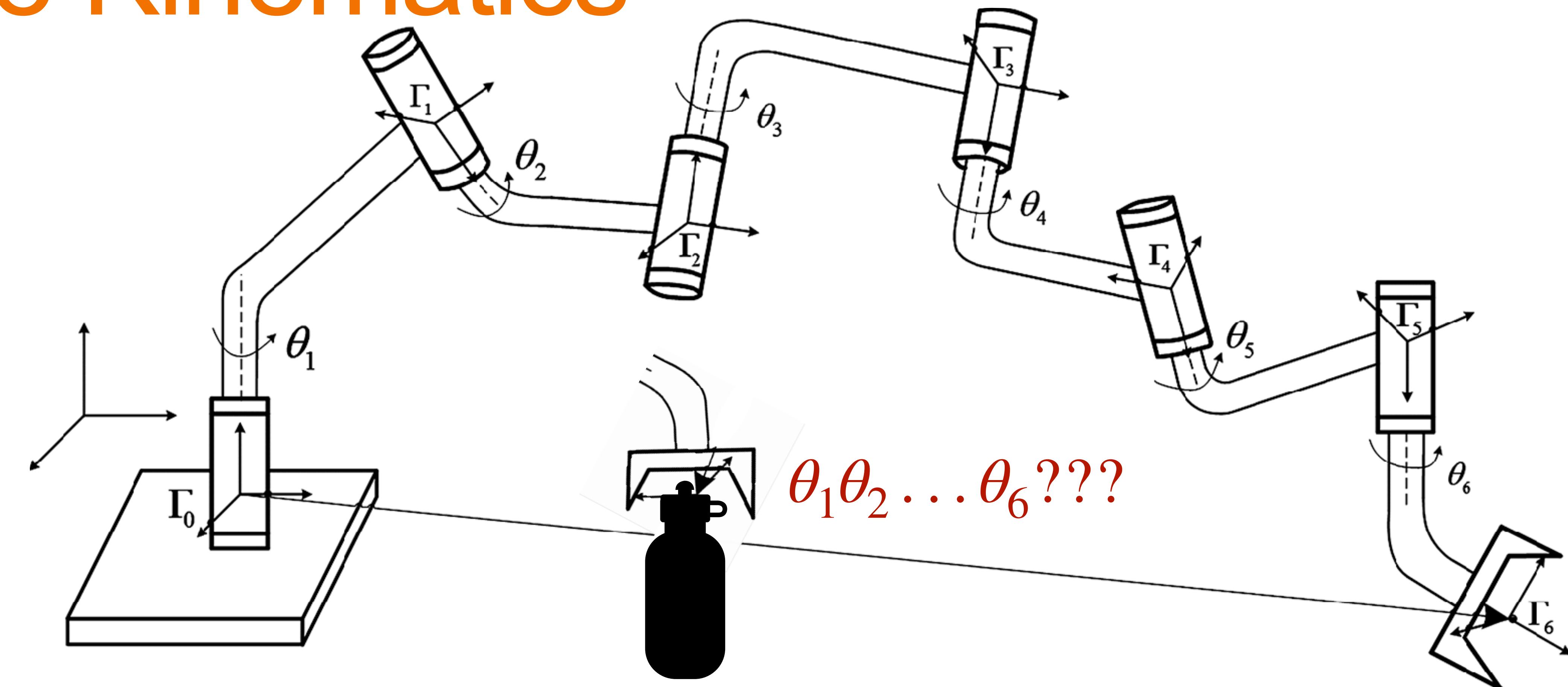


Lecture 08

Manipulation - III

Inverse Kinematics



Course Logistics

- Project 3 was posted on 02/07 and will be due 02/15 (not 02/14).
- Project 4 will be posted on 02/14 and will be due 02/28 (yes 2 weeks).
- Quiz 4 will be posted tomorrow at **noon** (*based on requests on Ed*) and will be due on Wed at noon.



Previously

PID Control

Error signal:

$$e(t) = x_{desired}(t) - x(t)$$

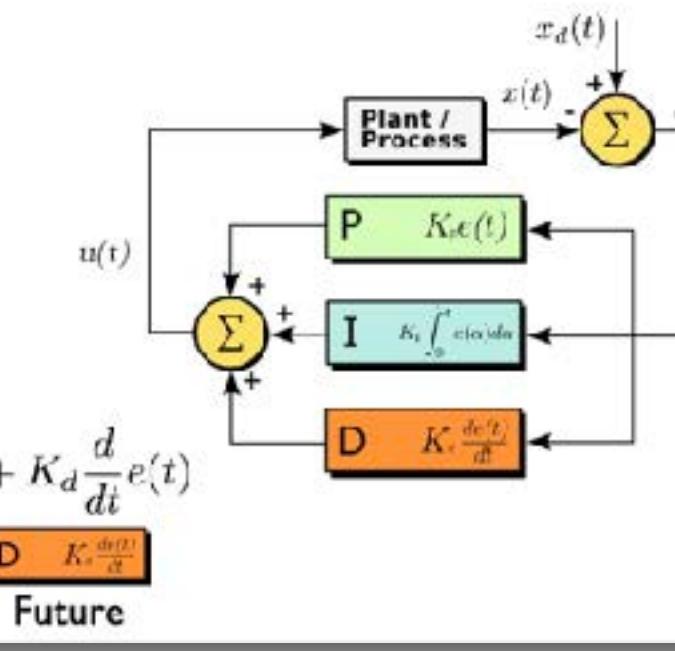
Control signal:

$$u(t) = K_p e(t) + K_i \int_0^t e(\alpha) d\alpha + K_d \frac{d}{dt} e(t)$$

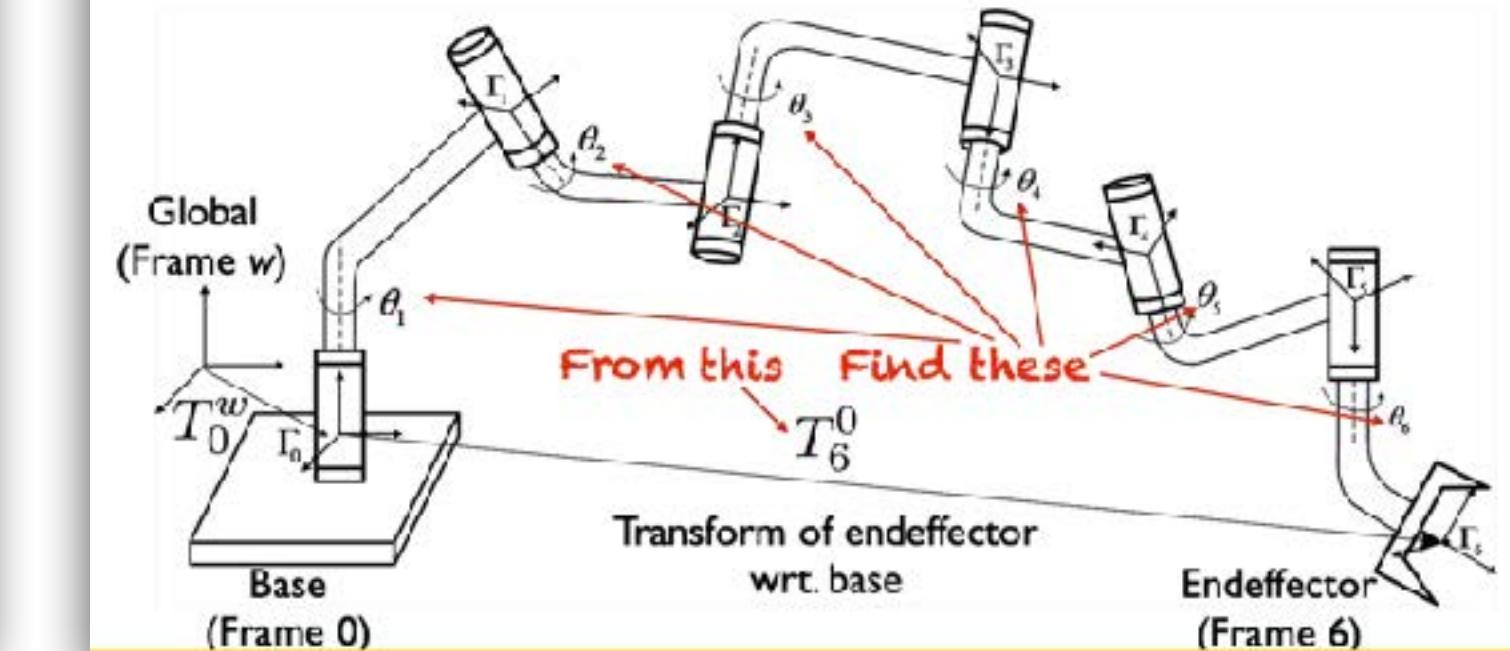
P $K_p e(t)$
Current

I $K_i \int_0^t e(\alpha) d\alpha$
Past

D $K_d \frac{d}{dt} e(t)$
Future



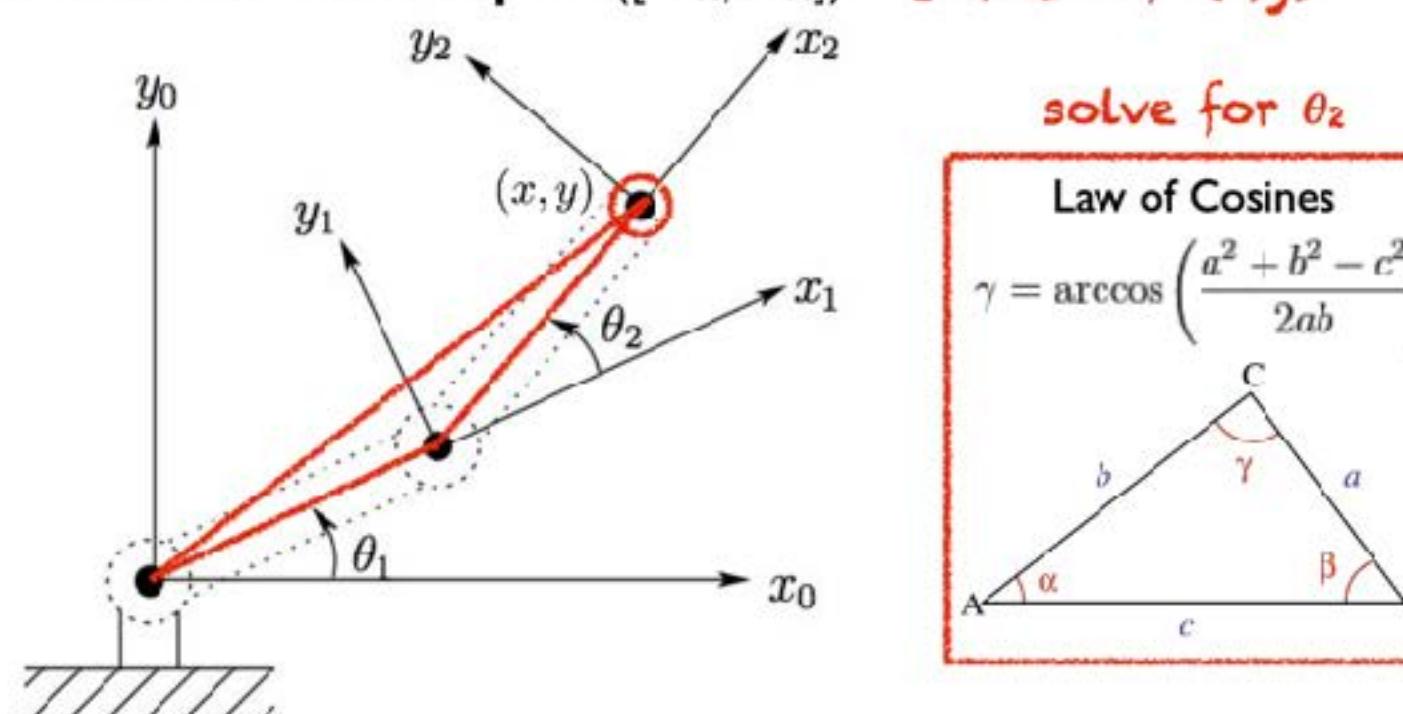
Inverse kinematics: one-to-many mapping of workspace endeffector pose to robot configuration



Inverse kinematics: how to solve for $q = \{\theta_1, \dots, \theta_N\}$ from T^0_N ?

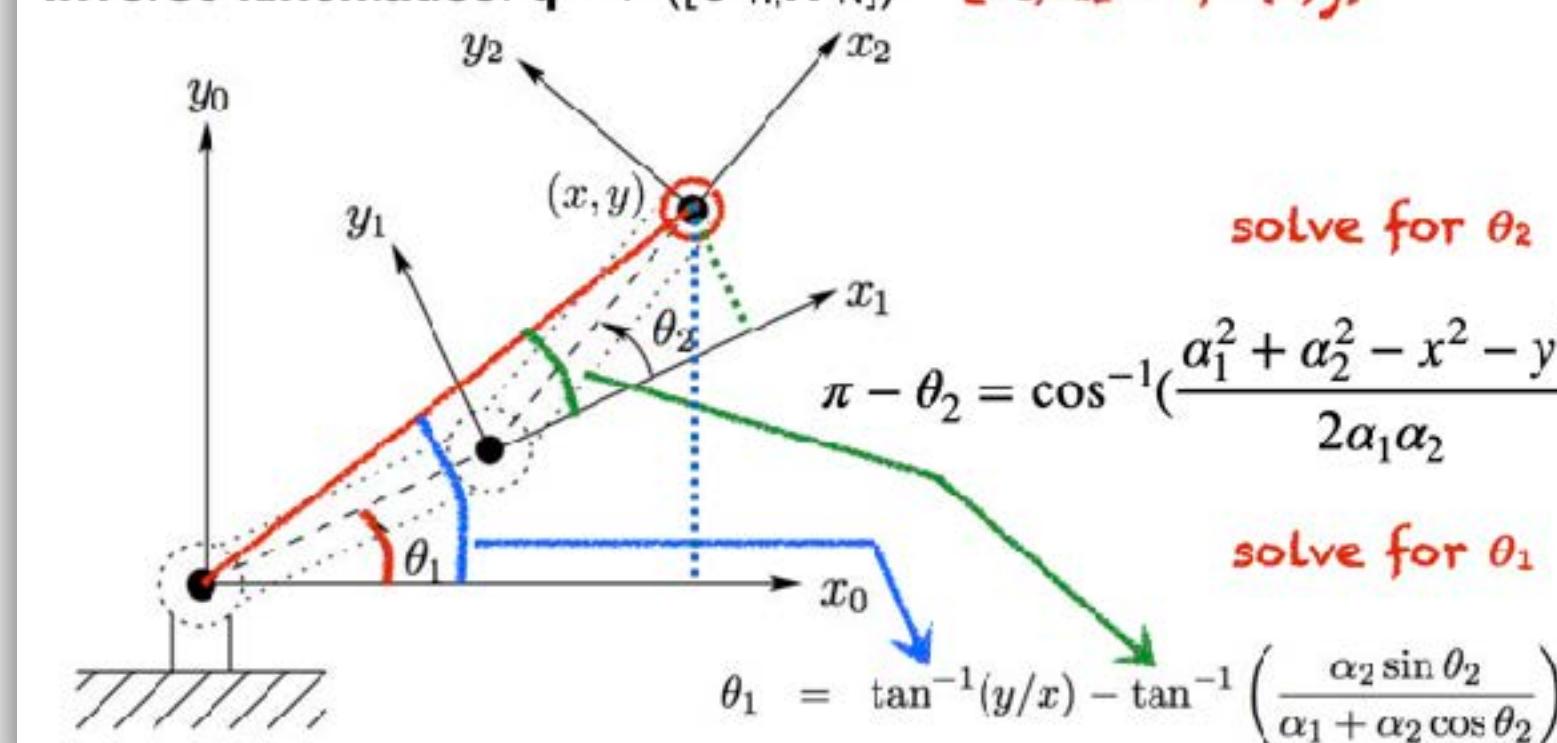
Inverse kinematics: $q = f^{-1}([o^0_N, R^0_N])$

$[\theta_1, \theta_2] = f^{-1}(x, y)$



Inverse kinematics: $q = f^{-1}([o^0_N, R^0_N])$

$[\theta_1, \theta_2] = f^{-1}(x, y)$



Inverse Kinematics: 3D

Configuration

$$q = \begin{bmatrix} q_1 \\ q_2 \\ \vdots \\ q_6 \end{bmatrix}$$

$T_n^0(q_1, \dots, q_n) = H$ ← Transform from endeffector

$$H = \begin{bmatrix} R & o \\ 0 & 1 \end{bmatrix}$$

Closed form solution?

$$q_k = f_k(h_{11}, \dots, h_{34}), \quad k = 1, \dots, n.$$

$$H = \begin{bmatrix} r_{11} & r_{12} & r_{13} & o_x \\ r_{21} & r_{22} & r_{23} & o_y \\ r_{31} & r_{32} & r_{33} & o_z \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

6 DOF position and orientation of endeffector

Why Closed Form?

- Advantages
- Speed: IK solution computed in constant time
- Predictability: consistency in selecting satisfying IK solution
- Disadvantage
- Generality: general form for arbitrary kinematics difficult to express

Closed form solution?

$$q_k = f_k(h_{11}, \dots, h_{34}), \quad k = 1, \dots, n.$$

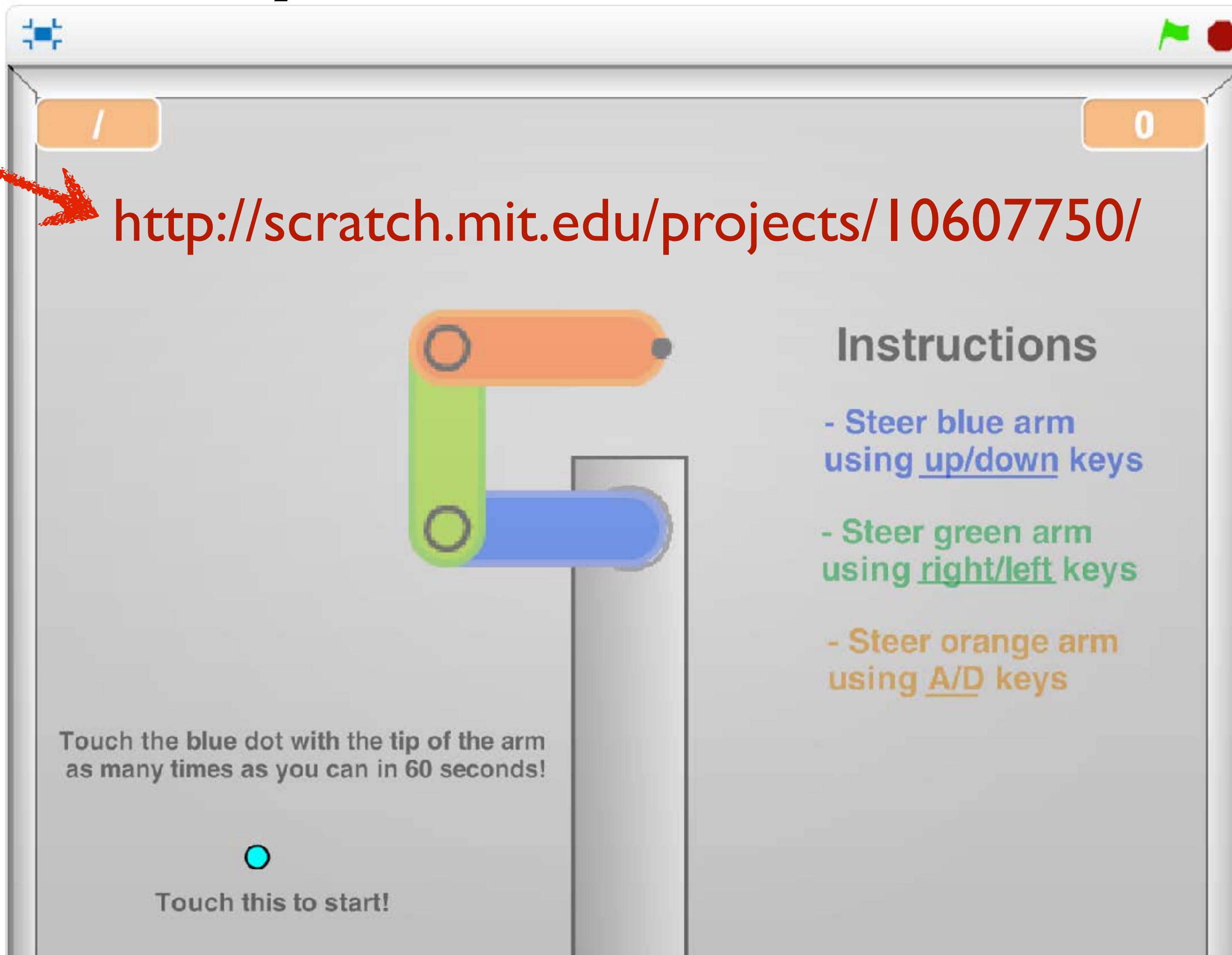
Inverse Kinematics: 2 possibilities

- **Closed-form solution:** geometrically infer satisfying configuration
 - *Speed:* solution often computed in constant time
 - *Predictability:* solution is selected in a consistent manner
- **Solve by optimization:** minimize error of endeffector to desired pose
 - often some form of Gradient Descent (a la Jacobian Transpose)
 - *Generality:* same solver can be used for many different robots

