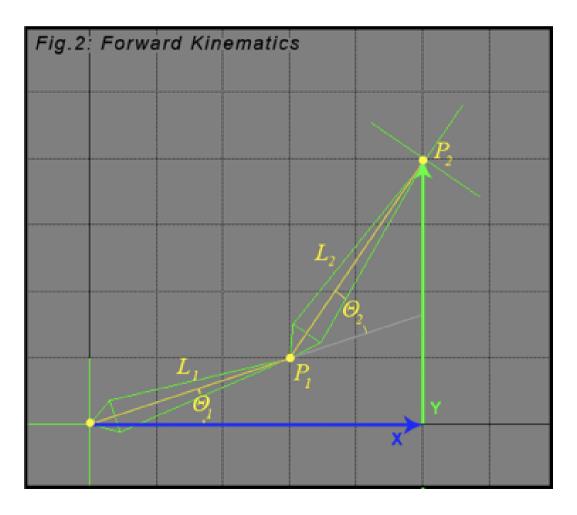




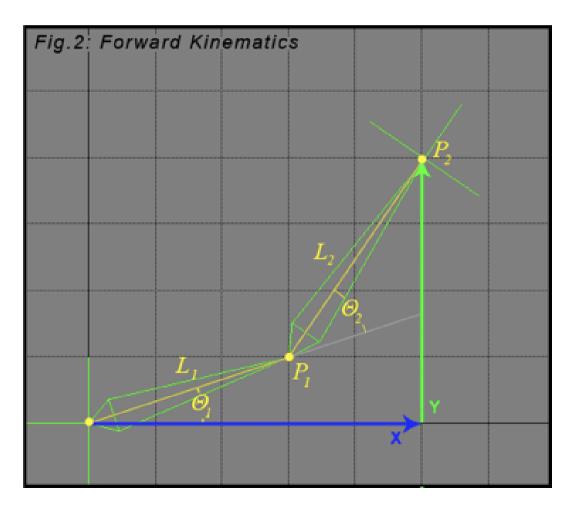
$$Px_1 = L_1 \times \cos(\Theta_1)$$
$$Py_1 = L_1 \times \sin(\Theta_1)$$



$$Px_{2} = Px_{1} + L_{2} \times \cos(\Theta_{1} + \Theta_{2})$$
$$Py_{2} = Py_{1} + L_{2} \times \sin(\Theta_{1} + \Theta_{2})$$



$$Px_{2} = L_{1} \times \cos(\Theta_{1}) + L_{2} \times \cos(\Theta_{1} + \Theta_{2})$$
$$Py_{2} = L_{1} \times \sin(\Theta_{1}) + L_{2} \times \sin(\Theta_{1} + \Theta_{2})$$



- Links should be drawn from the outermost link (nearest the end effector) to the innermost (nearest the root)
- The positioning of each link requires translations and rotations from preceding links

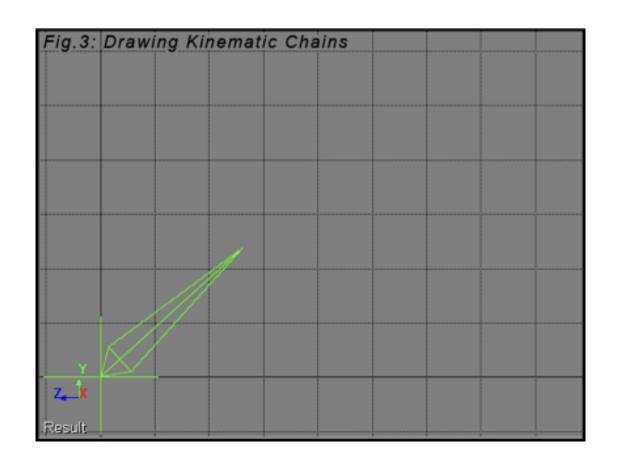
• Starting with the end effector's object:

Fig.3: Draw	ing Kinematic Cha	ins	
		******	
Z_X			
Result			

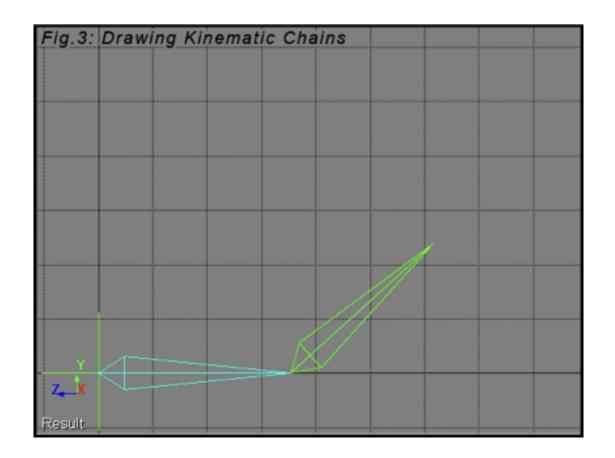
- Starting with the end effector's object:
  - Translate base to origin

