

# Inverse Kinematics

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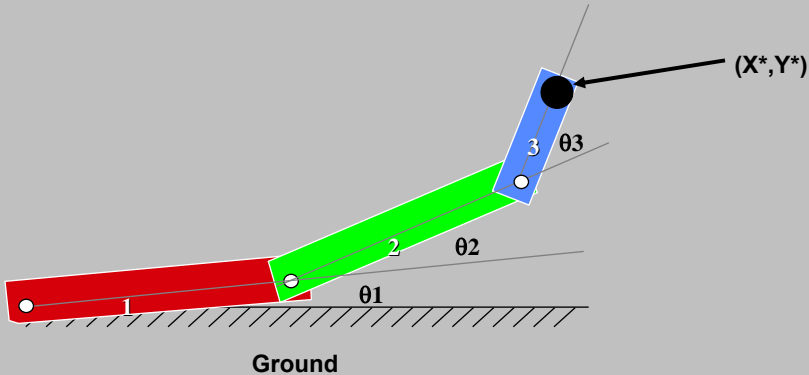
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## Inverse Kinematics

Forward Kinematics solves the problem “if I know the link transformation parameters, where are the links?”.

Inverse Kinematics (IK) solves the problem “If I know where I want the parts to be, what part transformation parameters will put them there?”



## Inverse Kinematics

Note that you can differentiate the equations

$$X^* = f(\Theta_1, \Theta_2, \Theta_3) \text{ and } Y^* = g(\Theta_1, \Theta_2, \Theta_3)$$

If you have a good guess for  $(\Theta_1, \Theta_2, \Theta_3)$  which currently produce an  $X^*$  and  $Y^*$  that are not quite what you want, then you can refine your values for  $(\Theta_1, \Theta_2, \Theta_3)$  and try again. The fundamental equations for this are:

$$X^* - X^*_{approx} = \Delta X = \frac{\partial X^*}{\partial \theta_1} \Delta \theta_1 + \frac{\partial X^*}{\partial \theta_2} \Delta \theta_2 + \frac{\partial X^*}{\partial \theta_3} \Delta \theta_3$$

$$Y^* - Y^*_{approx} = \Delta Y = \frac{\partial Y^*}{\partial \theta_1} \Delta \theta_1 + \frac{\partial Y^*}{\partial \theta_2} \Delta \theta_2 + \frac{\partial Y^*}{\partial \theta_3} \Delta \theta_3$$



## Inverse Kinematics

Or, in matrix form:

$$\begin{bmatrix} \frac{\partial X^*}{\partial \theta_1} & \frac{\partial X^*}{\partial \theta_2} & \frac{\partial X^*}{\partial \theta_3} \\ \frac{\partial Y^*}{\partial \theta_1} & \frac{\partial Y^*}{\partial \theta_2} & \frac{\partial Y^*}{\partial \theta_3} \end{bmatrix} \begin{Bmatrix} \Delta \theta_1 \\ \Delta \theta_2 \\ \Delta \theta_3 \end{Bmatrix} = \begin{Bmatrix} \Delta X \\ \Delta Y \end{Bmatrix}$$

2 × 3
3 × 1
2 × 1

The matrix:

$$\begin{bmatrix} \frac{\partial X^*}{\partial \theta_1} & \frac{\partial X^*}{\partial \theta_2} & \frac{\partial X^*}{\partial \theta_3} \\ \frac{\partial Y^*}{\partial \theta_1} & \frac{\partial Y^*}{\partial \theta_2} & \frac{\partial Y^*}{\partial \theta_3} \end{bmatrix}$$

is called the Jacobian, and is abbreviated as  $[J]$ :

$$[J] \begin{Bmatrix} \Delta \theta_1 \\ \Delta \theta_2 \\ \Delta \theta_3 \end{Bmatrix} = \begin{Bmatrix} \Delta X \\ \Delta Y \end{Bmatrix}$$



### Inverse Kinematics

Note that  $[J]$  is not a square matrix, so this system of equations cannot be solved for directly. But, if we pre-multiply by the transpose of  $[J]$ , we get:

$$([J]^T [J]) \begin{Bmatrix} \Delta\theta_1 \\ \Delta\theta_2 \\ \Delta\theta_3 \end{Bmatrix} = [J]^T \begin{Bmatrix} \Delta X \\ \Delta Y \end{Bmatrix}$$

$$(3 \times 2 \quad 2 \times 3) \quad 3 \times 1 \quad (3 \times 2 \quad 2 \times 1)$$

$$3 \times 3 \quad 3 \times 1 \quad 3 \times 1$$

which is solvable.