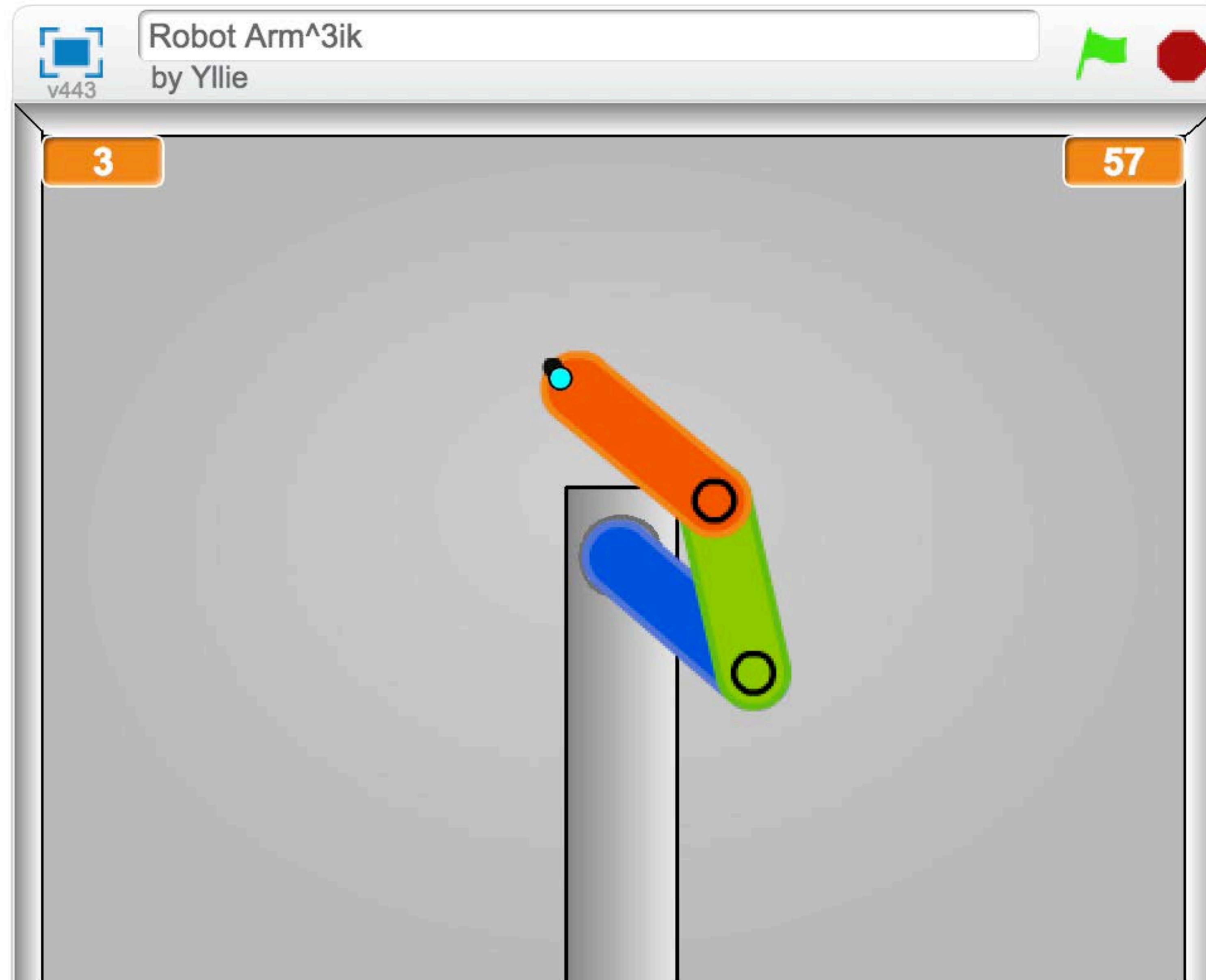


Anyone tried this?

Try this



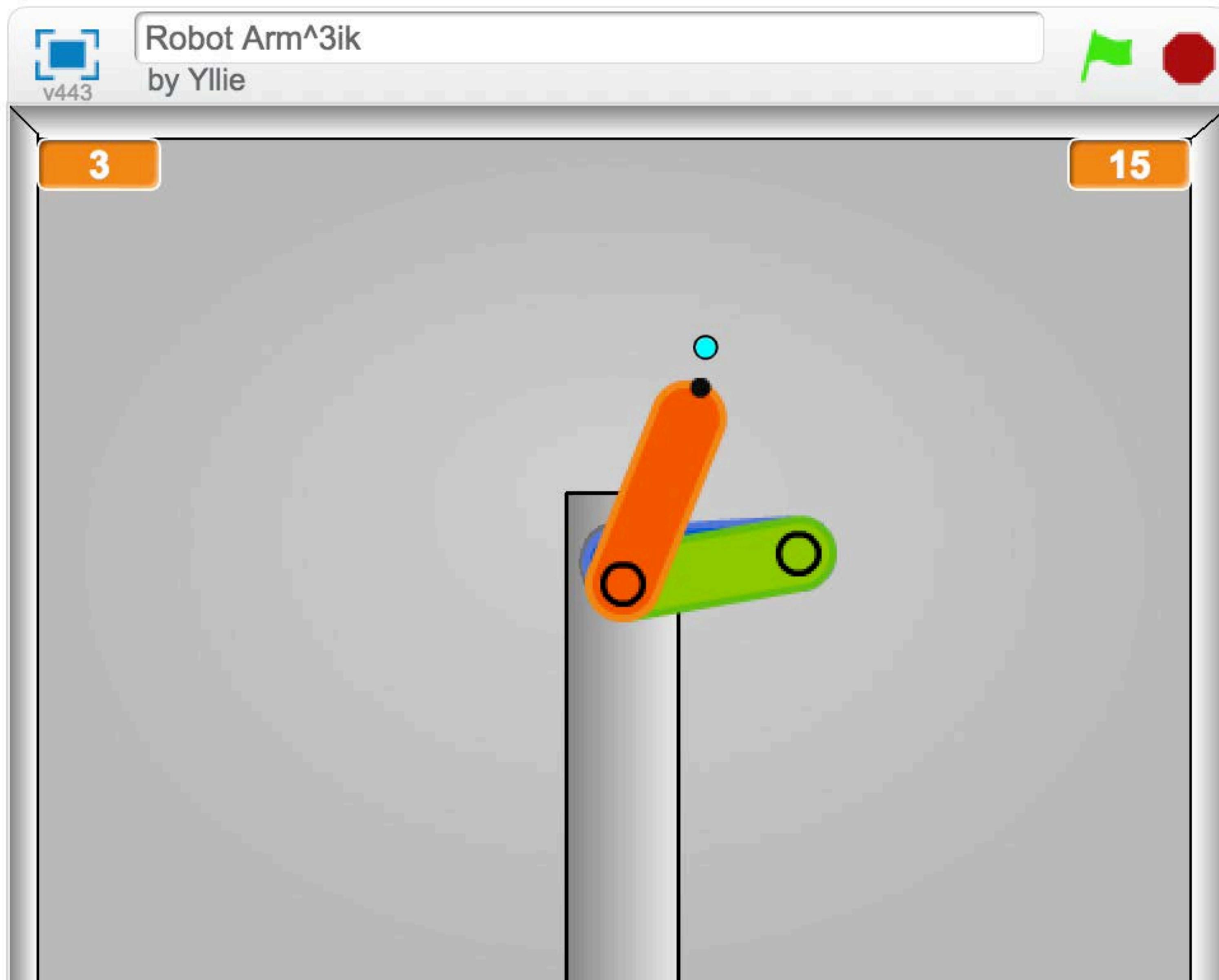
Aggressively tuned IK



By Dr. Jenkins



Conservatively tuned IK



By Dr. Jenkins



How to programmatically do this?



How to programmatically do
this?