Fig.3: Drawing Kinematic Chains					
Y T					
Z_X Result					
Result					

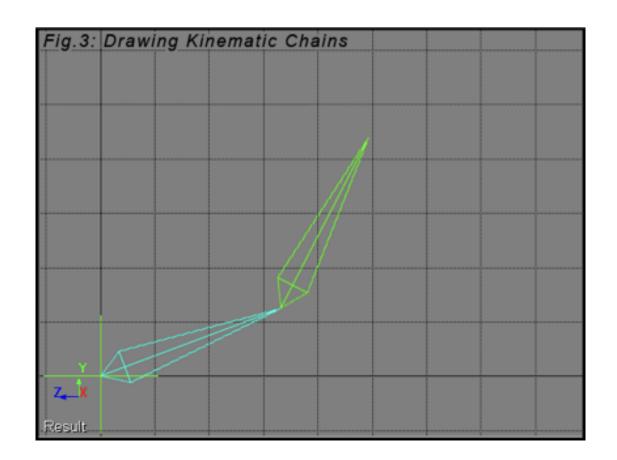
- Starting with the end effector's object:
  - Translate base to origin
  - Rotate by angle



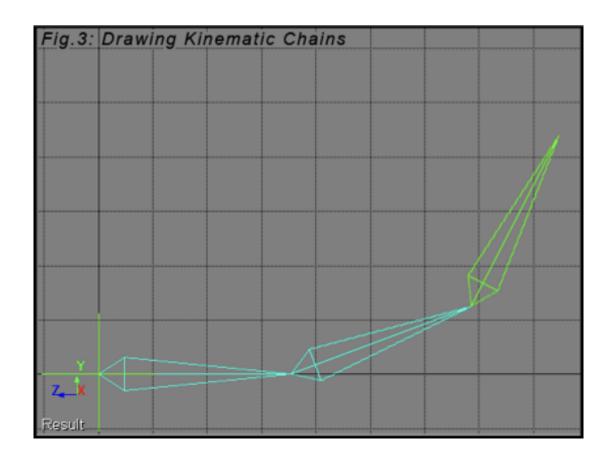
- Starting with the end effector's object:
  - Now translate by length of next link...



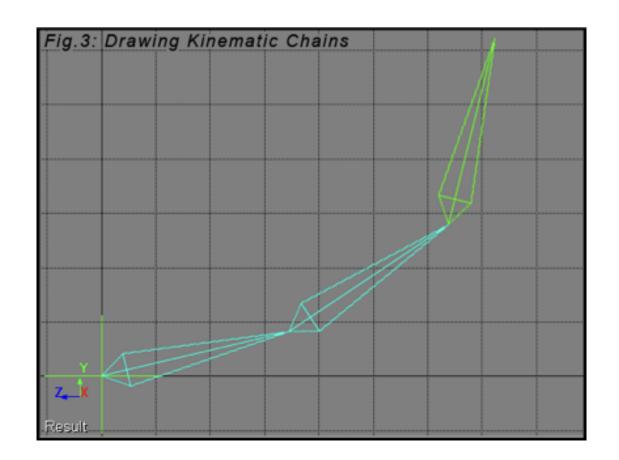
- Starting with the end effector's object:
  - Now translate by length of next link...
  - ... and rotate the entire chain by the angle of that link



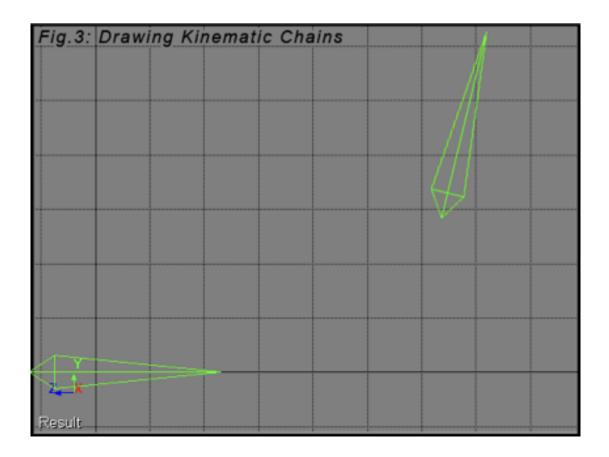
- Starting with the end effector's object:
  - Translate again by the length of the next link in the chain...



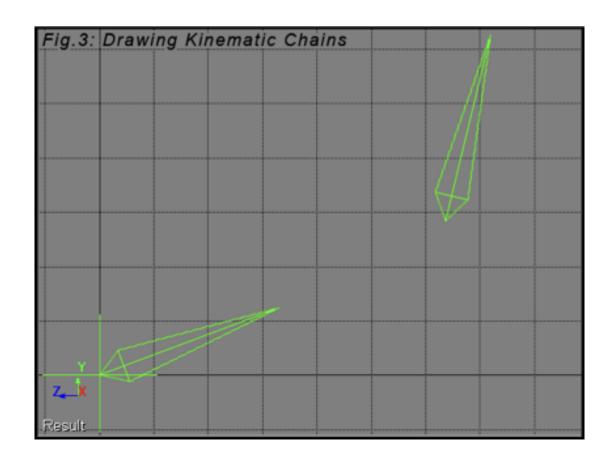
- Starting with the end effector's object:
  - Translate again by the length of the next link in the chain...
  - ... and rotate the entire chain by the angle of that link