### Inverse Kinematics

$$\begin{Bmatrix} \Delta\theta_1 \\ \Delta\theta_2 \\ \Delta\theta_3 \end{Bmatrix} = ([J]^T [J])^{-1} [J]^T \begin{Bmatrix} \Delta X \\ \Delta Y \end{Bmatrix}$$

$$3 \times 1 \quad 3 \times 3 \quad 3 \times 2 \quad 2 \times 1$$

It is not obvious, but this is the *Least Squares* formulation. It will give an optimum  $(\Delta\theta_1, \Delta\theta_2, \Delta\theta_3)$  to make  $(X^*, Y^*)$  move closer to the desired values.



## Inverse Kinematics

Differentiate the equation  $X^* = f(\theta_1, \theta_2, \theta_3)$  and  $Y^* = g(\theta_1, \theta_2, \theta_3)$

Pick a starting  $\theta_1, \theta_2$ , and  $\theta_3$

1. Compute  $X^*_{\text{approx}}$  and  $Y^*_{\text{approx}}$  from  $\theta_1, \theta_2, \theta_3$

2. Compute  $\Delta X = X^* - X^*_{\text{approx}}$  and  $\Delta Y = Y^* - Y^*_{\text{approx}}$

3. If  $\Delta X$  and  $\Delta Y$  are "small enough", we're done

4. Compute:  $\frac{\partial X^*}{\partial \theta_1}, \frac{\partial X^*}{\partial \theta_2}, \frac{\partial X^*}{\partial \theta_3}, \frac{\partial Y^*}{\partial \theta_1}, \frac{\partial Y^*}{\partial \theta_2}, \frac{\partial Y^*}{\partial \theta_3}$

5. Form the Jacobian [ J ]

6. Solve the system of equations: 
$$([J]^T [J]) \begin{Bmatrix} \Delta \theta_1 \\ \Delta \theta_2 \\ \Delta \theta_3 \end{Bmatrix} = [J]^T \begin{Bmatrix} \Delta X \\ \Delta Y \end{Bmatrix}$$

7. Refine:  $\theta_1 = \theta_1 + \Delta \theta_1$  ;  $\theta_2 = \theta_2 + \Delta \theta_2$  ;  $\theta_3 = \theta_3 + \Delta \theta_3$



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## Numerical Differentiation

Rather than explicitly differentiating an equation, it is often easier to calculate the derivative numerically. This is known as the **Central Difference** method, where you look both forward and backwards to see how the dependent variable is changing with respect to the independent variable:

$$\frac{\partial X}{\partial \theta} \approx \frac{X(\theta + \Delta \theta) - X(\theta - \Delta \theta)}{2\Delta \theta}$$

Pick a delta that is small, but not so small that floating point accuracy becomes an issue



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## Numerical Differentiation

If there are other independent variables, hold them constant:

$$\frac{\partial X}{\partial \theta_1} \approx \frac{X(\theta_1 + \Delta\theta_1, \theta_2, \theta_3) - X(\theta_1 - \Delta\theta_1, \theta_2, \theta_3)}{2\Delta\theta_1}$$



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## Numerical Differentiation

In case you ever need it, here is how to compute the second derivative

$$\frac{\partial^2 X}{\partial \theta^2} \approx \frac{X(\theta + \Delta\theta) - 2X(\theta) + X(\theta - \Delta\theta)}{\Delta\theta^2}$$

As before, if there are other independent variables, hold them constant:



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