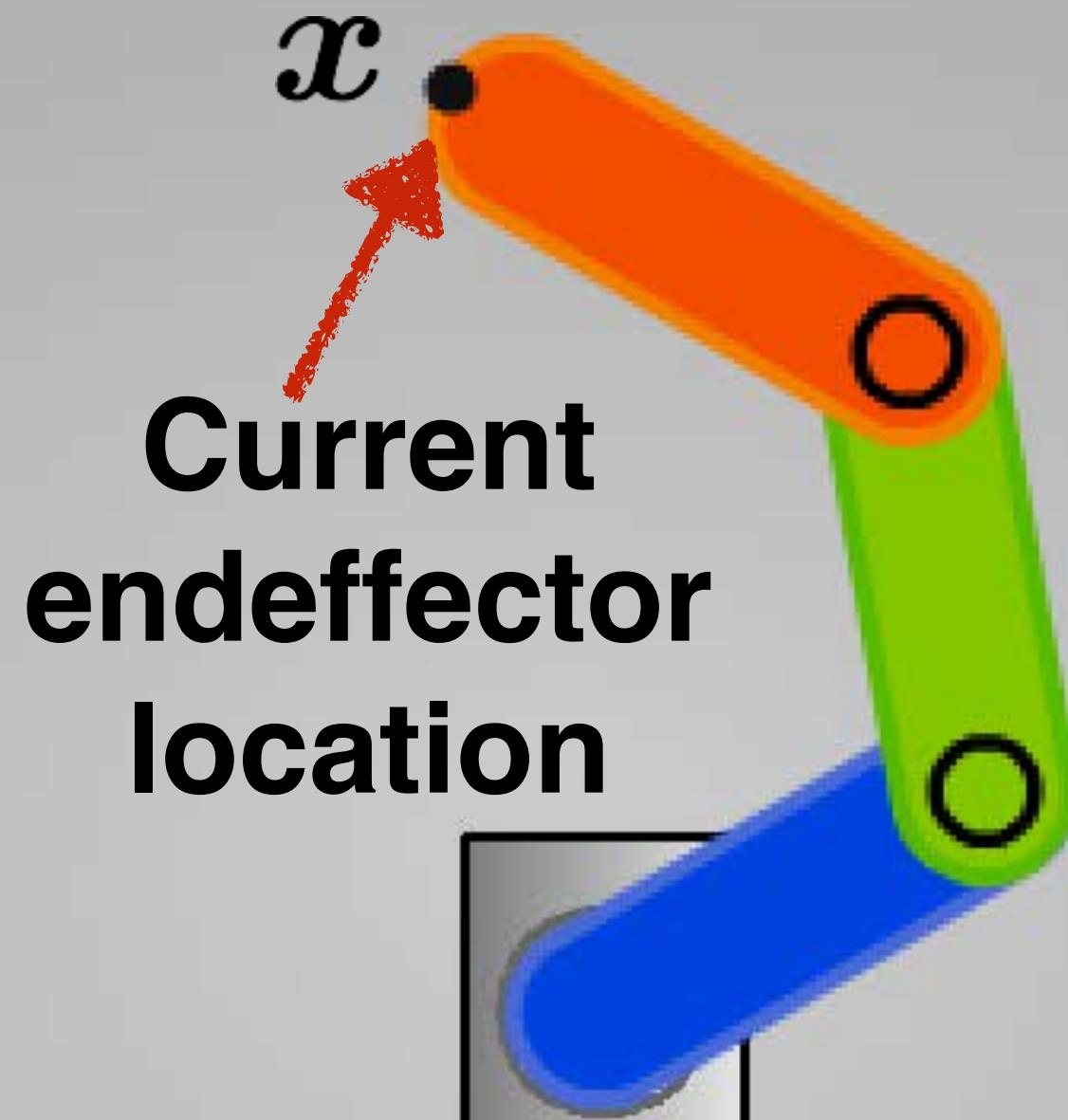
Jacobian Transpose

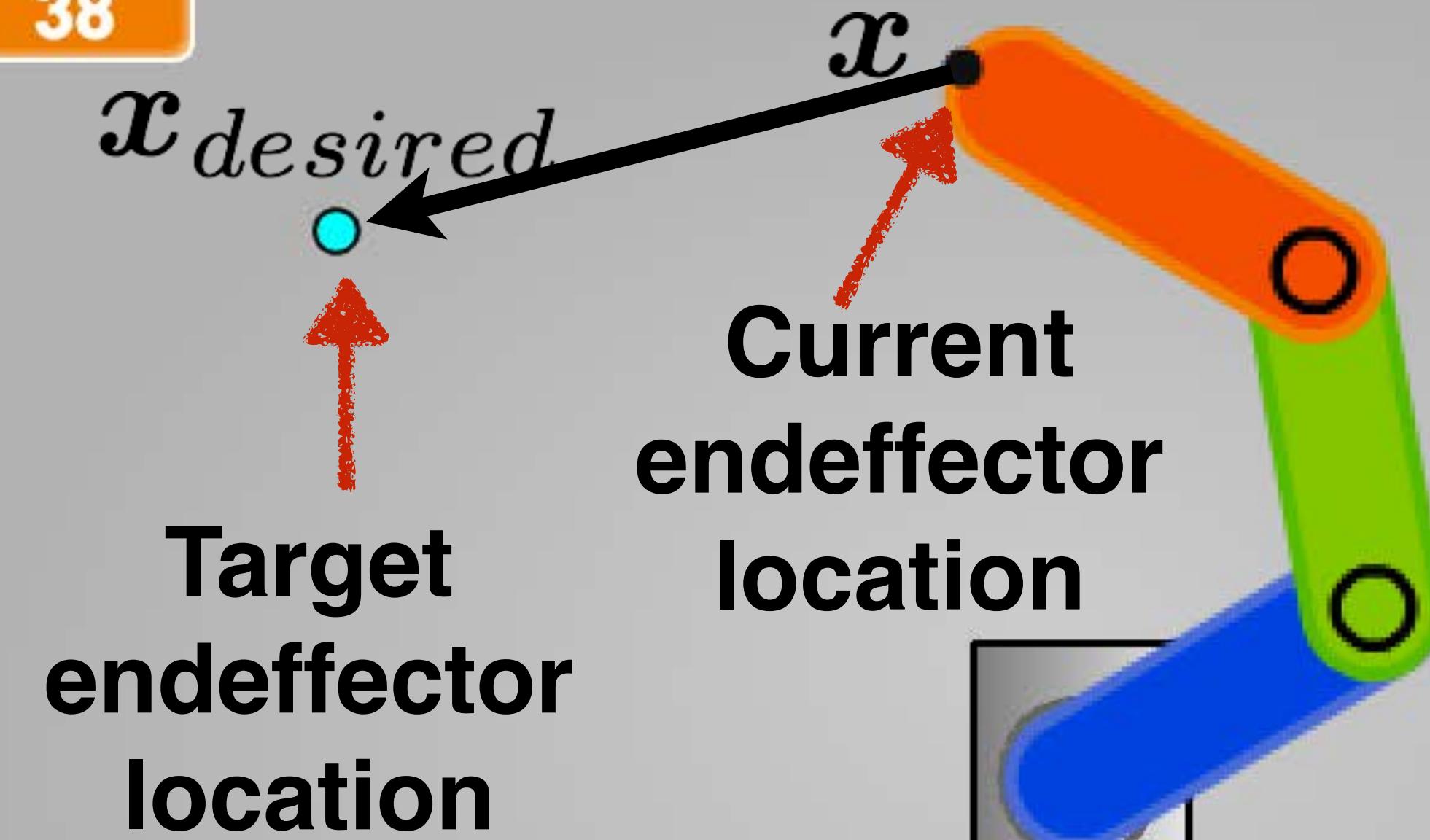


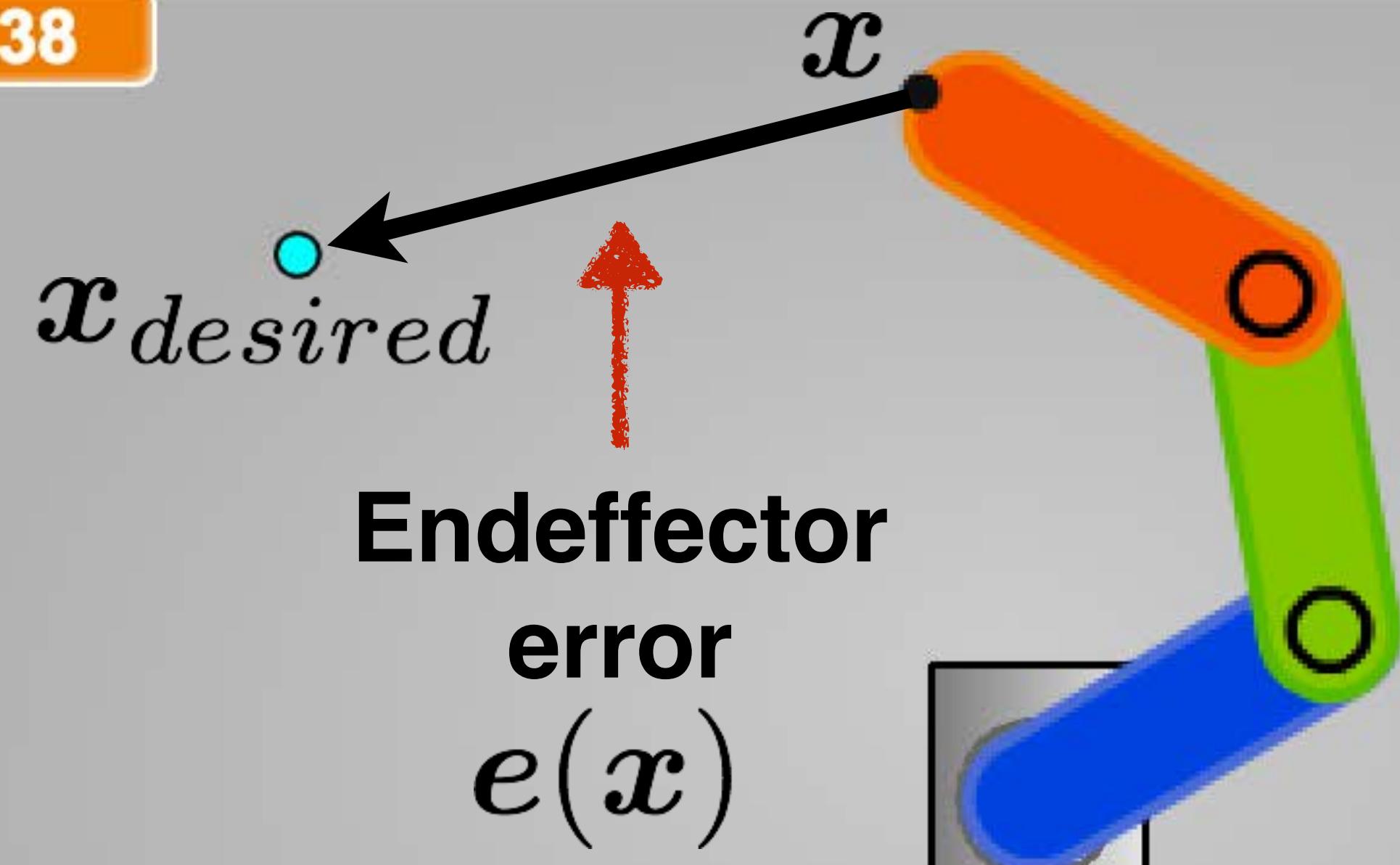
$x_{desired}$



Target
endeffector
location

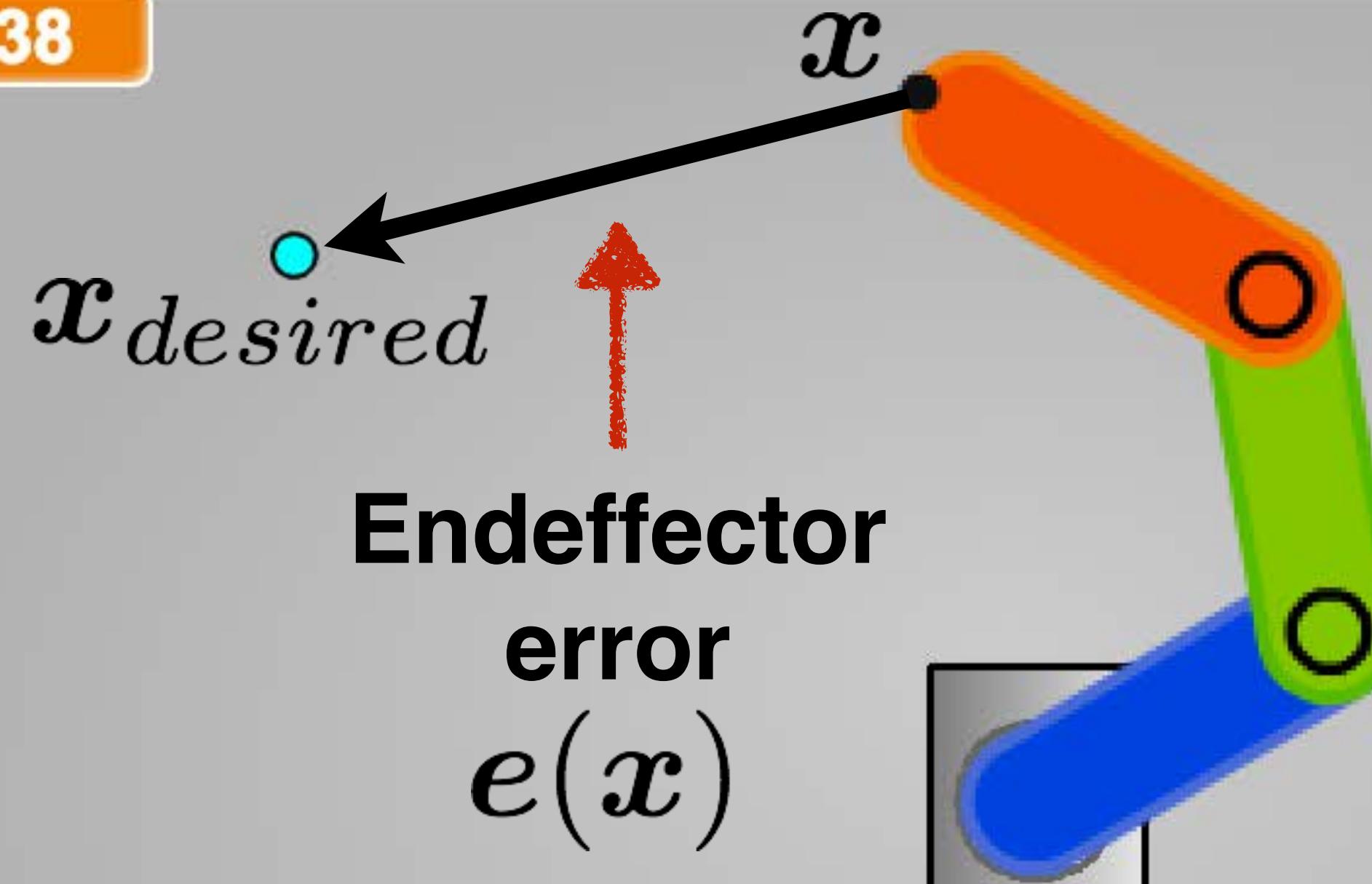






**Endeffector
error**
 $e(x)$

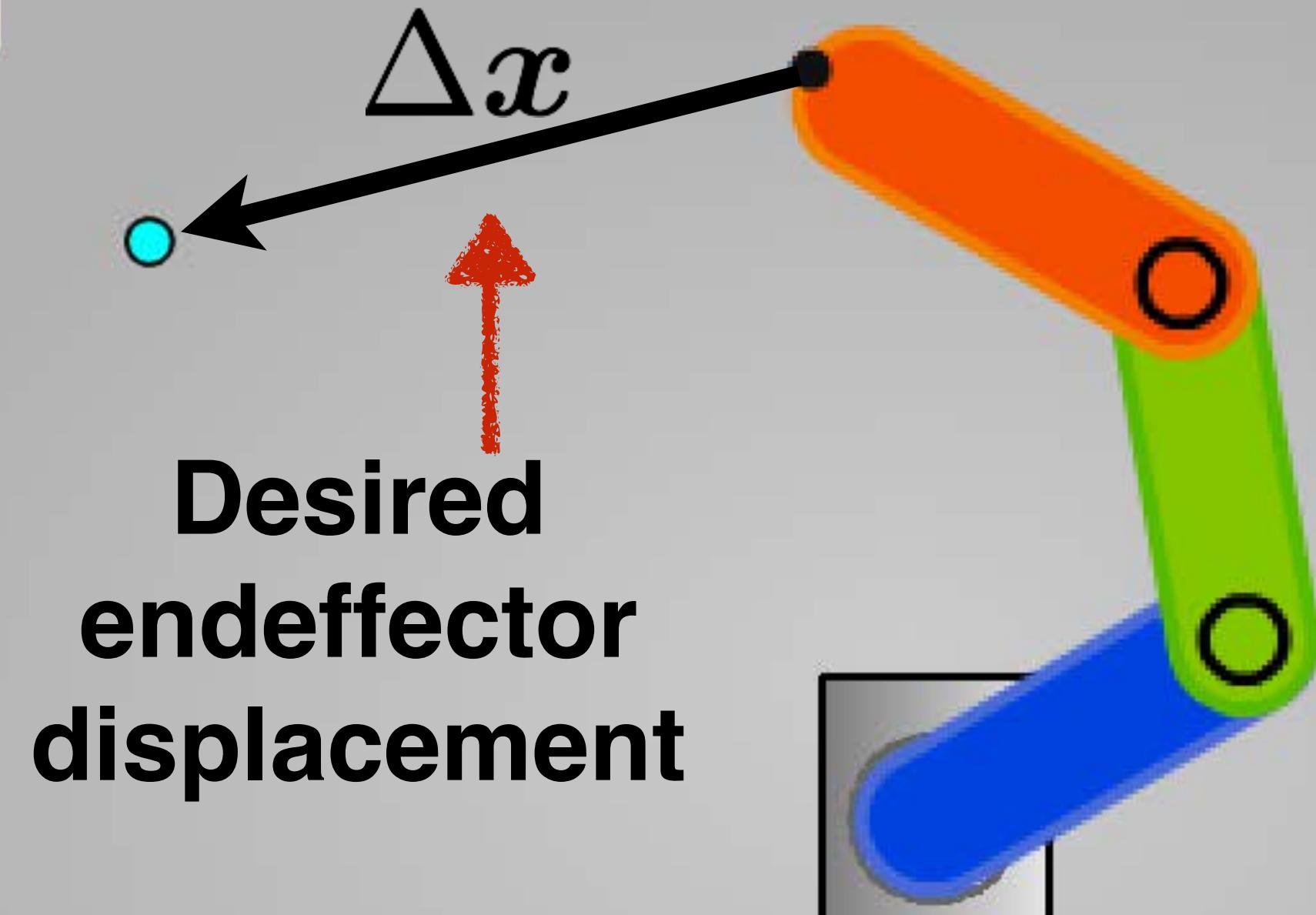
Can we move the
endeffector to
minimize error?



Can we move the
end effector to
minimize error?

Yes!

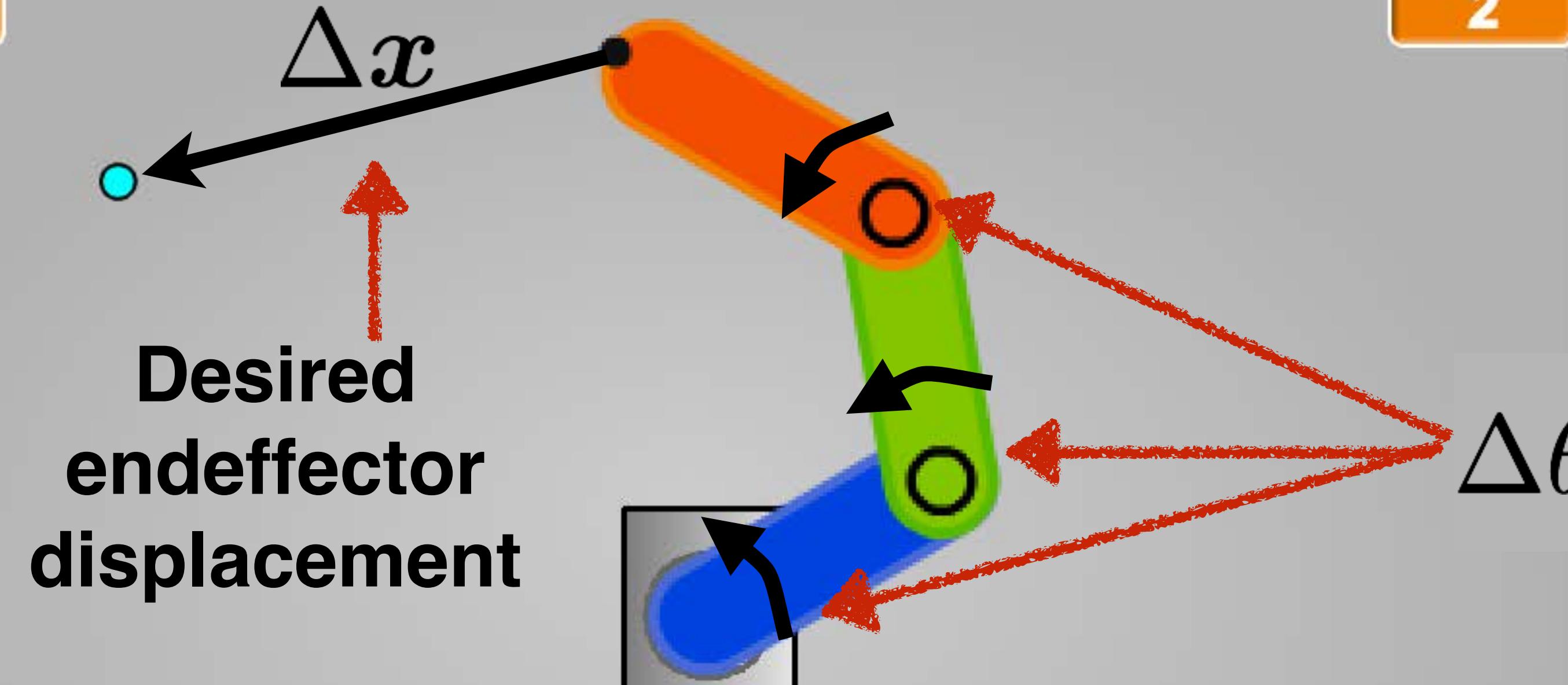
convert linear velocity at end effector
to angular velocities at joints.



Can we move the
endeffector to
minimize error?

Yes!

convert linear velocity at endeffector
to angular velocities at joints.



Desired
endeffector
displacement

Joint angle
displacements

Can we move the
endeffector to
minimize error?

Yes!

convert linear velocity at endeffector
to angular velocities at joints.

How are linear and angular velocity related?

How are linear and angular velocity related?

Consider the velocity of a point

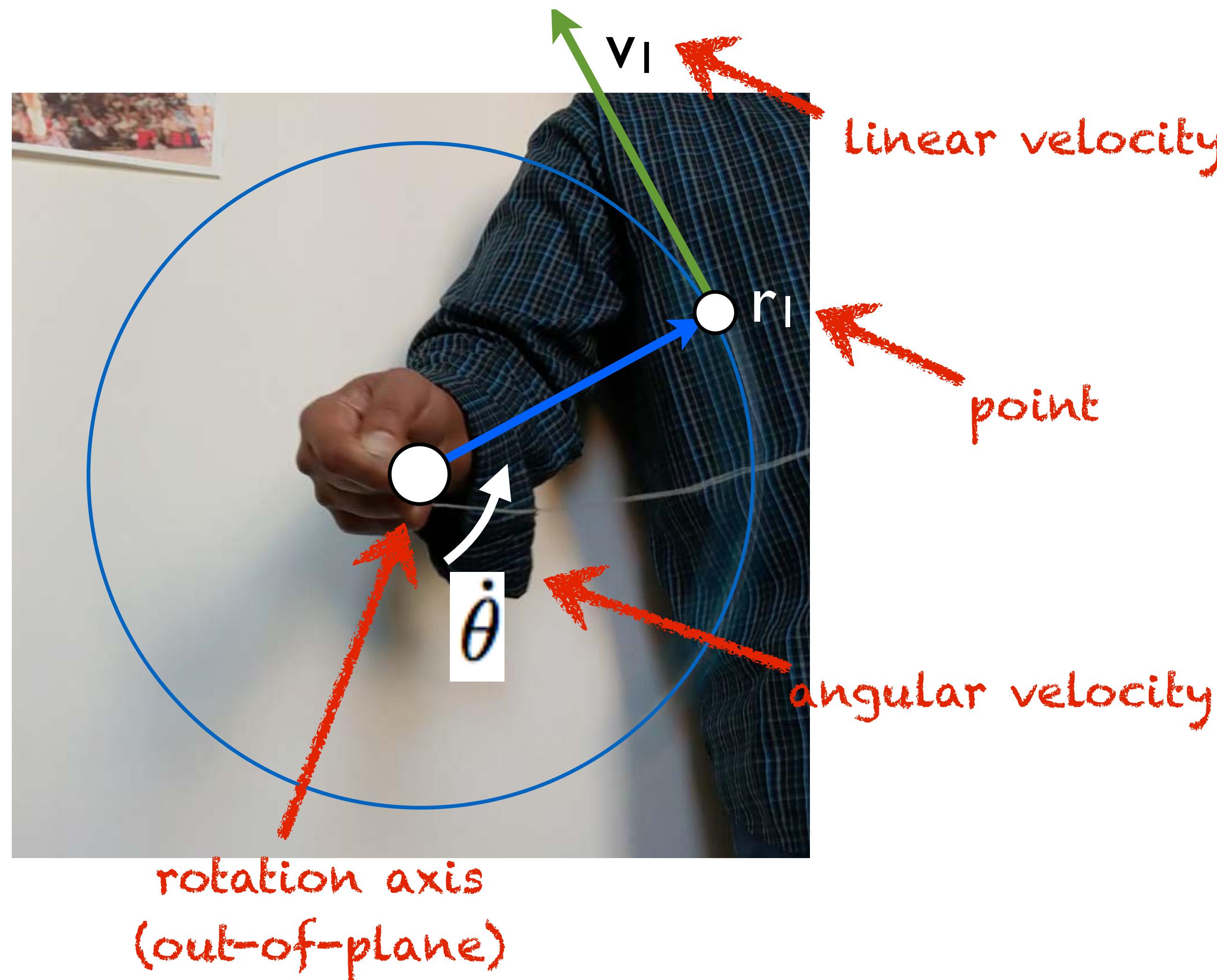


Consider the velocity of a point

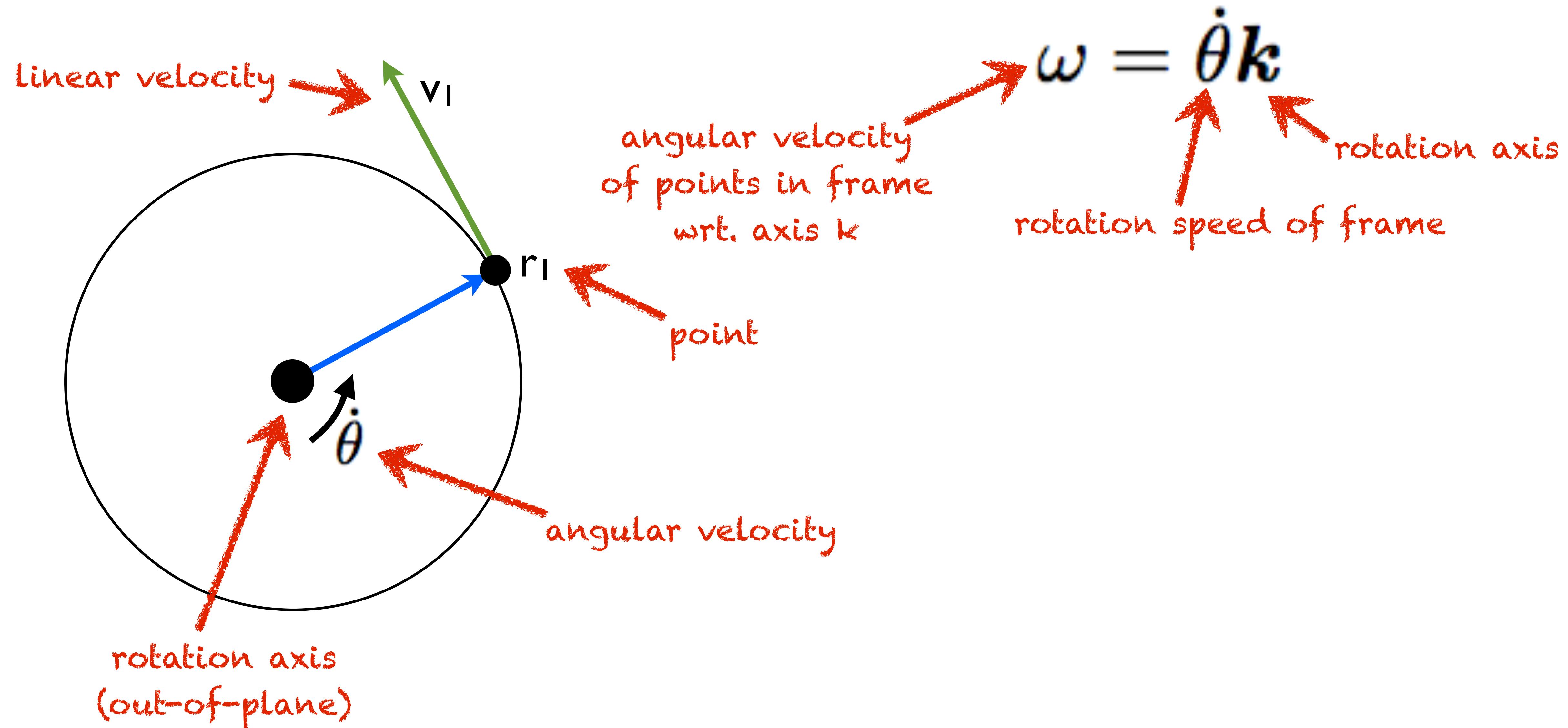


Consider the velocity of a point

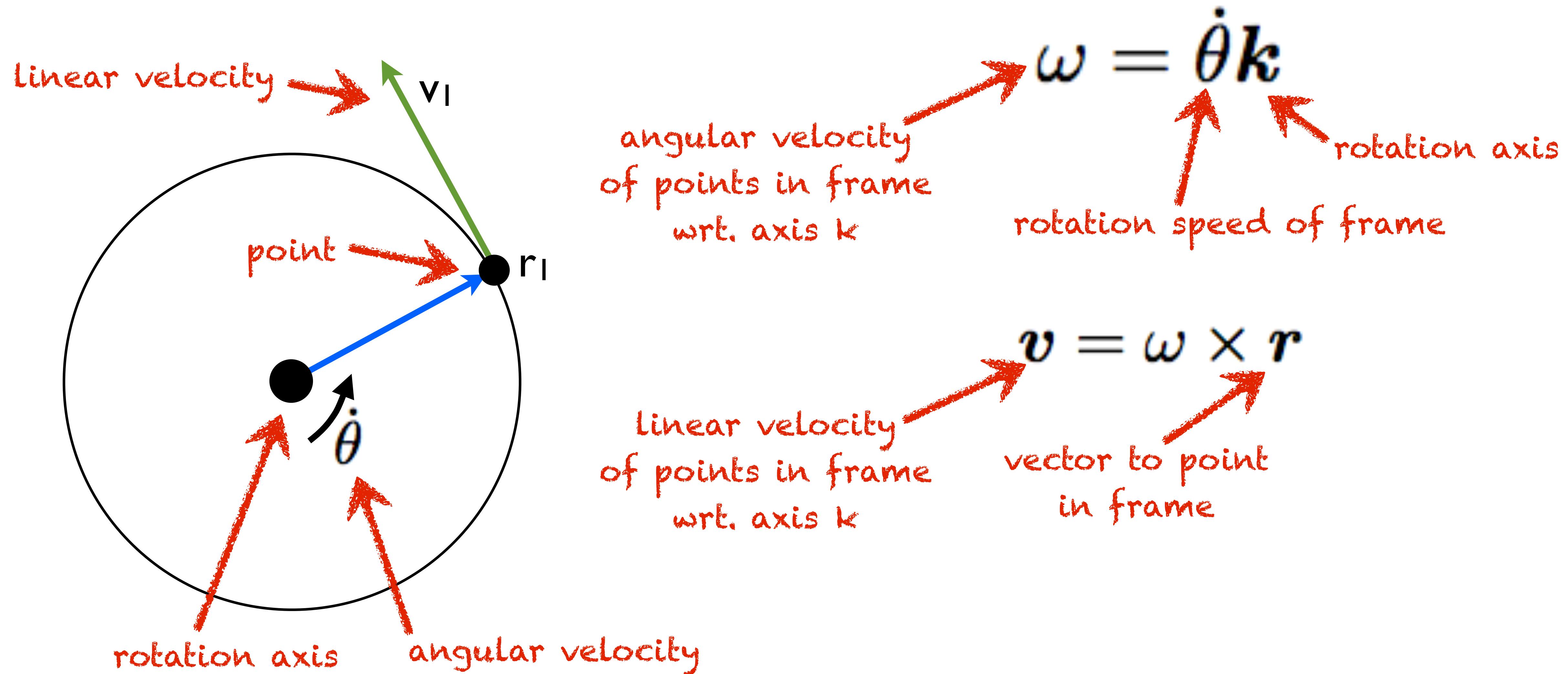
Velocity of Point Rotating in Fixed Frame