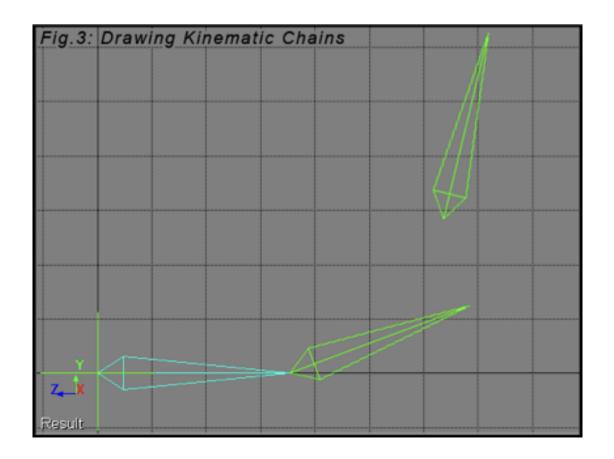
• Starting with the next link's object:



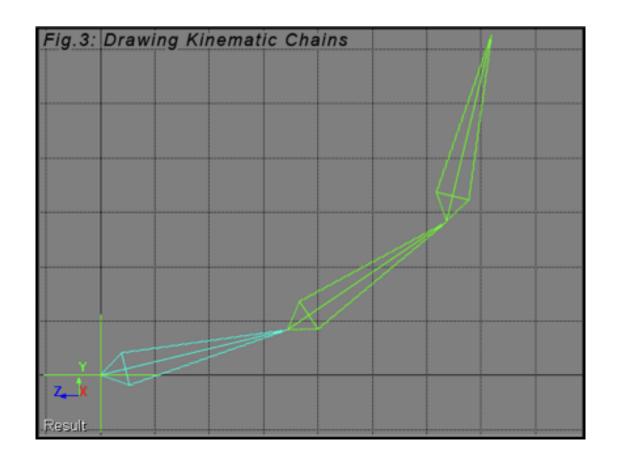
- Starting with the next link's object:
  - Translate the object's base to the origin...
  - ... and rotate by the object's angle



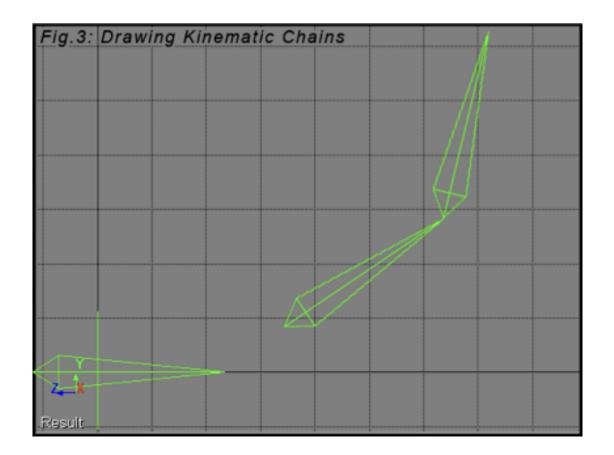
- Starting with the next link's object:
  - Translate by the next object's length...



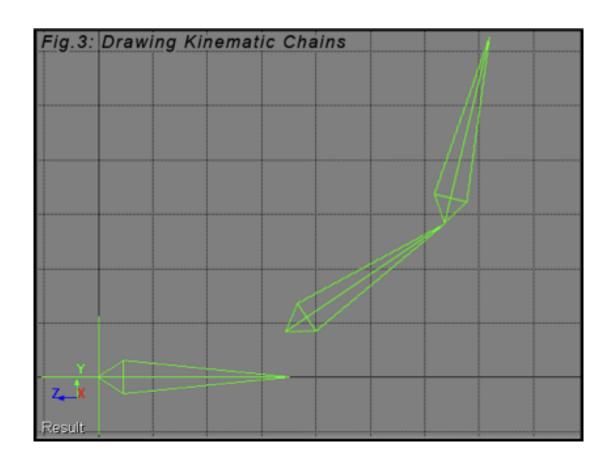
- Starting with the next link's object:
  - Translate by the next object's length...
  - ... and rotate the entire chain by that object's angle



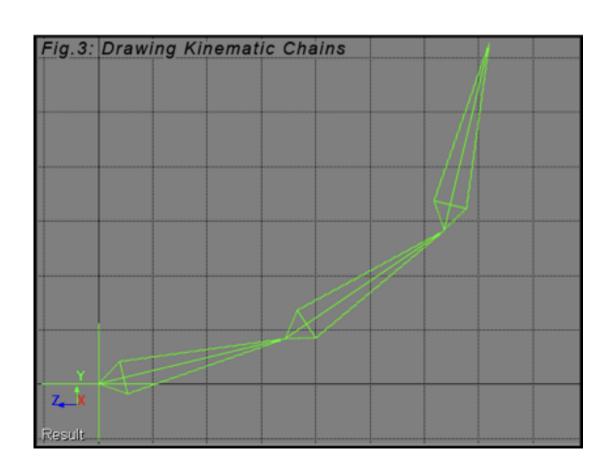
- So on, and so forth...
  - Place the next link



- So on, and so forth...
  - Place the next link
  - Translate



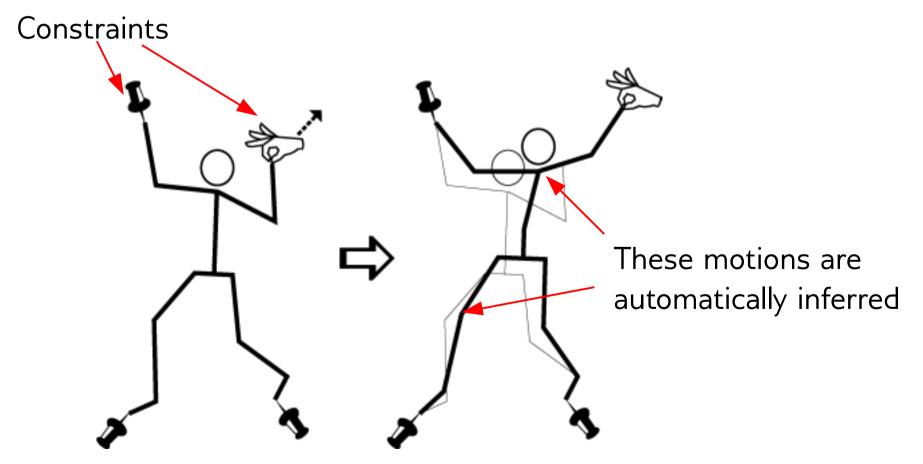
- So on, and so forth, until the chain is complete
  - Place the next link
  - Translate
  - Rotate



• Now that we know what Forward Kinematics (FK) is, what is Inverse Kinematics (IK)?

- Now that we know what Forward Kinematics (FK) is, what is Inverse Kinematics (IK)?
- Inverse Kinematics: Mathematically determining the positions and angles of joints in a flexible, jointed object, given the position and orientation of some subset of the joints (typically the end effectors)

• Now that we know what Forward Kinematics (FK) is, what is Inverse Kinematics (IK)?



- What is IK used for?
  - Originally used in industrial robotics for assembly plants
    - To get the robot to weld this point, how do I have to position all the links in its arm?
  - In computer graphics, IK is typically used for character animation
    - Animator manipulates a few handles (e.g. hands, feet)
    - The system infers the pose of the rest of the skeleton

# Types of IK solutions

- Closed form/analytical solution
  - Calculate sequence of joint angles from the root to the effector, allowing us to determine if a solution is even possible
  - By parametrizing the solution, we can get a range of feasible configurations
  - For a two link chain the solution is (almost) unique:

$$\Theta_{2} = \cos^{-1} \left( \frac{x^{2} + y^{2} - L_{1}^{2} - L_{2}^{2}}{2L_{1}L_{2}} \right)$$
$$\Theta_{1} = \tan^{-1} \left( \frac{-L_{2}\sin\Theta_{2}x + (L_{1} + L_{2}\cos\Theta_{2})y}{L_{2}\sin\Theta_{2}y + (L_{1} + L_{2}\cos\Theta_{2})x} \right)$$

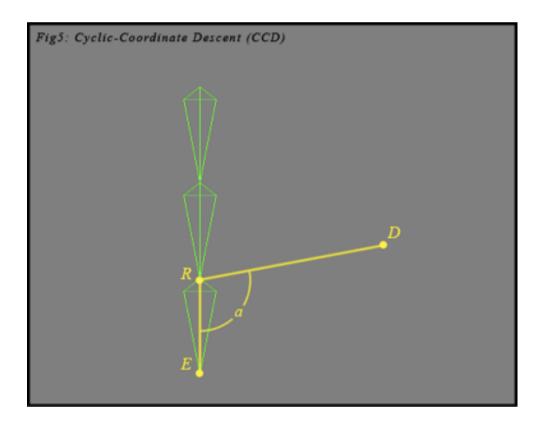
# Types of IK solutions

- Closed form/analytical solution
  - Relatively simple solution for smaller problems
  - However, as the chain increases in length, each new element adds new degrees of freedom, and the problem quickly becomes complex