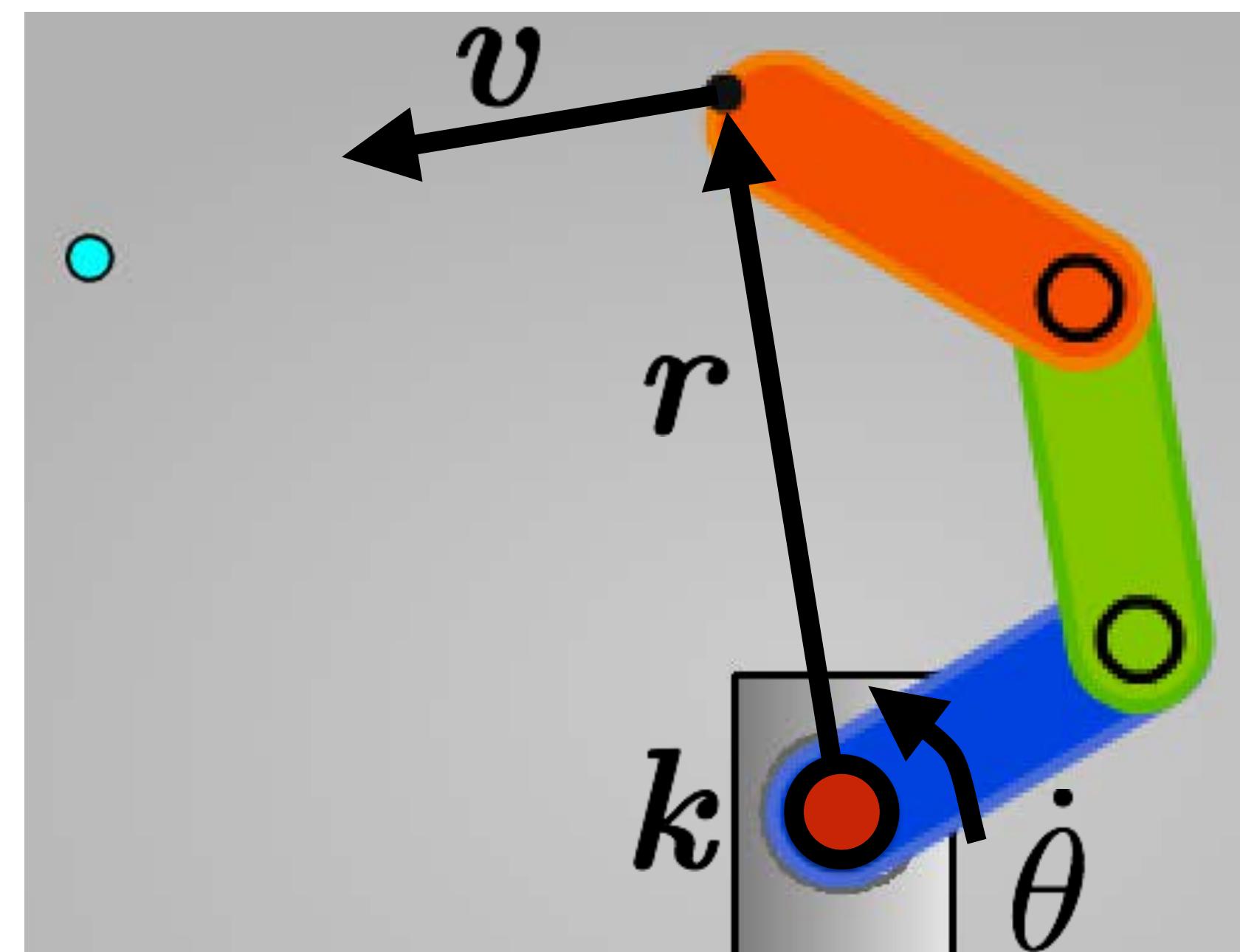
Velocity of Point Rotating in Fixed Frame



Velocity of Point Rotating in Fixed Frame



Velocity of Point Rotating in Fixed Frame



angular velocity
of points in frame
wrt. axis k

$$\omega = \dot{\theta} k$$

Linear velocity
of points in frame
wrt. axis k

rotation speed of frame

endeffector
linear velocity

$$\vec{v} = \omega \times \vec{r}$$

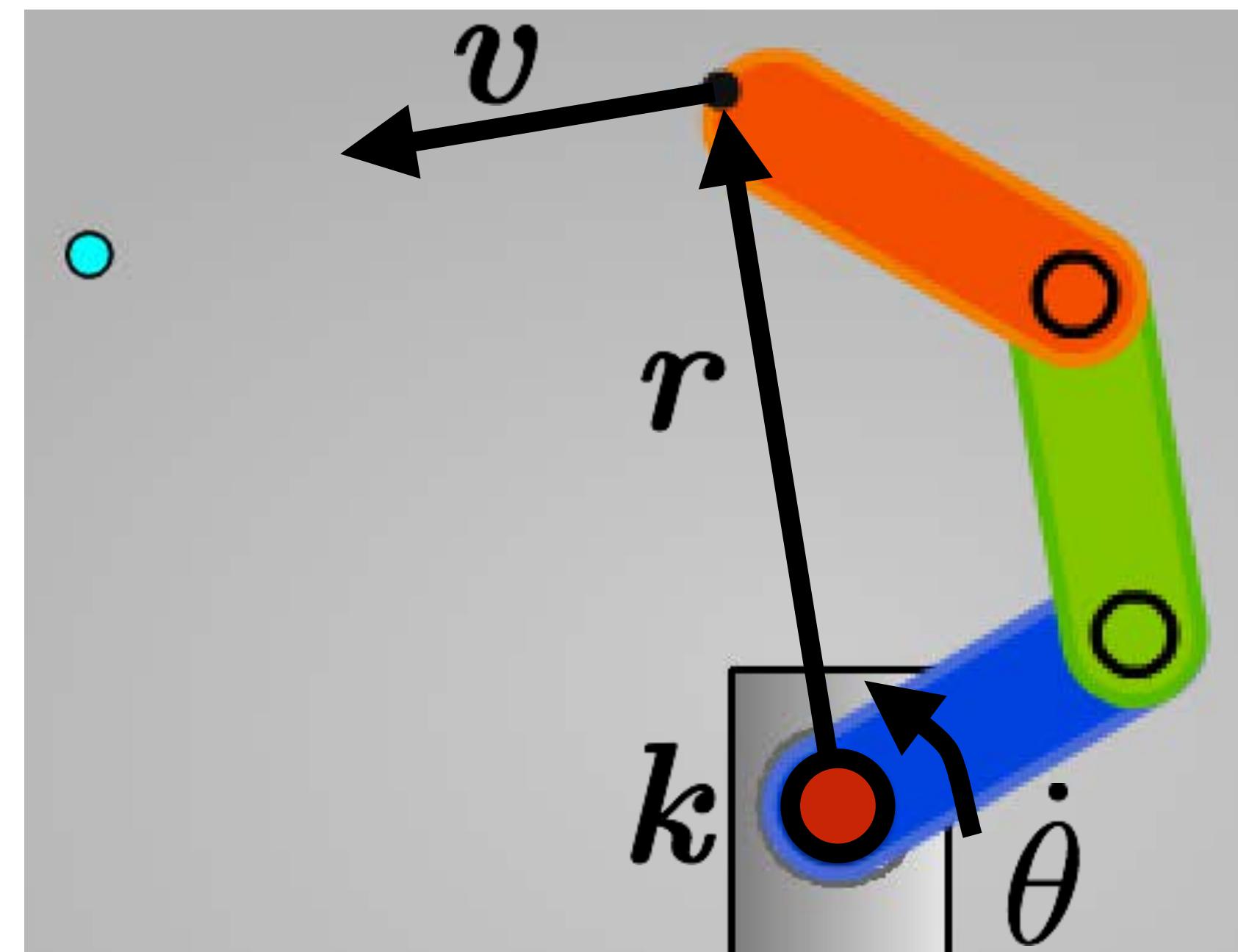
velocity
in frame
S k

vector to point
in frame

$$\vec{v} = \dot{\theta} \vec{k} \times \vec{r}$$

* joint rotation axis

vector from joint origin to endeffector



endeffector
Linear velocity

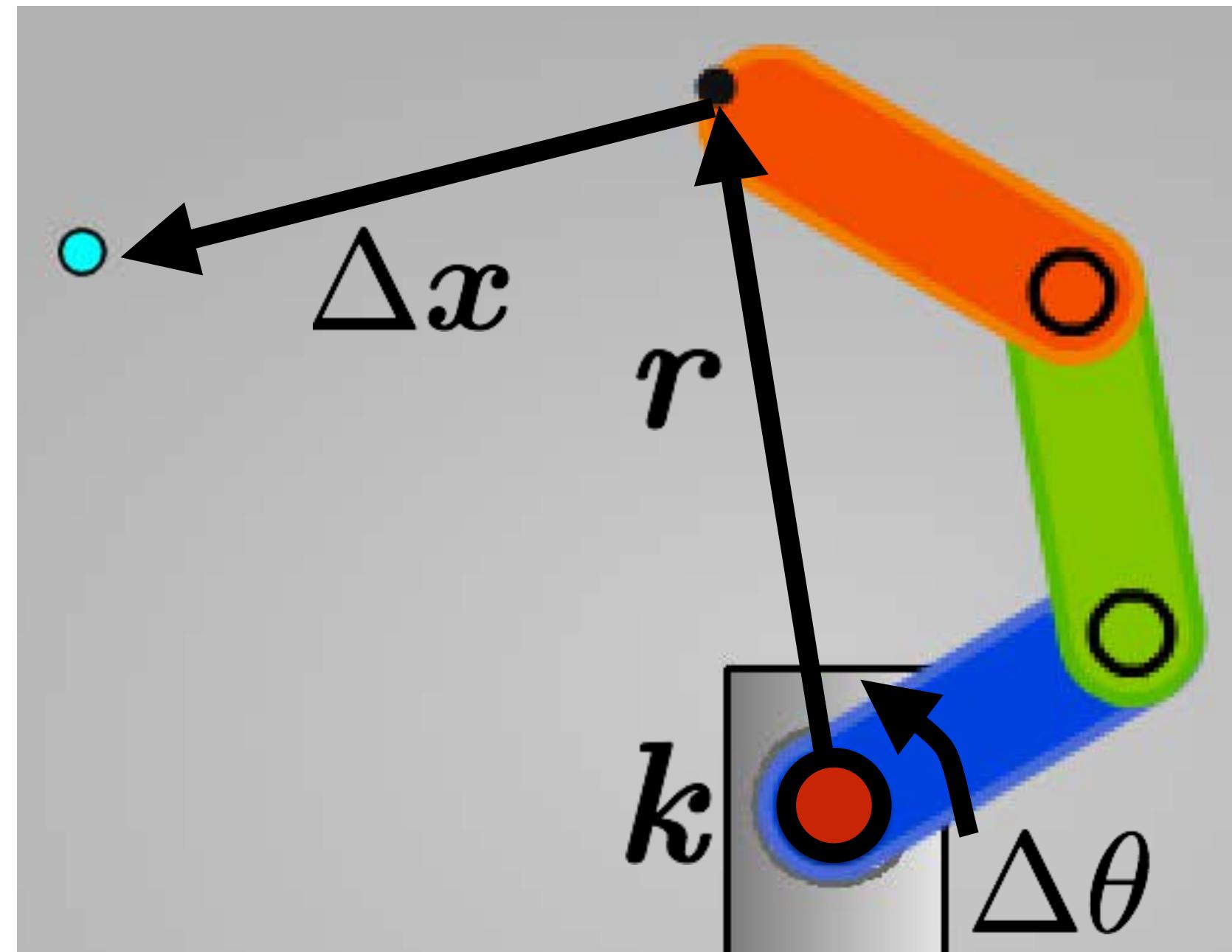
$$\vec{v} = \dot{\theta} k \times r$$

vector from
joint origin to
endeffector
joint rotation axis

This is not what we wanted.

Why?

Jacobian Transpose



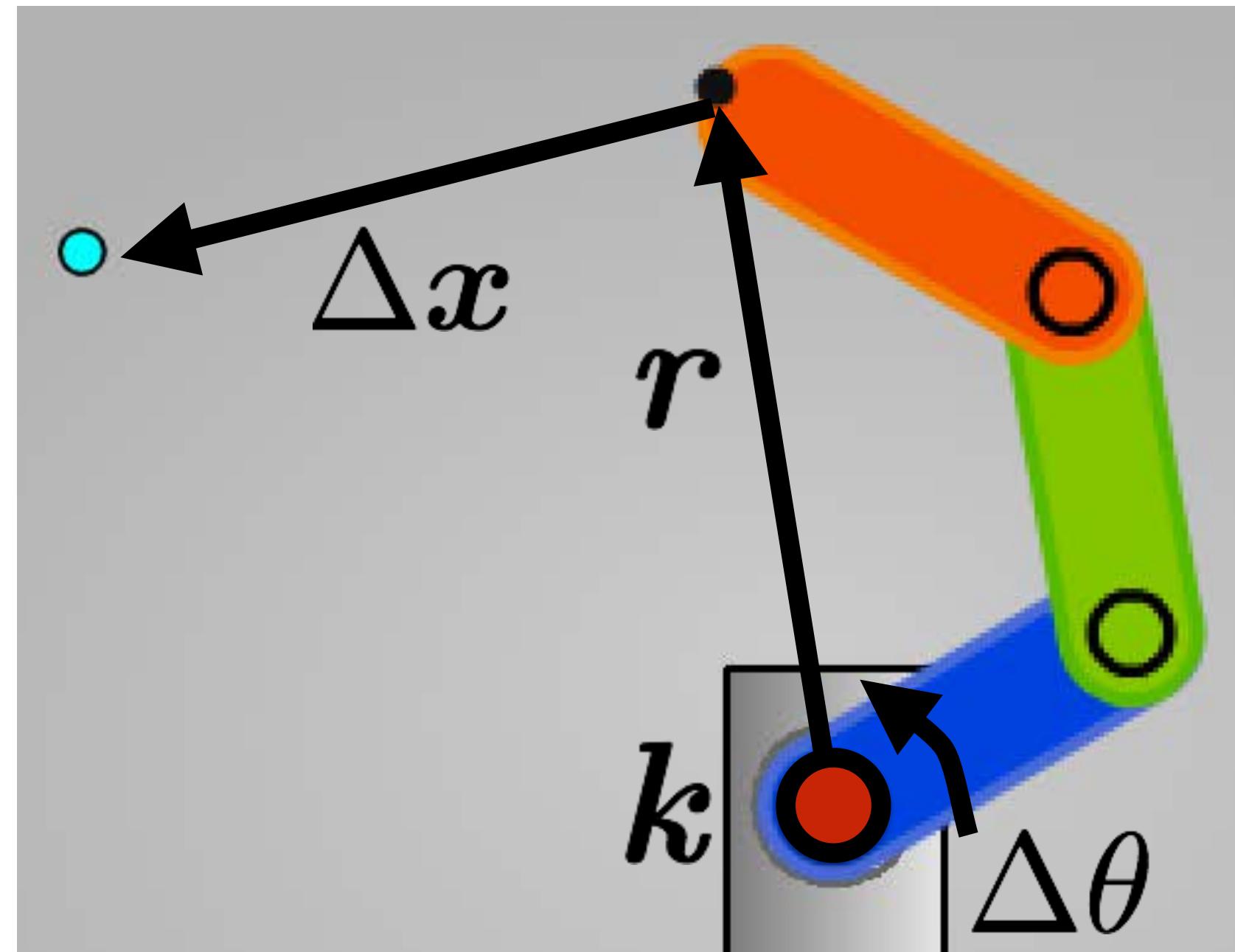
$$v = \dot{\theta}k \times r$$

vector from joint origin to endeffector
joint rotation axis
endeffector Linear velocity

This is not what we wanted.

How to obtain joint angular velocity from endeffector Linear velocity?

Jacobian Transpose



$$\vec{v} = \dot{\theta} k \times r$$

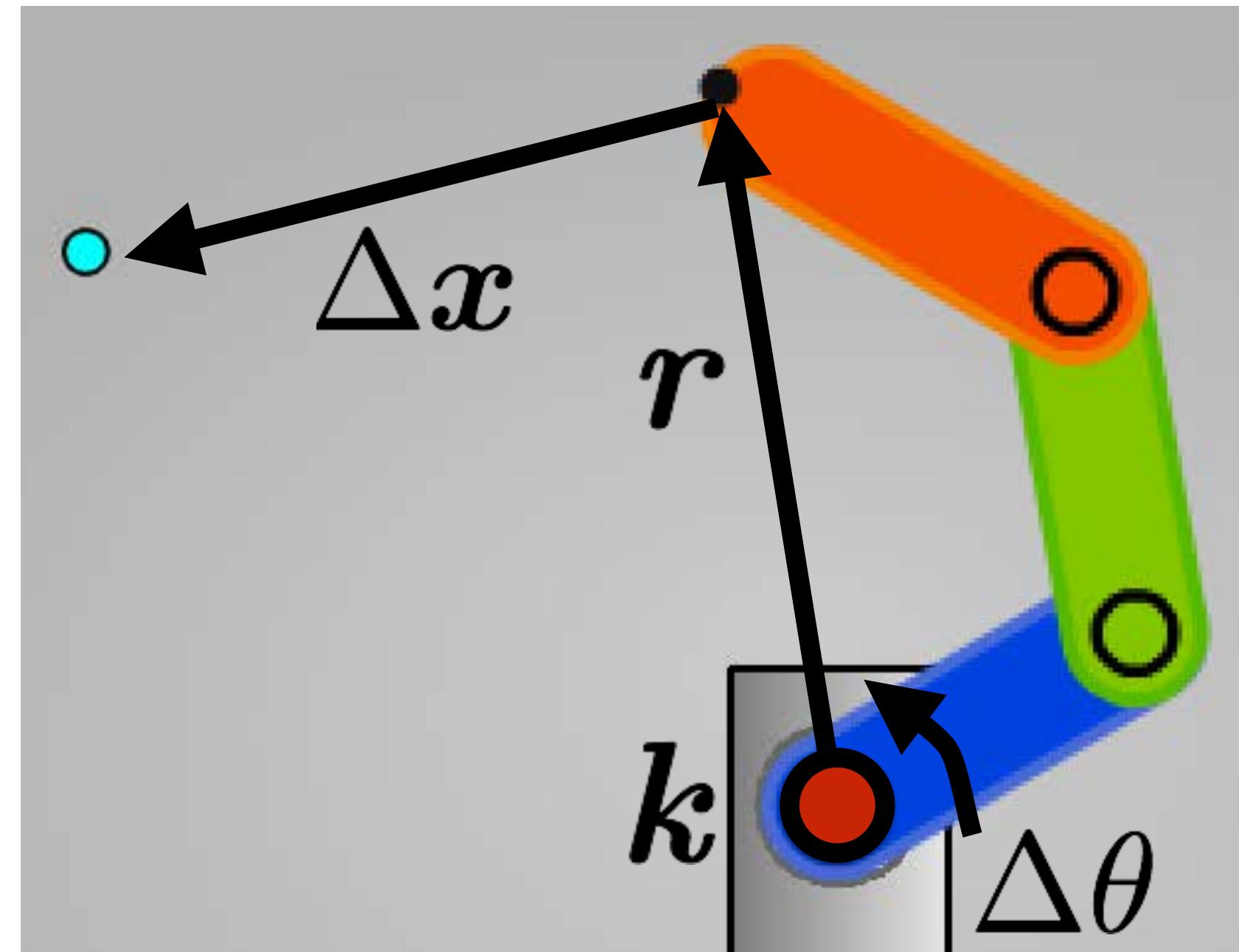
vector from joint origin to endeffector
joint rotation axis

This is not what we wanted.