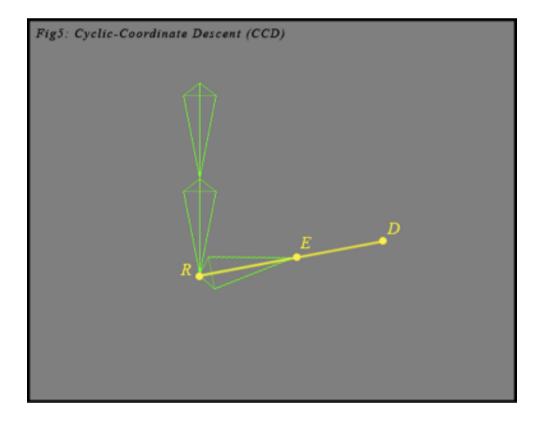
# Types of IK solutions

- Numerical solutions
  - Suitable for complex linkages
  - Iteratively improve solution, progressing towards goal configuration
  - Cyclic Coordinate Descent (this class)
  - Inverse Jacobian Method (next class)

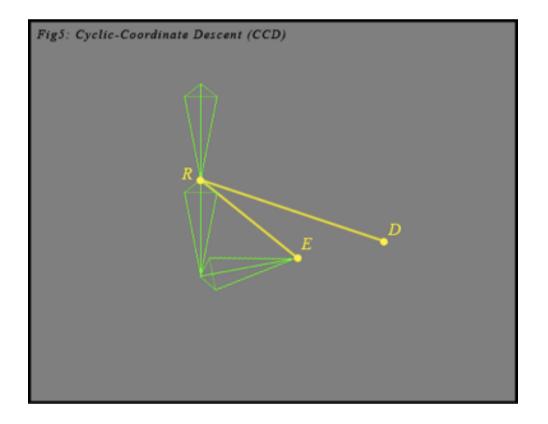
- Start with a vector from the root of our effector R to the current endpoint E
- Draw a vector from R to the desired endpoint D
- The inverse cosine of the dot product gives the angle between the vectors:  $\cos a = \vec{RD} \cdot \vec{RE}$

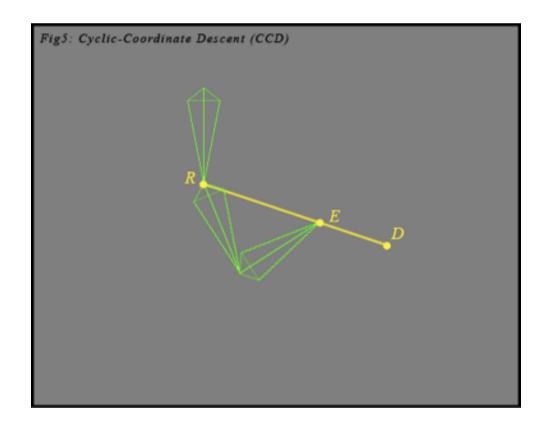


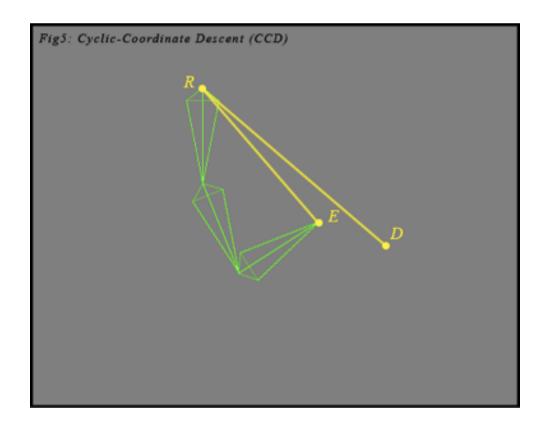
• Rotate our link so that RE falls on RD

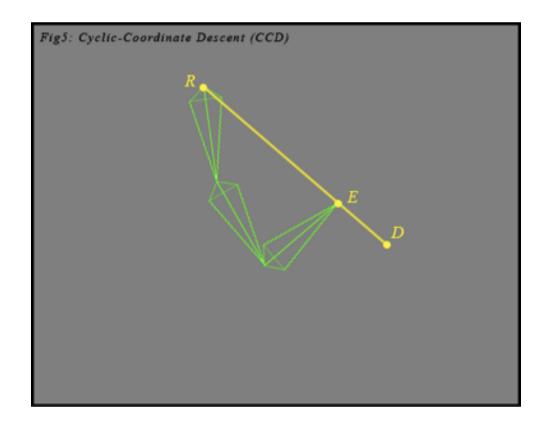


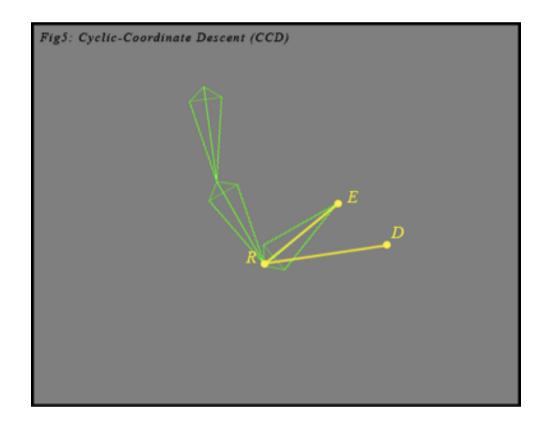
• Move one link up the chain and repeat the process

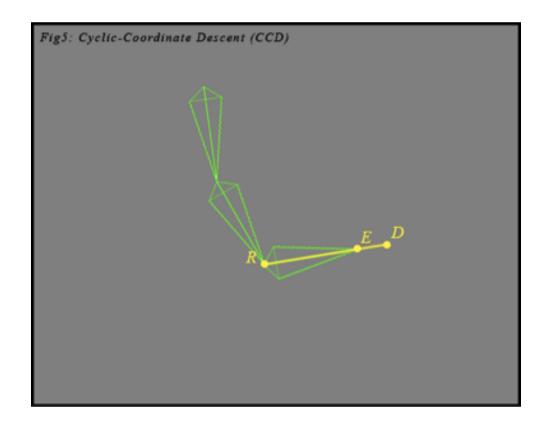


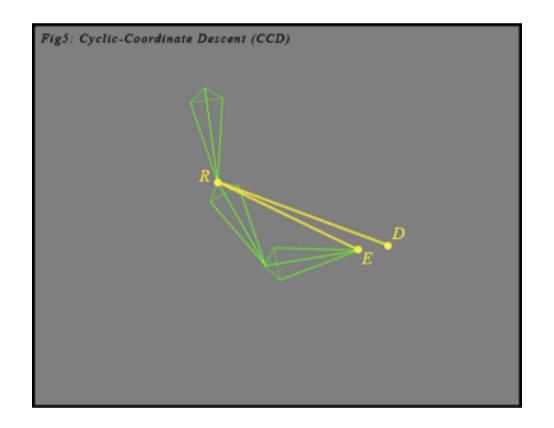


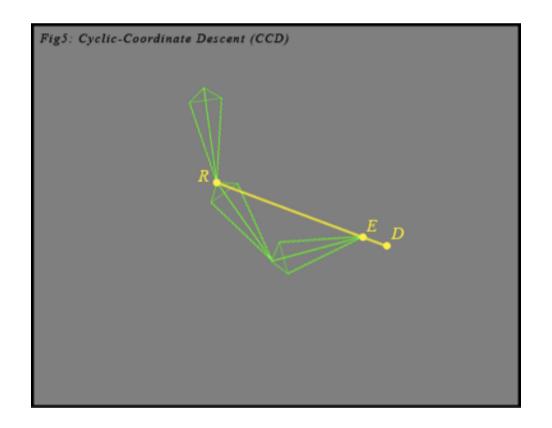


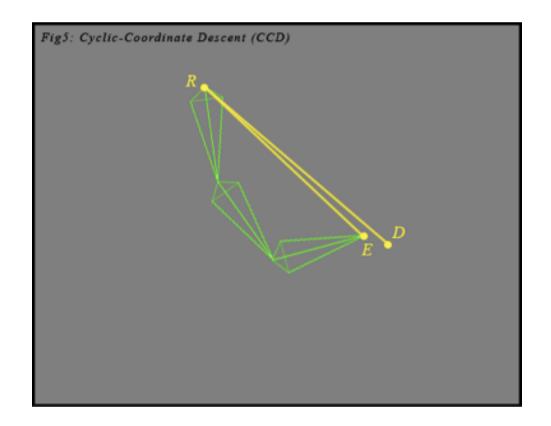


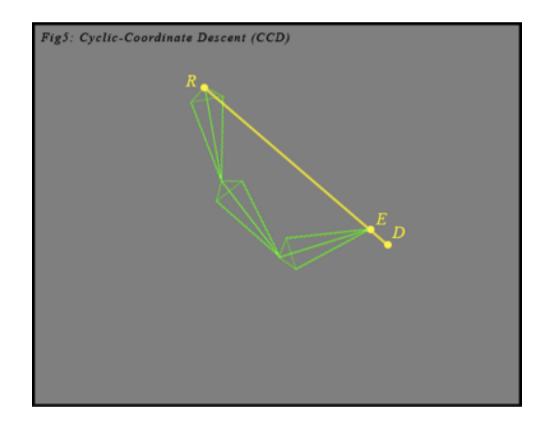










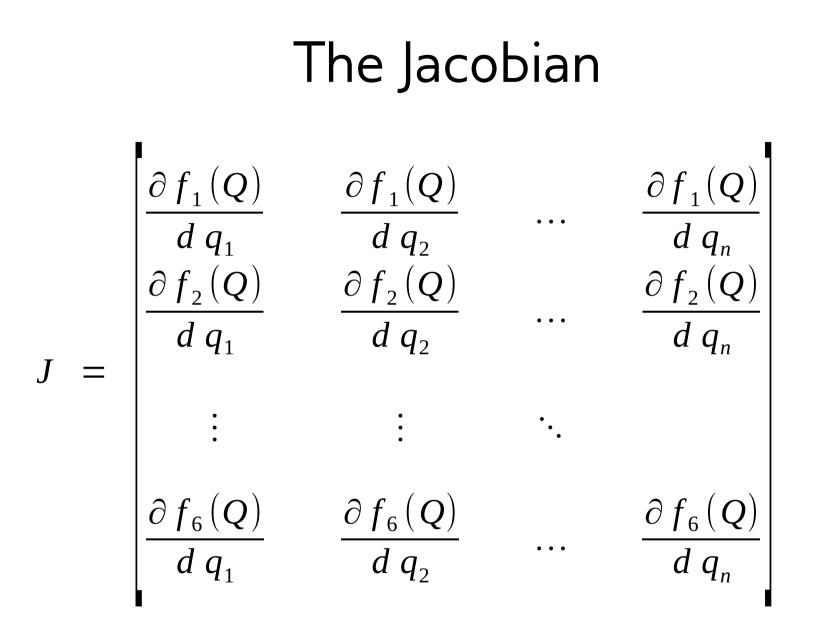


## Inverse Jacobian Method

- *Q*: *n*-vector of internal parameters (joint angles)
- X: 6-vector defining the endpoint's position and orientation
  - ... we can have other target configurations/constraints as well, but we'll currently stick with this for simplicity
- Assume  $n \ge 6$
- Forward kinematics: X = F(Q)

• 
$$dx_i = \frac{\partial f_i(Q)}{d q_1} dq_1 + \frac{\partial f_i(Q)}{d q_2} dq_2 + \dots + \frac{\partial f_i(Q)}{d q_n} dq_n$$

• dX = J dQ



If J is square, dX = J dQ implies  $dQ = J^{-1} dX$ 

#### Case where n = 6

- J is a square 6x6 matrix
- **Problem:** Given *D* (desired endpoint location), find Q such that D = F(Q)
- **Solution**, starting at any initial configuration  $X_0$  with known parameters  $Q_0$ 
  - Interpolate linearly between  $X_0$  and D

- Produces sequence  $X_1, X_2, X_3, \ldots, X_p = D$ 

• For *i* = 1, 2,..., *p* do

- 
$$Q_i = Q_{i-1} + J^{-1}(Q_{i-1})(X_i - X_{i-1})$$

- Reset  $X_1$  to  $F(Q_i)$ 

#### Case where n > 6

- J is a 6 x n matrix
- Recall: we're trying to solve the linear system dX = J dQ
- for dQ, where  $dX = X_i X_{i-1}$ , and J is the Jacobian evaluated at  $Q_{i-1}$
- The system is under-determined!
  - Too few constraints (6)
  - Too many variables (n)
- We can't take the inverse of  ${\cal J}$

#### Non-square linear system

- Assume we're trying to solve Ax = b
  ... but A is not square!
- So we'll add some constraints:
  - If *A* has more rows than columns (over-determined), find **x** that minimizes