How to obtain joint angular velocity from endeffector linear velocity?

$$\Delta\theta = (k \times r)^T \Delta x$$

Jacobian Transpose



$$\Delta\theta = \frac{(\underline{k} \times \underline{r})^T \Delta x}{\text{desired endeffector displacement}}$$

joint rotation axis

vector from joint origin to endeffector

Angular displacement for joint i

Jacobian for joint i

Procedure (for each joint):

- 1) Compute Jacobian
- 2) Update joint angles using Jacobian transpose
- 3) Repeat forever (or until error minimized)

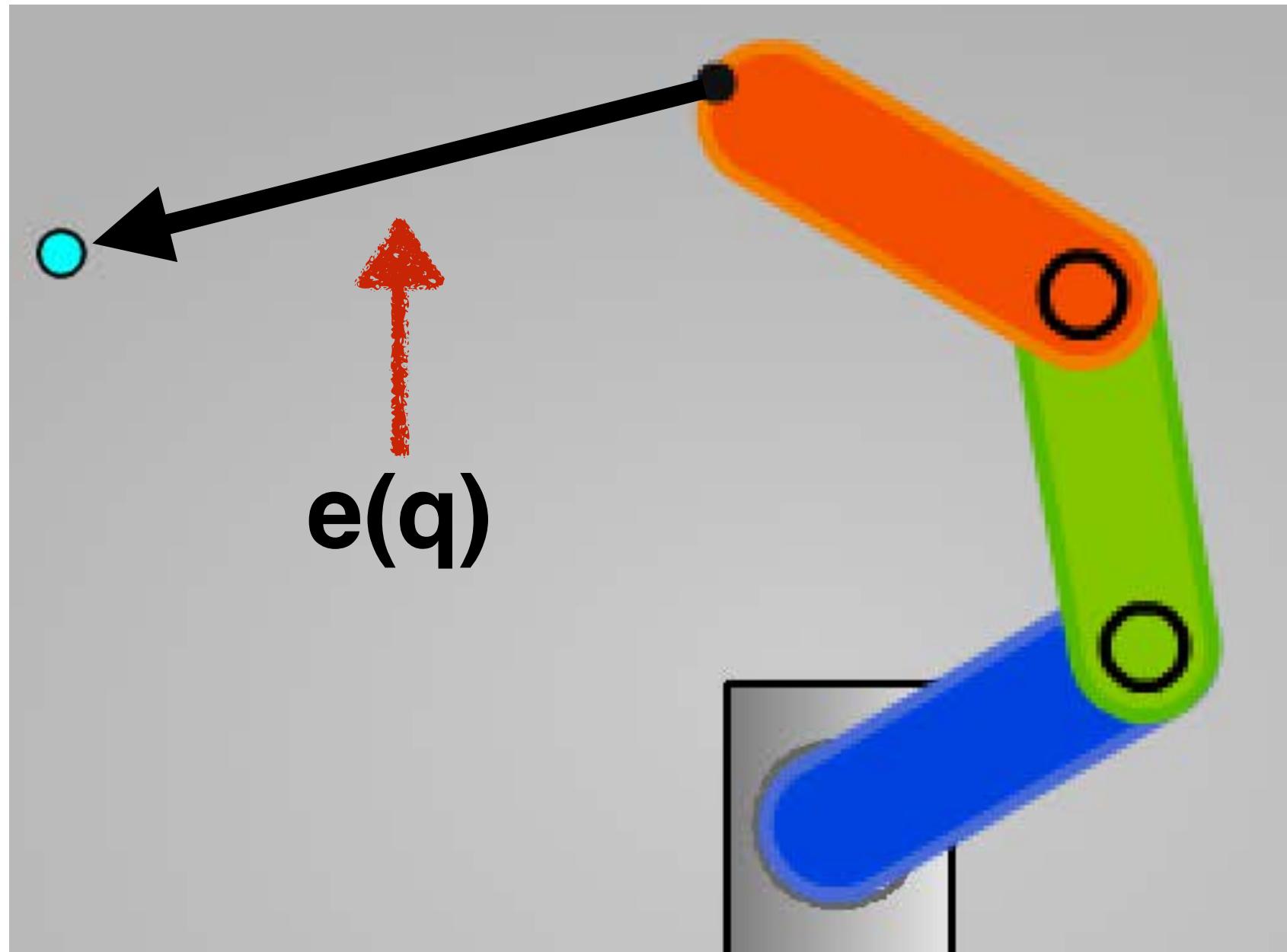
IK as Error Minimization

IK as Error Minimization

Gradient Descent Optimization



Inverse kinematics as error minimization

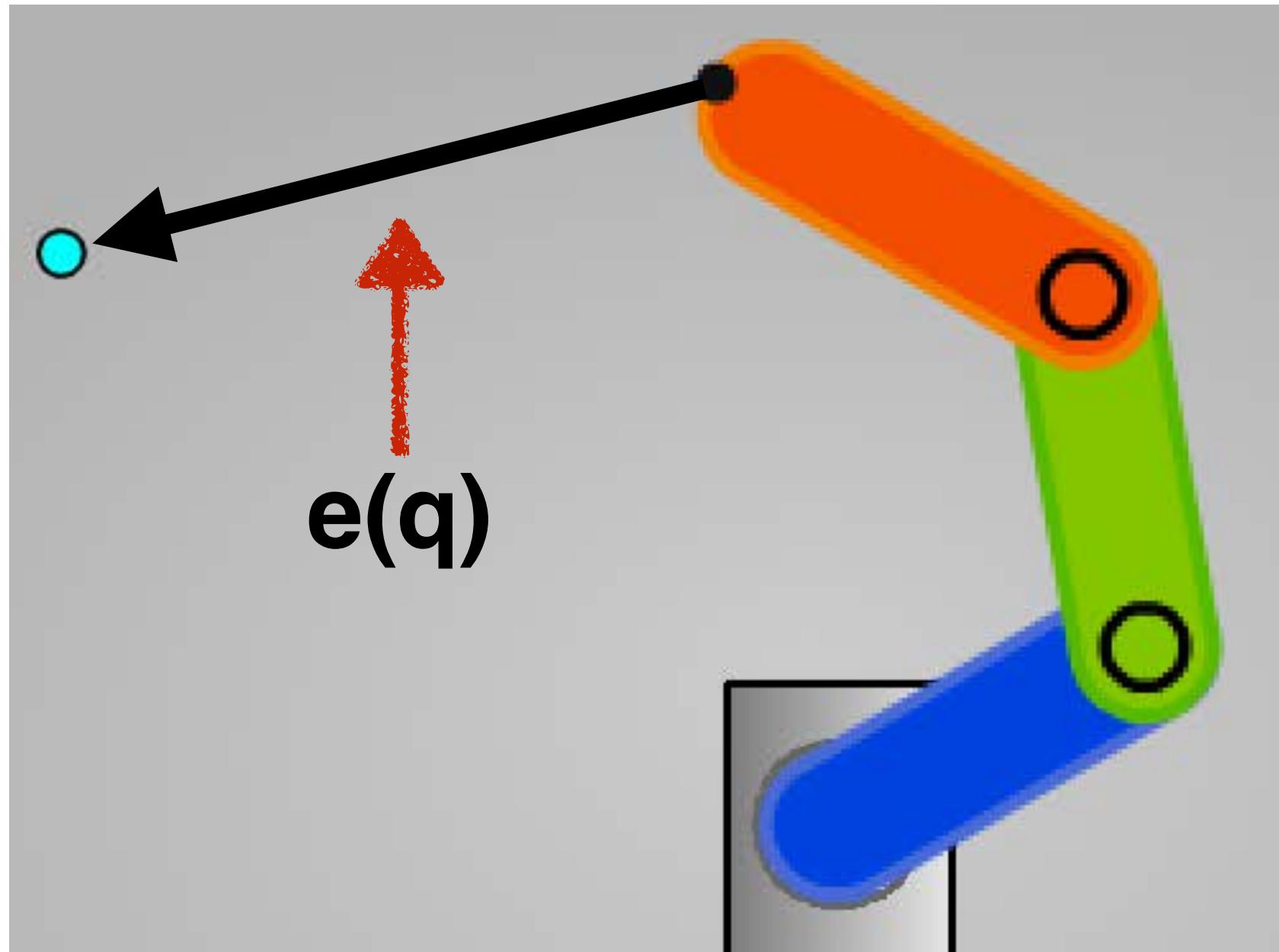


Define error function $e(\mathbf{q})$ as difference between current and desired endeffector poses

Error function parameterized by robot configuration \mathbf{q}

Find global minimum of $e(\mathbf{q})$: $\text{argmin}_{\mathbf{q}} e(\mathbf{q})$

Inverse kinematics as error minimization



Define error function $e(\mathbf{q})$ as difference between current and desired endeffector poses

Error function parameterized by robot configuration \mathbf{q}

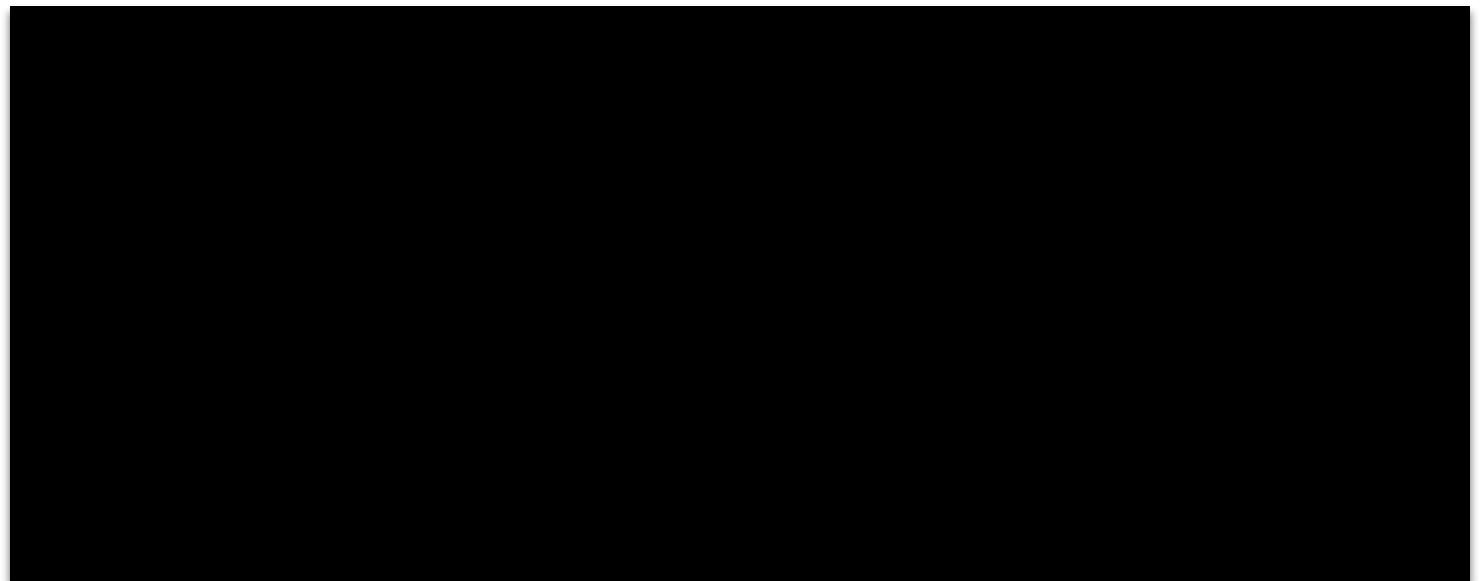
Find global minimum of $e(\mathbf{q})$: $\text{argmin}_{\mathbf{q}} e(\mathbf{q})$

How could we find $\text{argmin}_{\mathbf{q}} e(\mathbf{q})$ if we knew $e(\mathbf{q})$ in closed form?

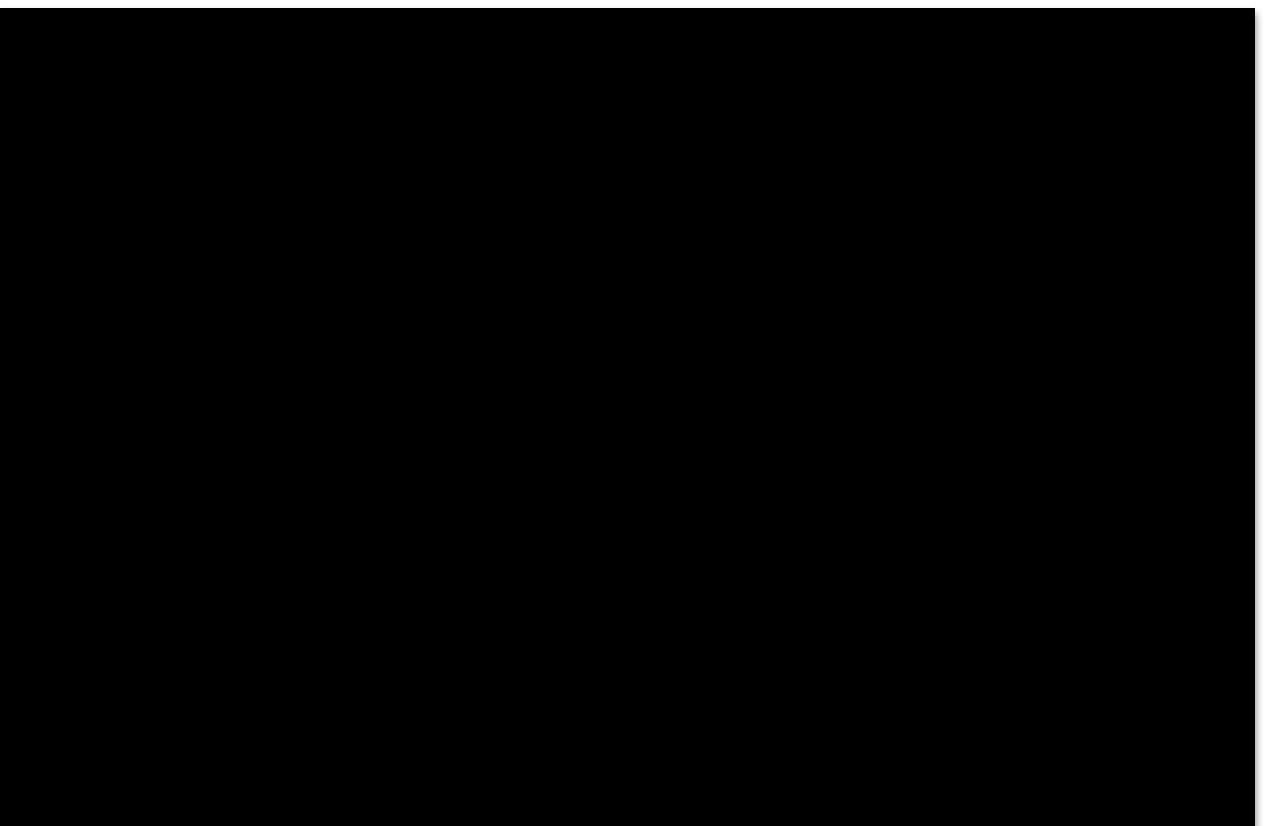
Example: Find global minimum of function

$$f(x) = 3(x - 2)^2$$

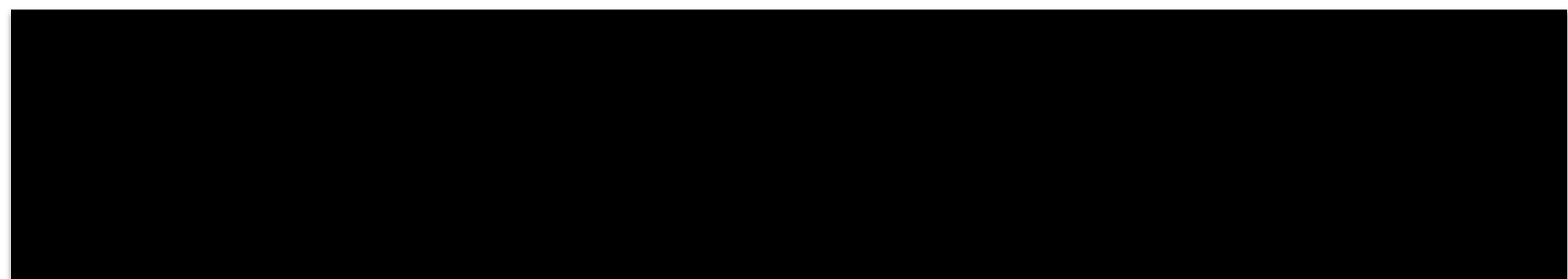
Take derivative



Solve for x where derivative is zero



Verify



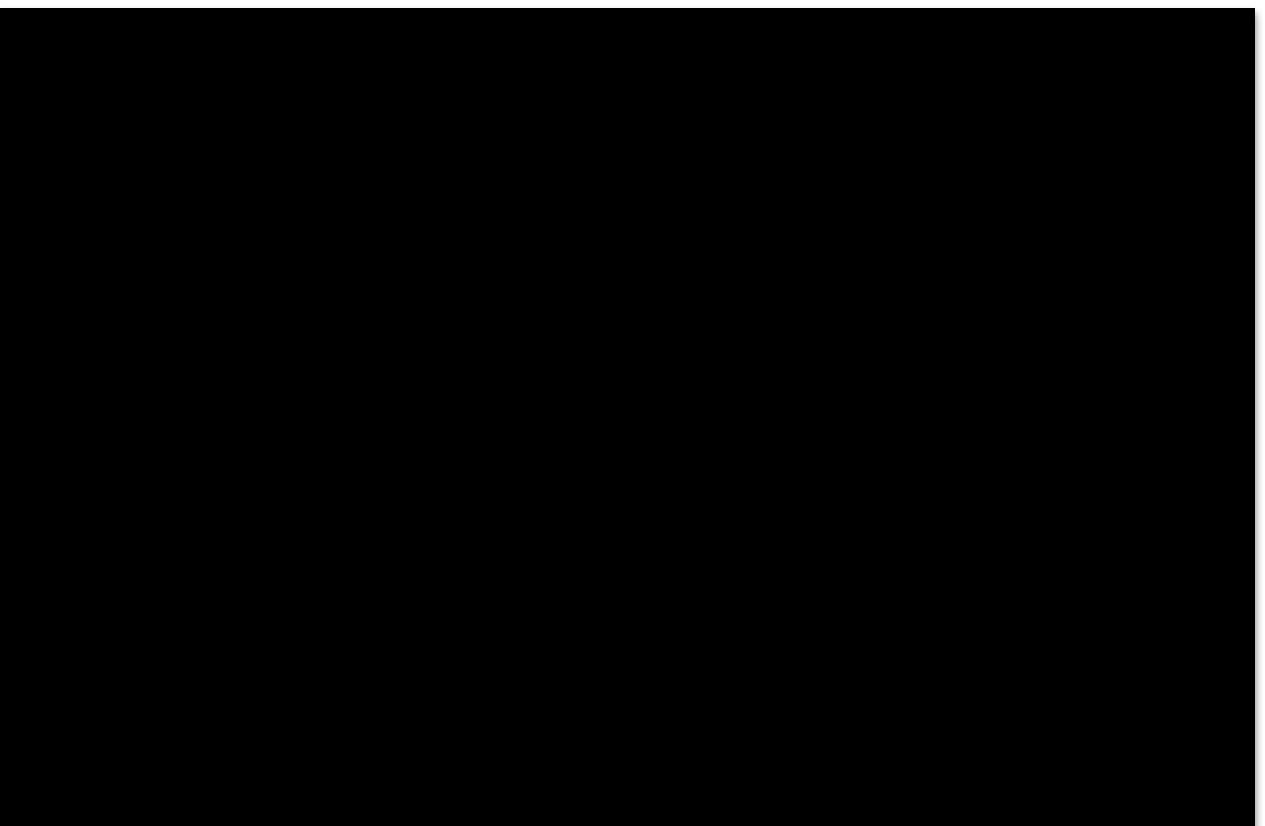
Example: Find global minimum of function