$$\|A\mathbf{x} - \mathbf{b}\|$$

 If A has more columns than rows (under-determined), find x satisfying Ax = b that minimizes

#### Enter the pseudoinverse

- If A has more columns than rows (underdetermined), find x satisfying Ax = b that minimizes || x ||
- It turns out that this solution is given by

 $\mathbf{x} = A^+ \mathbf{b}$ 

where  $A^+$  is the **pseudoinverse**  $A^T(AA^T)^{-1}$  of A

(we'll assume A is full rank so this pseudoinverse formula works)

#### Why least-norm solution?

Let  $\mathbf{x}^*$  be the pseudoinverse solution  $A^T(AA^T)^{-1}\mathbf{b}$ 

Proof that it is a solution:

• 
$$A \mathbf{x}^* = A A^T (AA^T)^{-1} \mathbf{b} = (A A^T) (AA^T)^{-1} \mathbf{b} = I \mathbf{b} = \mathbf{b}$$

## Why least-norm solution?

#### Proof that it is least-norm:

- Consider any solution **x** of A**x** = **b**
- ... we have  $A(\mathbf{x} \mathbf{x}^*) = \mathbf{b} \mathbf{b} = 0$

• 
$$(\mathbf{x} - \mathbf{x}^*)^{\mathrm{T}} \mathbf{x}^* = (\mathbf{x} - \mathbf{x}^*)^{\mathrm{T}} A^{\mathrm{T}} (AA^{\mathrm{T}})^{-1} \mathbf{b}$$
  
=  $(A(\mathbf{x} - \mathbf{x}^*))^{\mathrm{T}} (AA^{\mathrm{T}})^{-1} \mathbf{b}$   
=  $0^{\mathrm{T}} (AA^{\mathrm{T}})^{-1} \mathbf{b}$   
=  $0$ 

- ... so  $x x^*$  and  $x^*$  are orthogonal
- Hence,  $||\mathbf{x}||^2 = ||\mathbf{x}^* + \mathbf{x} \mathbf{x}^*||^2 = ||\mathbf{x}^*||^2 + ||\mathbf{x} \mathbf{x}^*||^2 \ge ||\mathbf{x}^*||^2$

# Derivation via Lagrange multipliers

- **Problem:** minimize  $||\mathbf{x}||^2 = \mathbf{x}^T \mathbf{x}$ subject to  $A\mathbf{x} = \mathbf{b}$
- Solution: Introduce Lagrange multiplier  $\lambda$  $L(\mathbf{x}, \lambda) = \mathbf{x}^{\mathrm{T}}\mathbf{x} + \lambda^{\mathrm{T}}(A\mathbf{x} - \mathbf{b})$