$$f(x) = 3(x - 2)^2$$

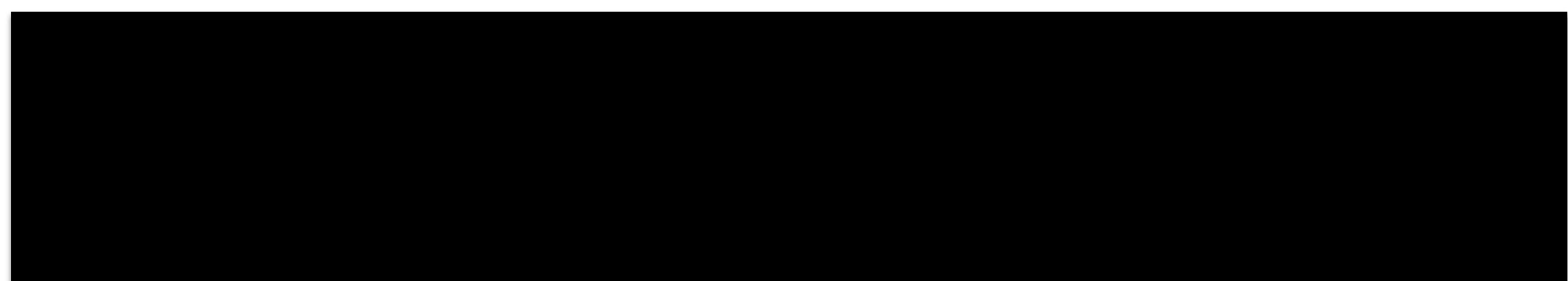
Take derivative

$$\frac{df}{dx} = 6(x - 2)$$

Solve for x where derivative is zero



Verify



Example: Find global minimum of function

$$f(x) = 3(x - 2)^2$$

Take derivative

$$\frac{df}{dx} = 6(x - 2)1$$

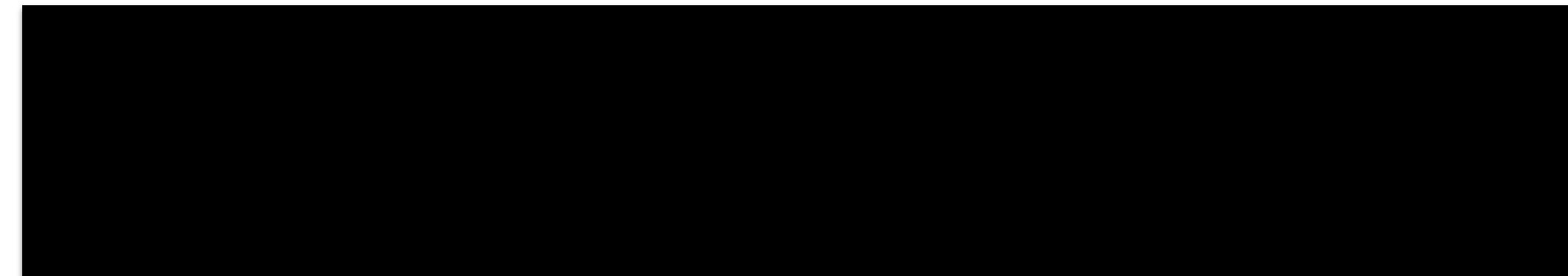
Solve for x where derivative is zero

$$6(x - 2) = 0$$

$$6x - 12 = 0$$

$$x = 2$$

Verify



Example: Find global minimum of function

$$f(x) = 3(x - 2)^2$$

Take derivative

$$\frac{df}{dx} = 6(x - 2)$$

Solve for x where derivative is zero

$$6(x - 2) = 0$$

$$6x - 12 = 0$$

$$x = 2$$

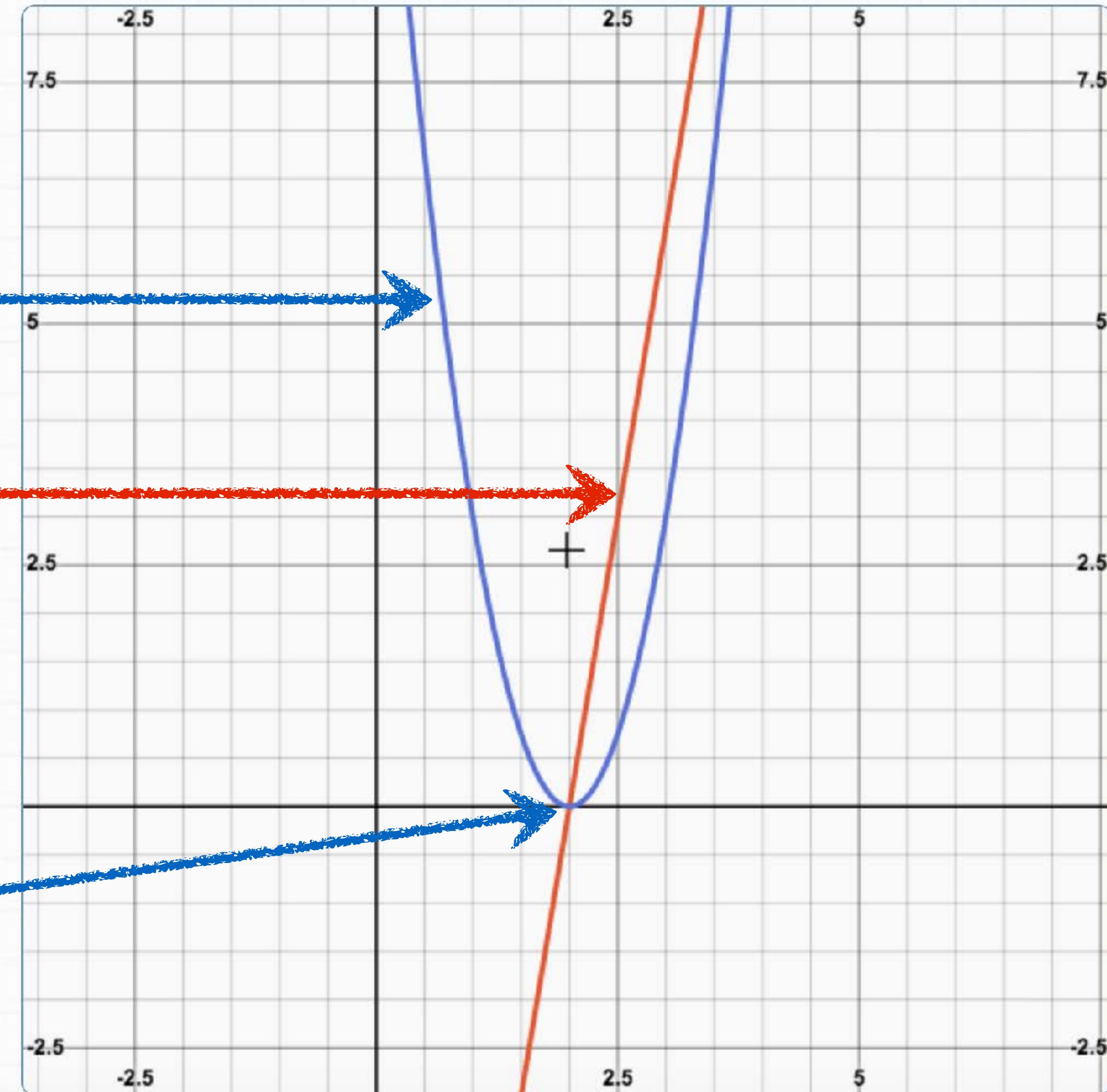
Verify $f(2) = 3((2) - 2)^2 = 0$



$$f(x) = 3(x - 2)^2$$

$$\frac{df}{dx} = 6(x - 2)$$

$$f(2) = 3((2) - 2)^2 = 0$$



Toggle graphs:

- $f(x)$
- $f'(x)$

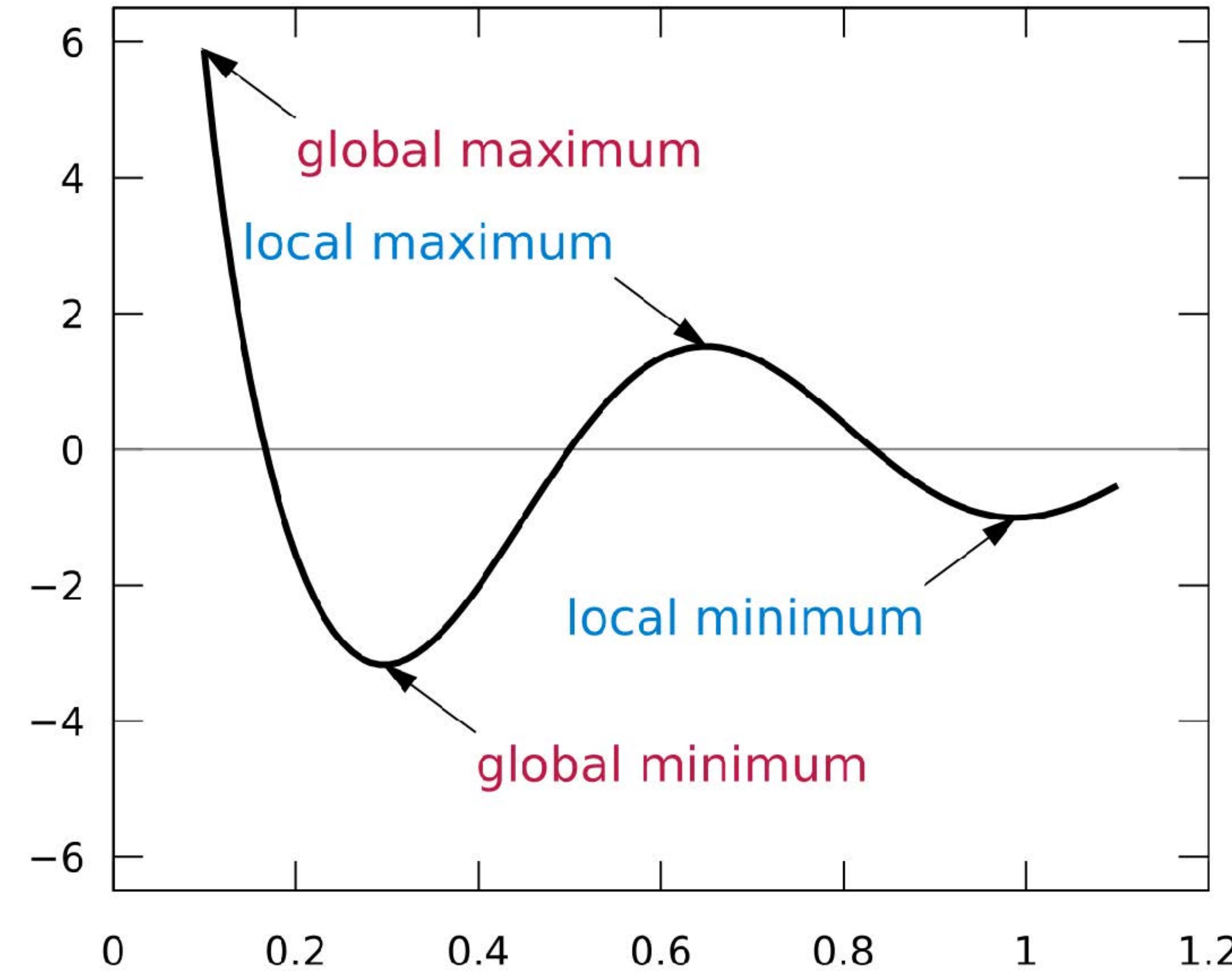
Table of values:

$x =$	<input type="text"/>
$f(x) =$	<input type="text"/>
$f'(x) =$	<input type="text"/>

Zoom mode:

- XY
- X
- Y
- 1:1**

Example: $\cos(3\pi x)/x$, $0.1 \leq x \leq 1.1$



 = Check your own answer
 = Export the expression (

commendation

Calculus for Dummies (2nd Edition)

An extremely well-written book for students taking Calculus for the first time as well as those who need a refresher. This book makes you realize that Calculus isn't that tough after all. → [to the book](#)

Did You Know? 

 Ad 


((SXM))
SEE WHAT YOU'VE BEEN HEARING. >

YOUR INPUT:
 $f(x) =$

$\frac{\cos(3\pi x)}{x}$

Simplify Roots/zeros

FIRST DERIVATIVE:

$\frac{d}{dx} [f(x)] = f'(x) =$

$$\frac{3\pi \sin(3\pi x)}{x} - \frac{\cos(3\pi x)}{x^2}$$

Simplify/rewrite:

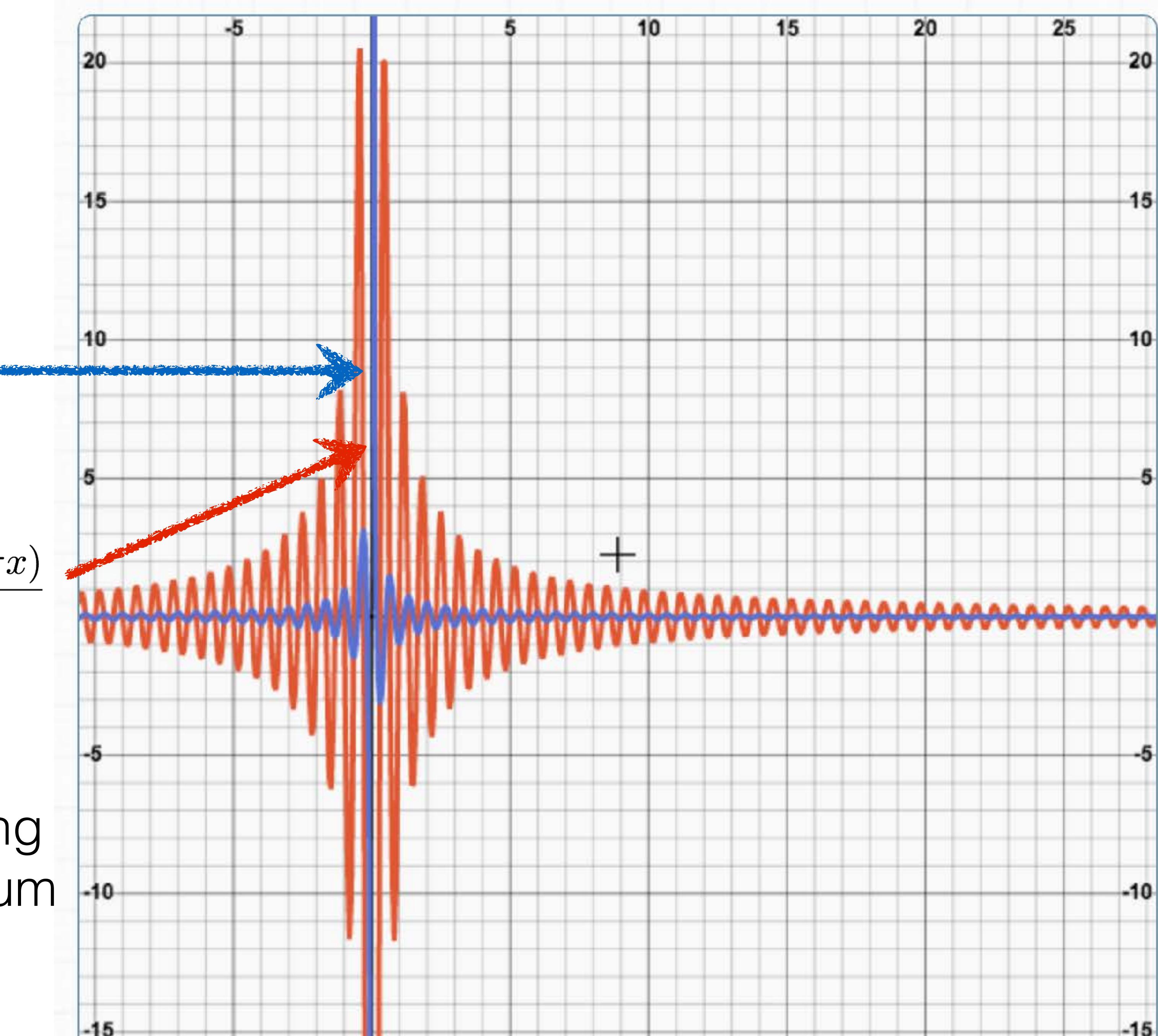
$$\frac{3\pi x \sin(3\pi x) + \cos(3\pi x)}{x^2}$$

Simplify Show steps Roots/zeros

$$\frac{\cos(3\pi x)}{x}$$

$$-\frac{3\pi x \sin(3\pi x) + \cos(3\pi x)}{x^2}$$

Every zero crossing
of $f'(x)$ is an optimum



Toggle graphs:

- $f(x)$
- $f'(x)$

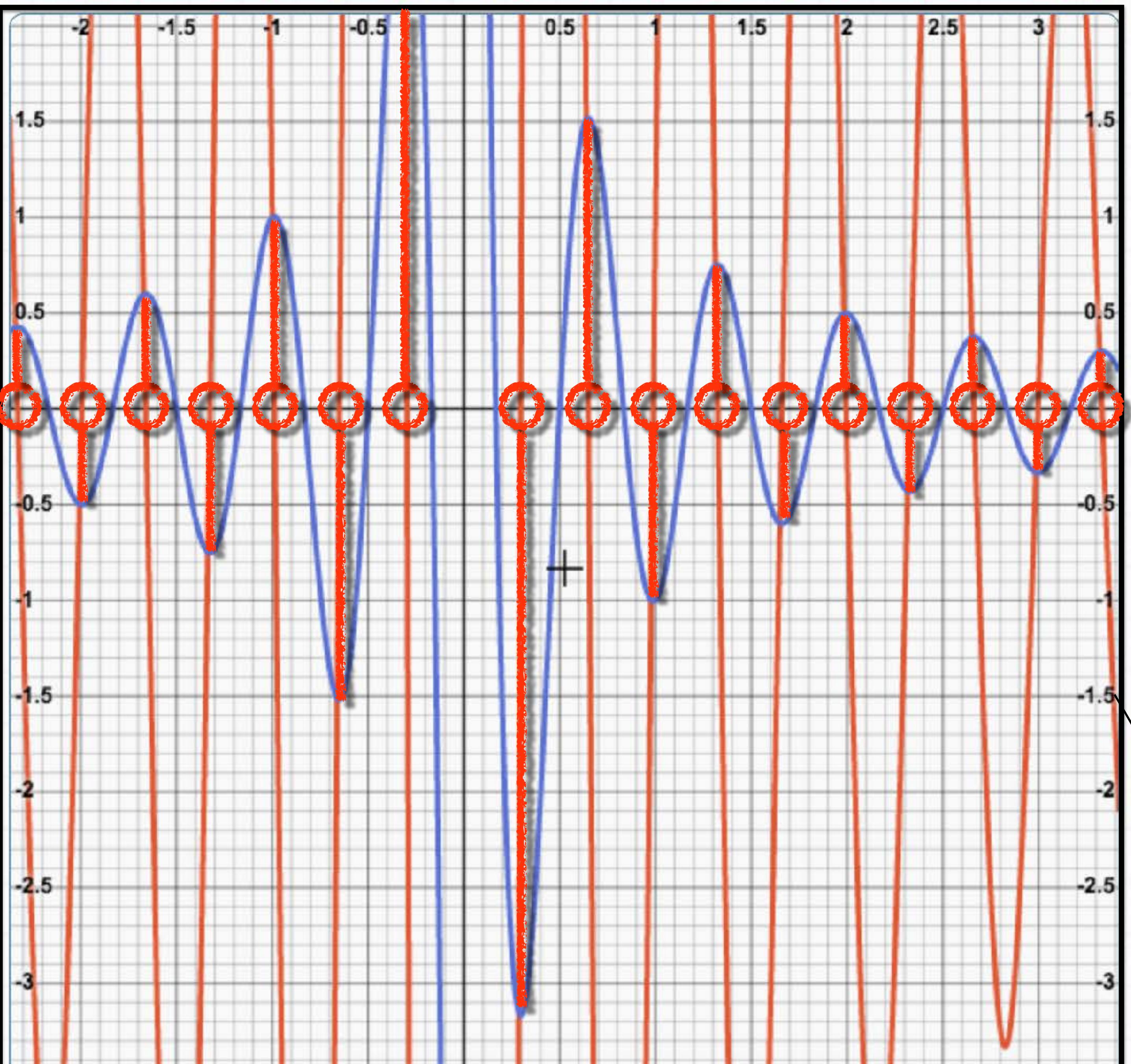
Table of values:

$x =$	<input type="text"/>
$f(x) =$	<input type="text"/>
$f'(x) =$	<input type="text"/>

Zoom mode:

- XY
- X
- Y
- 1:1**

Every zero crossing
of $f'(x)$ is an optimum



Toggle graphs:

- $f(x)$
- $f'(x)$

Table of values:

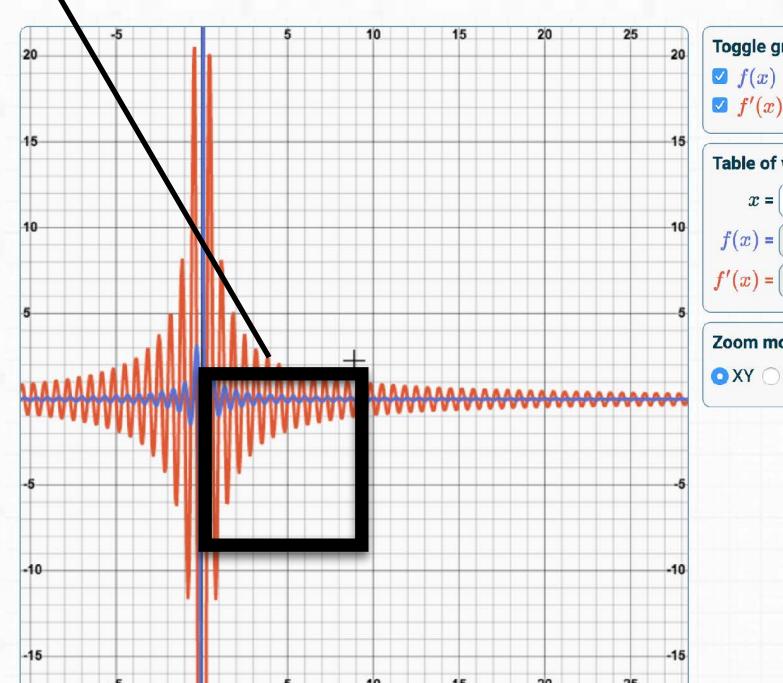
$x =$

$f(x) =$

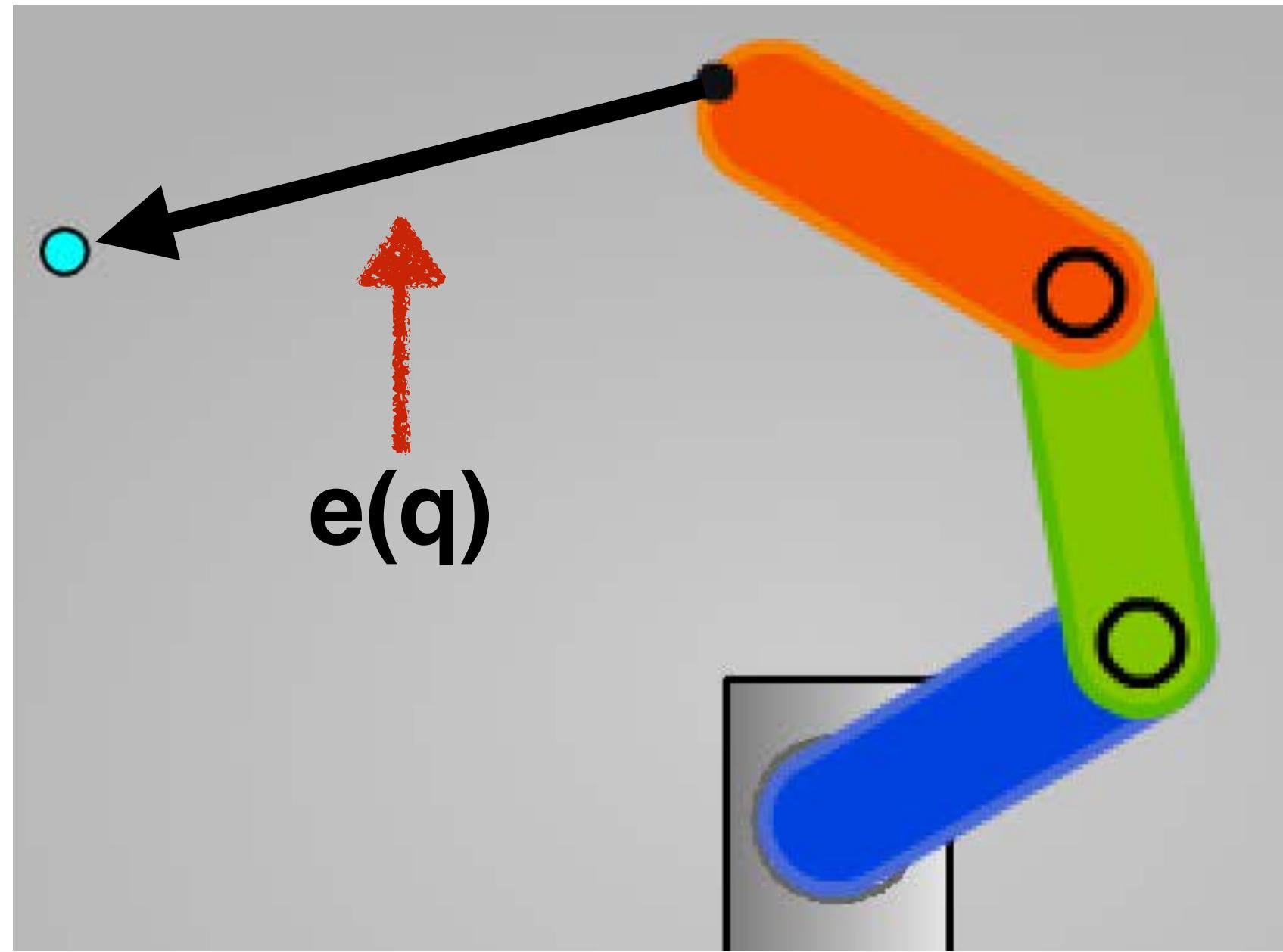
$f'(x) =$

Zoom mode:

- XY
 - X
 - Y
- 1:1



Inverse kinematics as error minimization



Define error function $e(\mathbf{q})$ as difference between current and desired endeffector poses

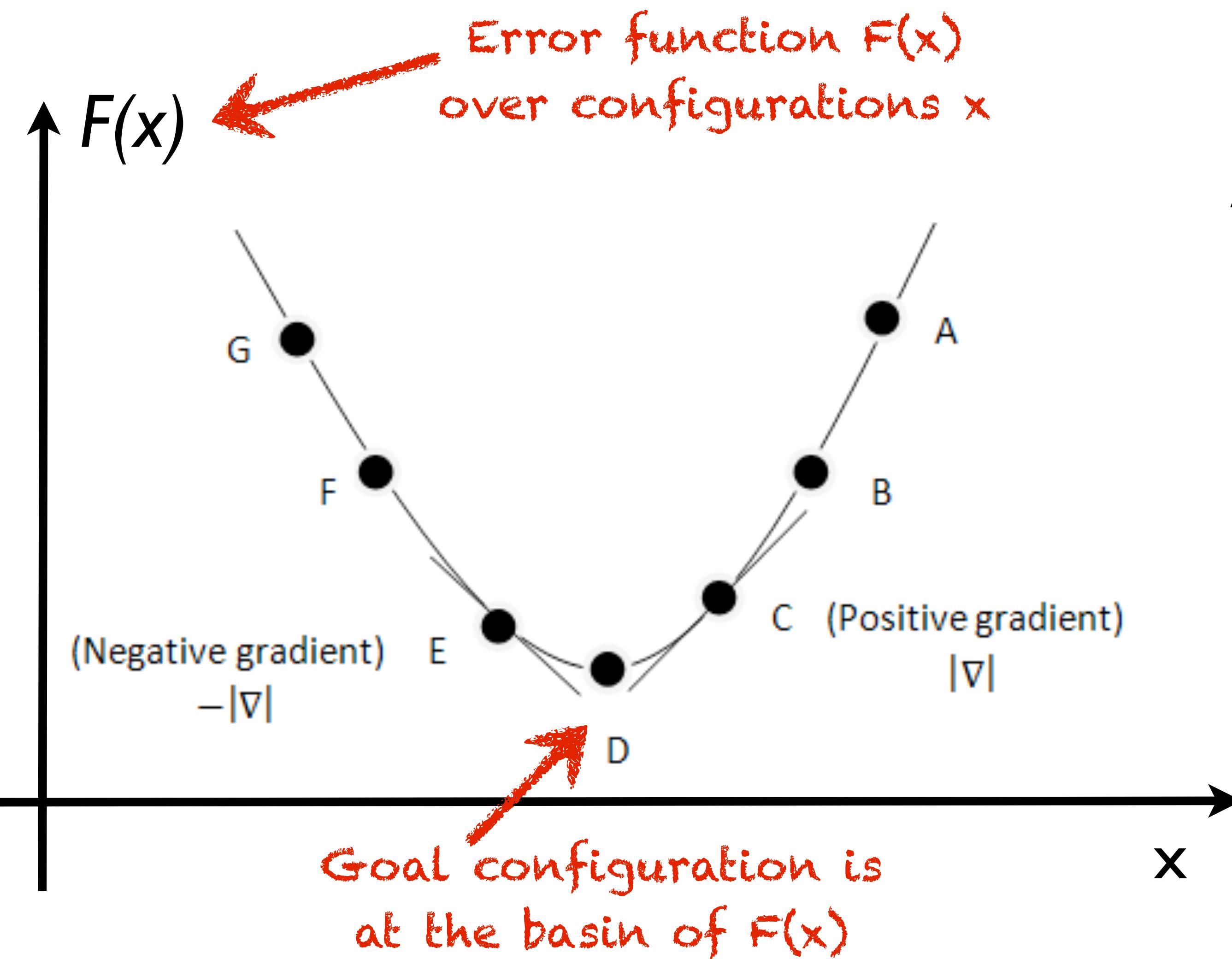
Error function parameterized by robot configuration \mathbf{q}

Find global minimum of $e(\mathbf{q})$,
or, $\text{argmin}_{\mathbf{q}} e(\mathbf{q})$

But, do we know $e(\mathbf{q})$ in closed form?

Gradient descent

From Wikipedia, the free encyclopedia



Assign initial solution guess \mathbf{x}_0

Repeat $\mathbf{x}_{i+1} = \mathbf{x}_i - \gamma_i \nabla F(\mathbf{x}_i)$

until $\|\mathbf{x}_i - \mathbf{x}_{i-1}\|$ is “small”

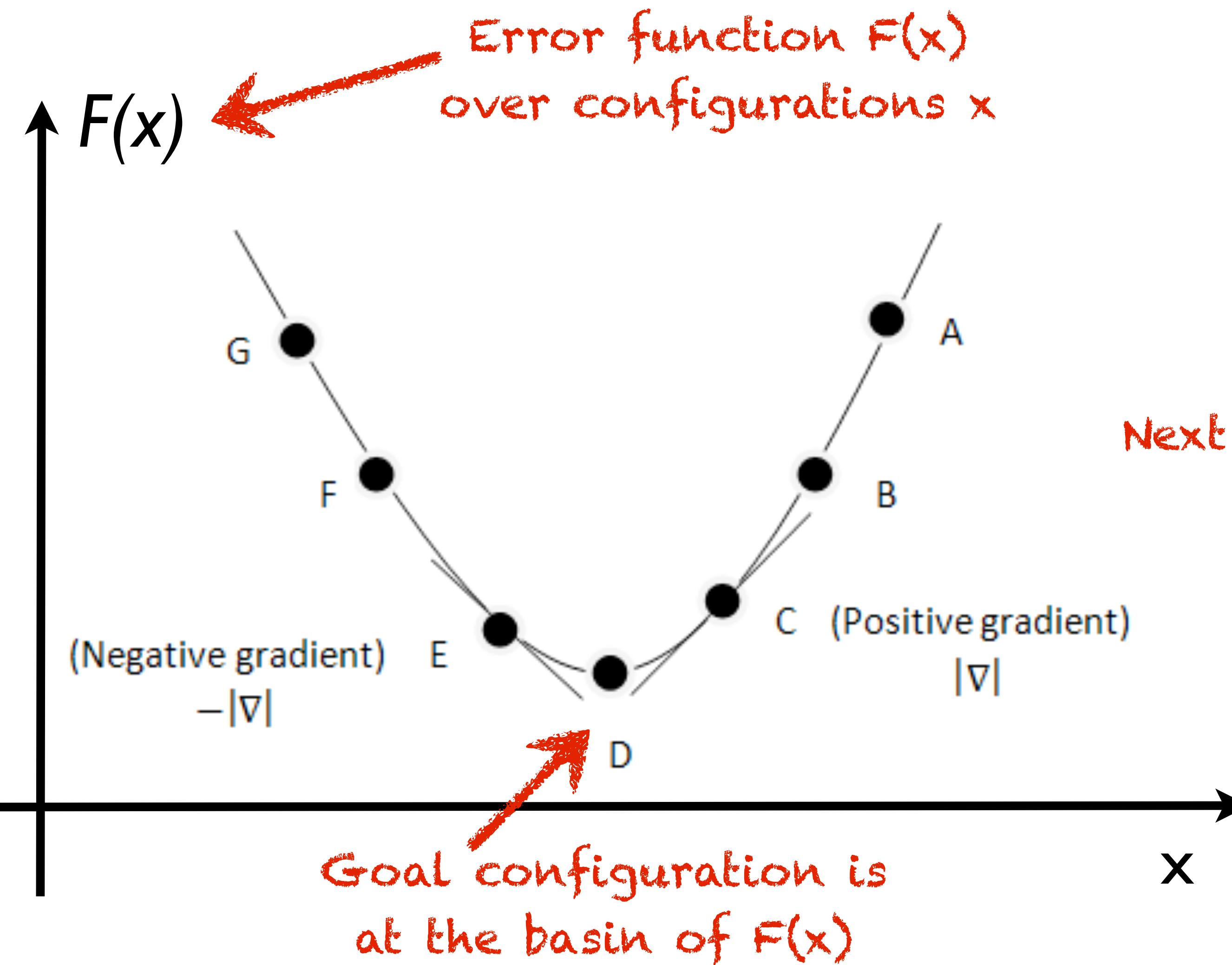
Gradient descent

From Wikipedia, the free encyclopedia



Gradient descent

From Wikipedia, the free encyclopedia



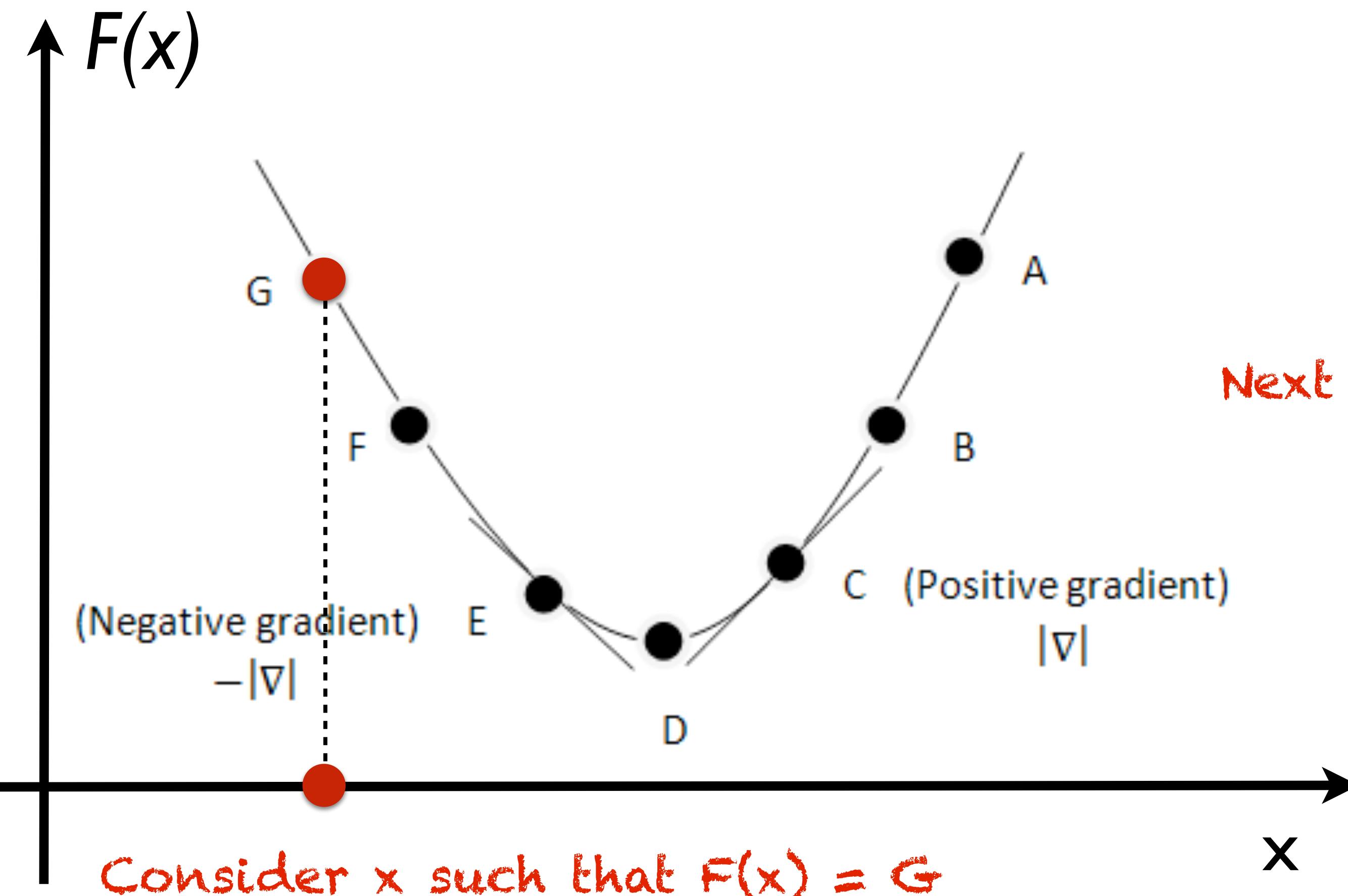
Current solution "Learning rate"

$$x_{i+1} = x_i - \gamma \nabla F(x_i)$$

Derivative assumed to be direction of steepest ascent away from goal

Gradient descent

From Wikipedia, the free encyclopedia



Current solution "Learning rate"

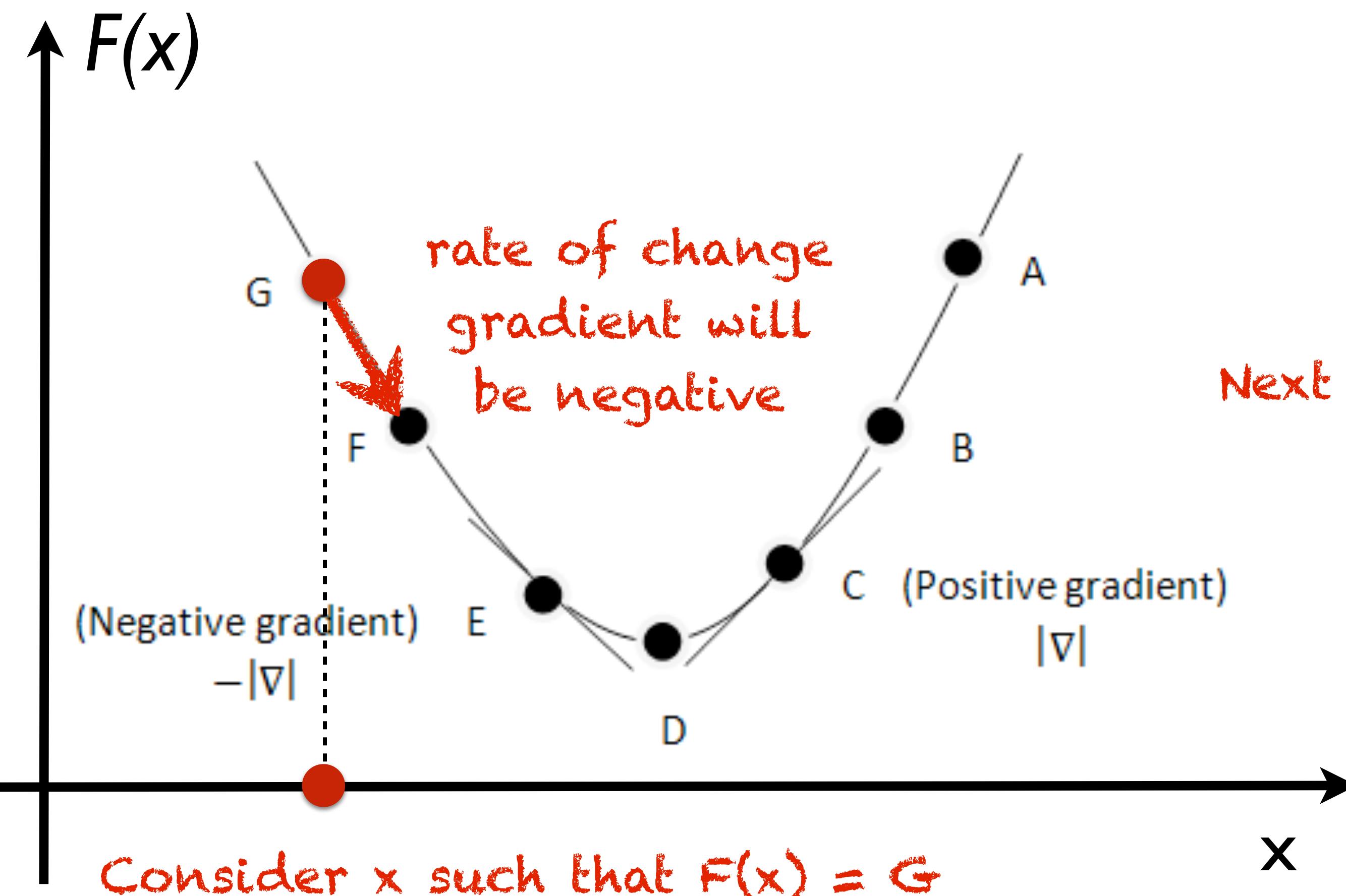
$$x_{i+1} = x_i - \gamma \nabla F(x_i)$$