Hence the minimum of L is also a solution of the linear system

• At the minimum of *L*, we have

 $\nabla_{\mathbf{x}} L = 2\mathbf{x} + A^{\mathrm{T}} \lambda = 0$  and  $\nabla_{\lambda} L = A\mathbf{x} - \mathbf{b} = 0$ 

- From first condition, we have  $x=-\!\!A^{\mathrm{T}}\,\lambda\,/\,2$
- Substituting into second condition,  $\lambda = -2(AA^{T})^{-1}\mathbf{b}$

• Hence, 
$$\mathbf{x} = A^{T}(AA^{T})^{-1}\mathbf{b}$$

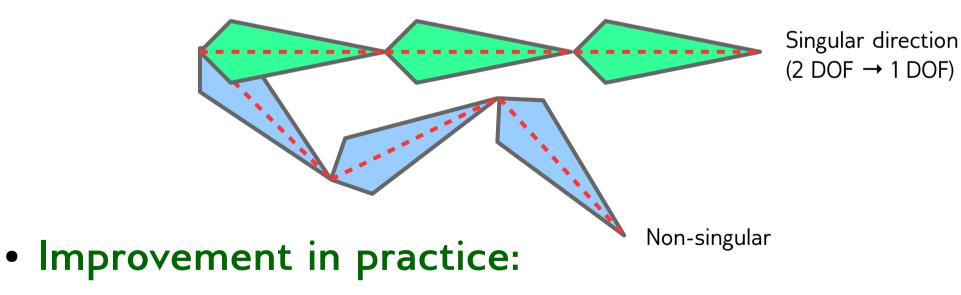
## Inverse Jacobian Method

- Need to reach some target configuration D from initial configuration  $X_{\!_0}$
- General Algorithm:
  - Interpolate linearly between  $X_0$  and D
    - Produces sequence  $X_1, X_2, X_3, \dots, X_p = D$
  - For *i* = 1, 2,..., *p* do
    - $-Q_{i} = Q_{i-1} + J^{+}(Q_{i-1})(X_{i} X_{i-1})$
    - Reset  $X_1$  to  $F(Q_i)$

#### Inverse Jacobian Method

#### • Disadvantages:

- Slow to compute inverse of  $AA^{T}$
- Instability around singularities (J loses full rank)



- Use the Jacobian transpose  $J^{\rm T}$  instead of  $J^{\scriptscriptstyle +}$ 

#### Jacobian Transpose Method

• Replace

 $\mathrm{d}Q = J^+ \,\mathrm{d}X$ 

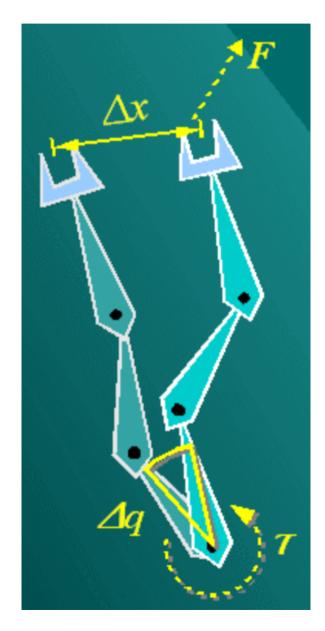
with

 $dQ = J^{\mathrm{T}} dX$ 

• Why does this work?!!!

# Principle of Virtual Work

- "External work = internal work"
- External work = force \* distance
- Internal Work = torque \* angle  $\mathbf{f}^{\mathrm{T}} \mathrm{d}X = \mathbf{\tau}^{\mathrm{T}} \mathrm{d}Q$
- dX = J dQ (forward kinematics)
- $\mathbf{f}^T J dQ = \mathbf{\tau}^T dQ$  (substituting)
- $\mathbf{f}^T J = \boldsymbol{\tau}^T$  (holds for any dQ)
  - i.e.  $\mathbf{\tau} = J^{\mathrm{T}} \mathbf{f}$



# Jacobian Transpose Method

• Virtual work equation:

$$\mathbf{\tau} = J^{\mathrm{T}} \mathbf{f}$$

• Compare with:

 $\mathrm{d}Q = J^{\mathrm{T}} \,\mathrm{d}X$ 

- We're taking the distance to the goal to be a force that pulls our end effector
- With  $J^+$ , we had an exact solution to linearized problem
- ... now no longer

#### Jacobian Transpose Method

- $dQ = J^T dX$  is not exact, but has the right trend
- Throw in a scaling factor *h* and iterate

$$(\Delta Q)_{i+1} = h J^{\mathrm{T}} (\Delta X)_i$$

• h can be thought of as a timestep  $\Delta t$ 

$$\frac{\Delta Q}{\Delta t} = J^T \Delta X$$

• So we're just solving the differential equation

$$\frac{dQ}{dt} = J^T X = J^T F(Q)$$