Next solution

Derivative assumed to be direction of steepest ascent away from goal

Gradient descent

From Wikipedia, the free encyclopedia



Current solution "Learning rate"

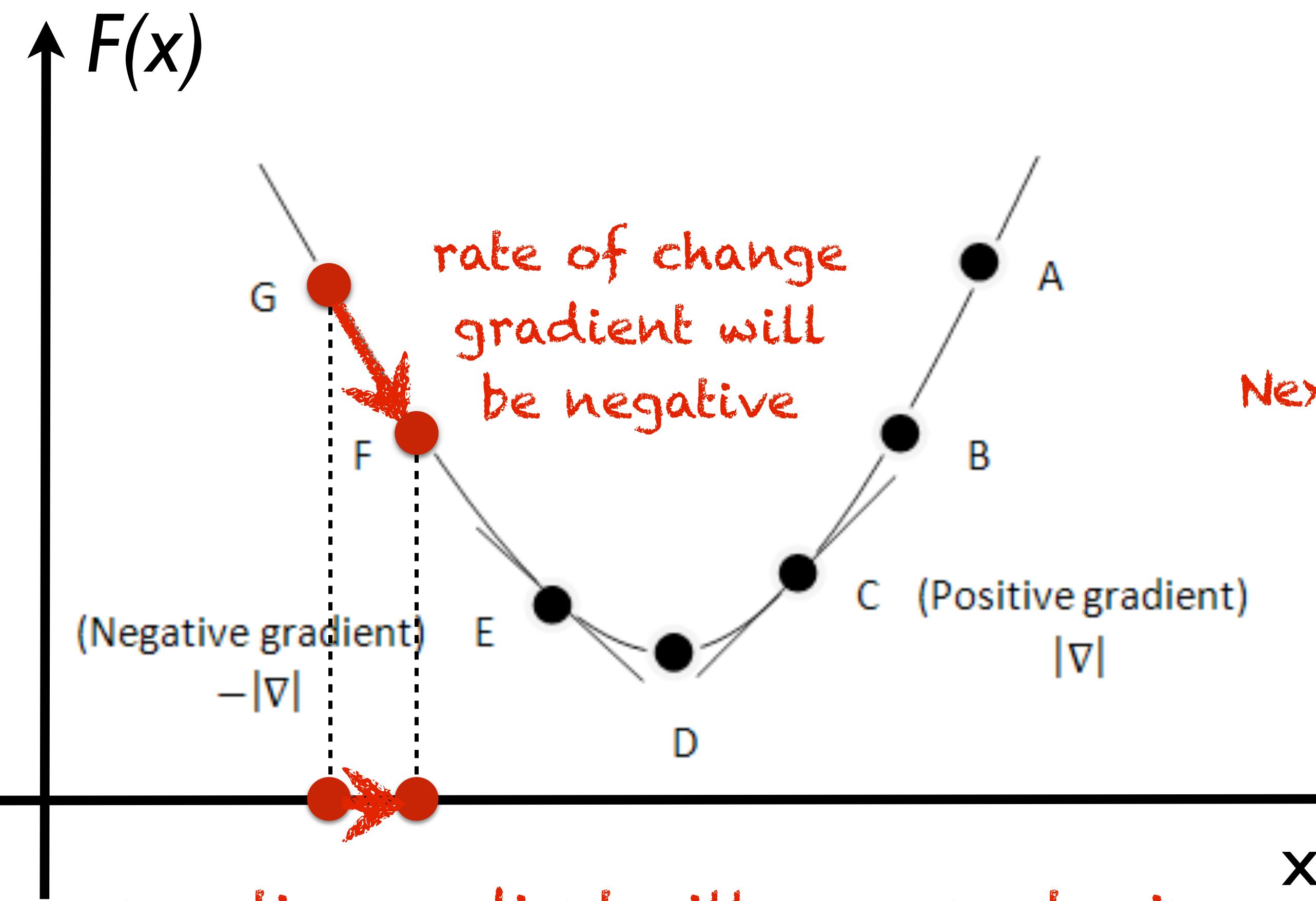
$$x_{i+1} = x_i - \gamma \nabla F(x_i)$$

Next solution

Derivative assumed to be direction of steepest ascent away from goal

Gradient descent

From Wikipedia, the free encyclopedia



Current solution "Learning rate"

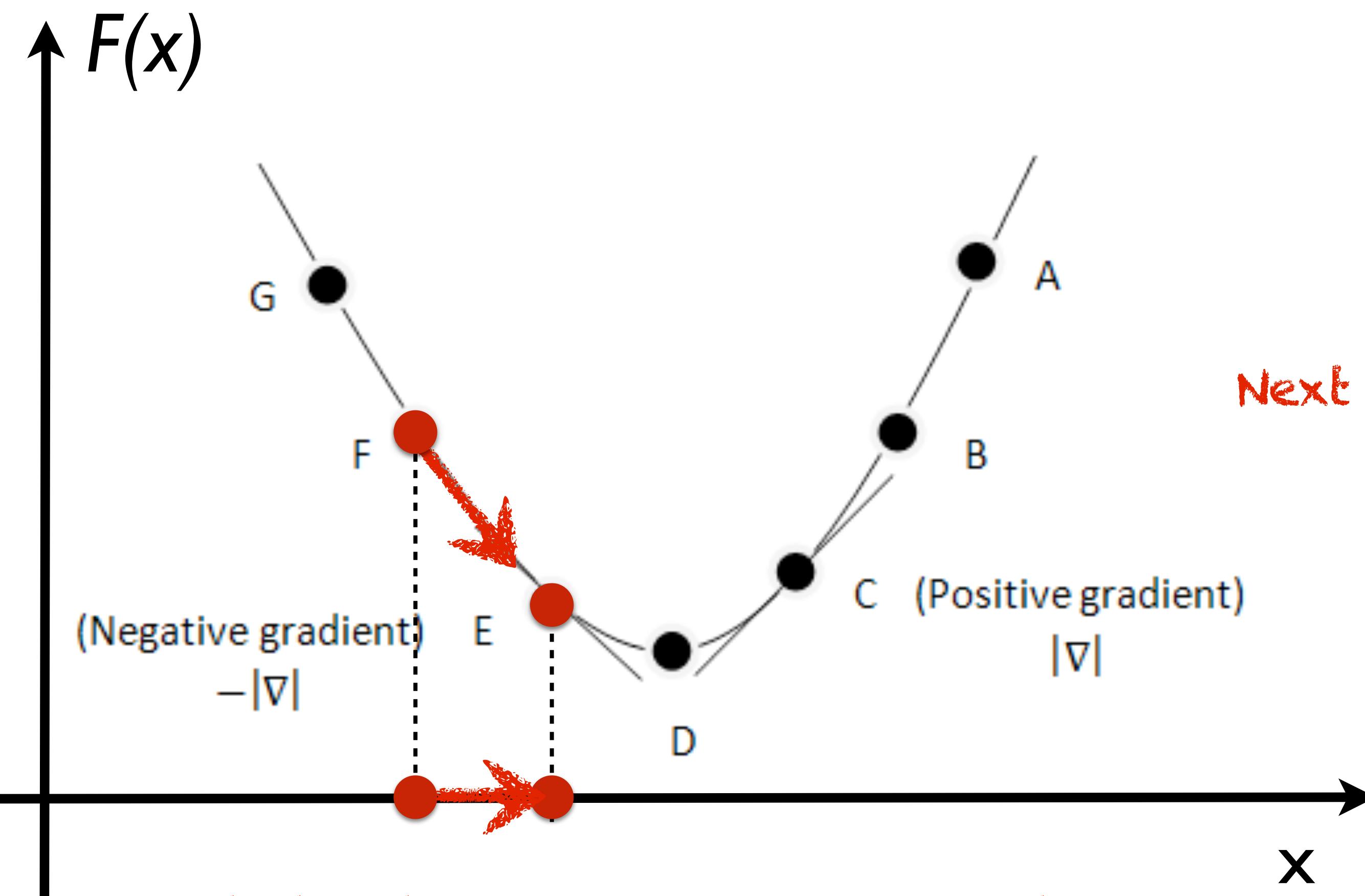
$$x_{i+1} = x_i - \gamma \nabla F(x_i)$$

Next solution

Derivative assumed to be direction
of steepest ascent away from goal

Gradient descent

From Wikipedia, the free encyclopedia



Current solution "Learning rate"

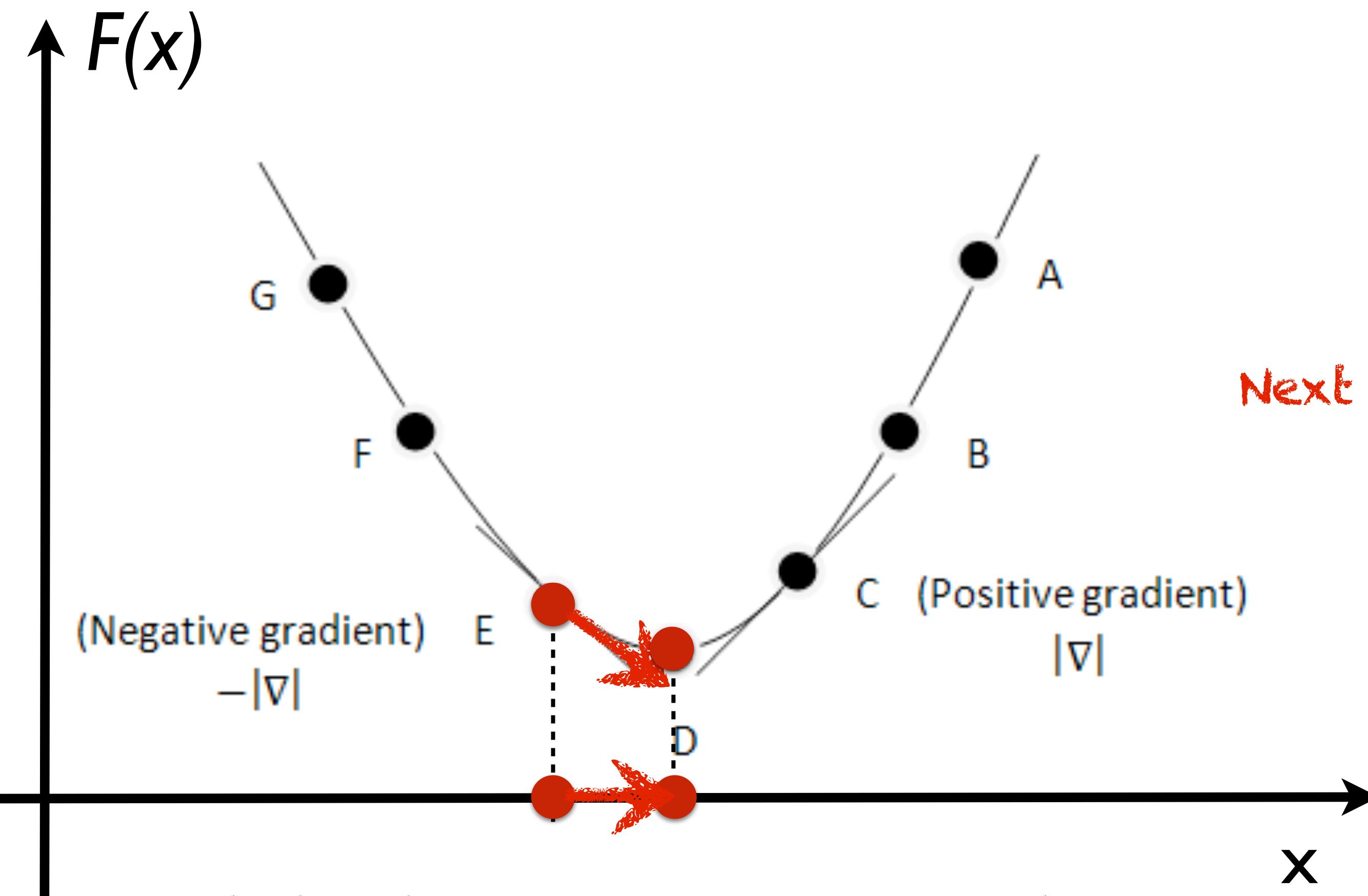
$$x_{i+1} = x_i - \gamma \nabla F(x_i)$$

Next solution

Derivative assumed to be direction of steepest ascent away from goal

Gradient descent

From Wikipedia, the free encyclopedia



next iteration will move closer to goal

Current solution "Learning rate"

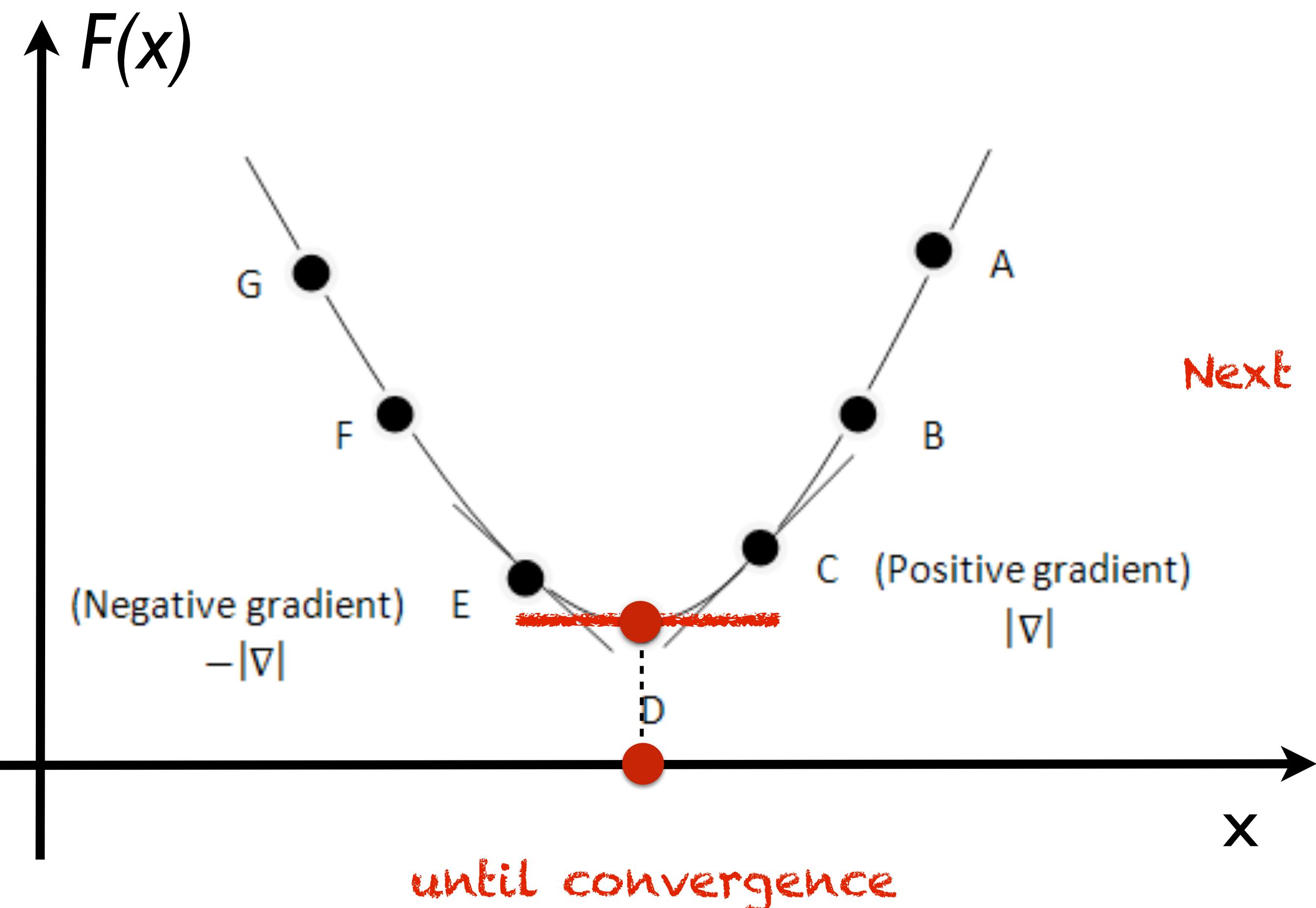
$$x_{i+1} = x_i - \gamma \nabla F(x_i)$$

Next solution

Derivative assumed to be direction of steepest ascent away from goal

Gradient descent

From Wikipedia, the free encyclopedia



Current solution "Learning rate"

$\mathbf{x}_{i+1} = \mathbf{x}_i - \gamma_i \nabla F(\mathbf{x}_i)$

Next solution

Derivative assumed to be direction of steepest ascent away from goal

What is the derivative of
robot configuration?

rate of change of
the endeffector
with respect to

What is the ~~-derivative of -~~
robot configuration?

What is the derivative of
robot configuration?

Geometric Jacobian



Geometric The \checkmark Jacobian

A $6 \times N$ matrix

$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

assuming forward kinematics:

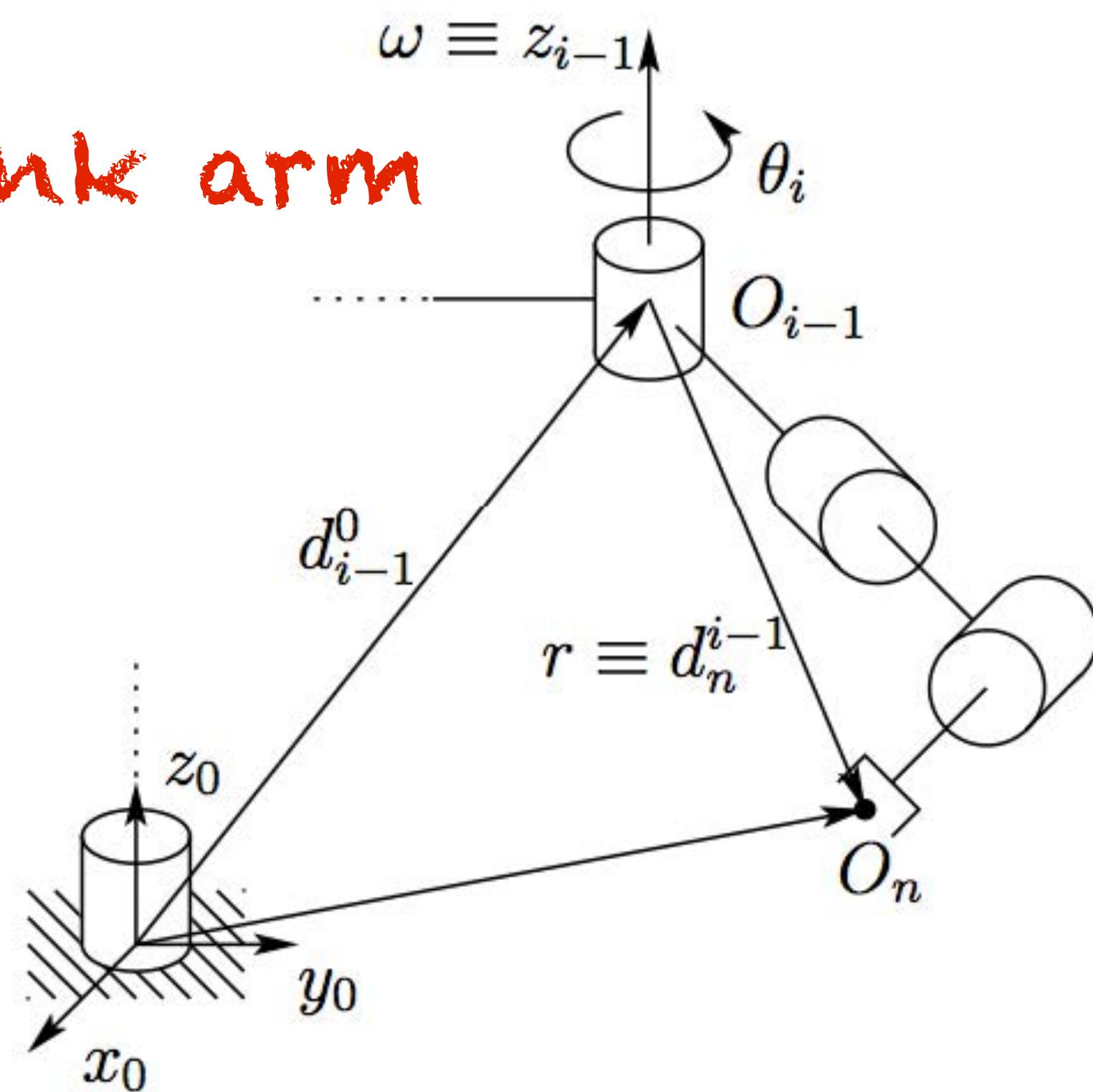
$$\mathbf{x} = f(\mathbf{q})$$

represents partial derivative:

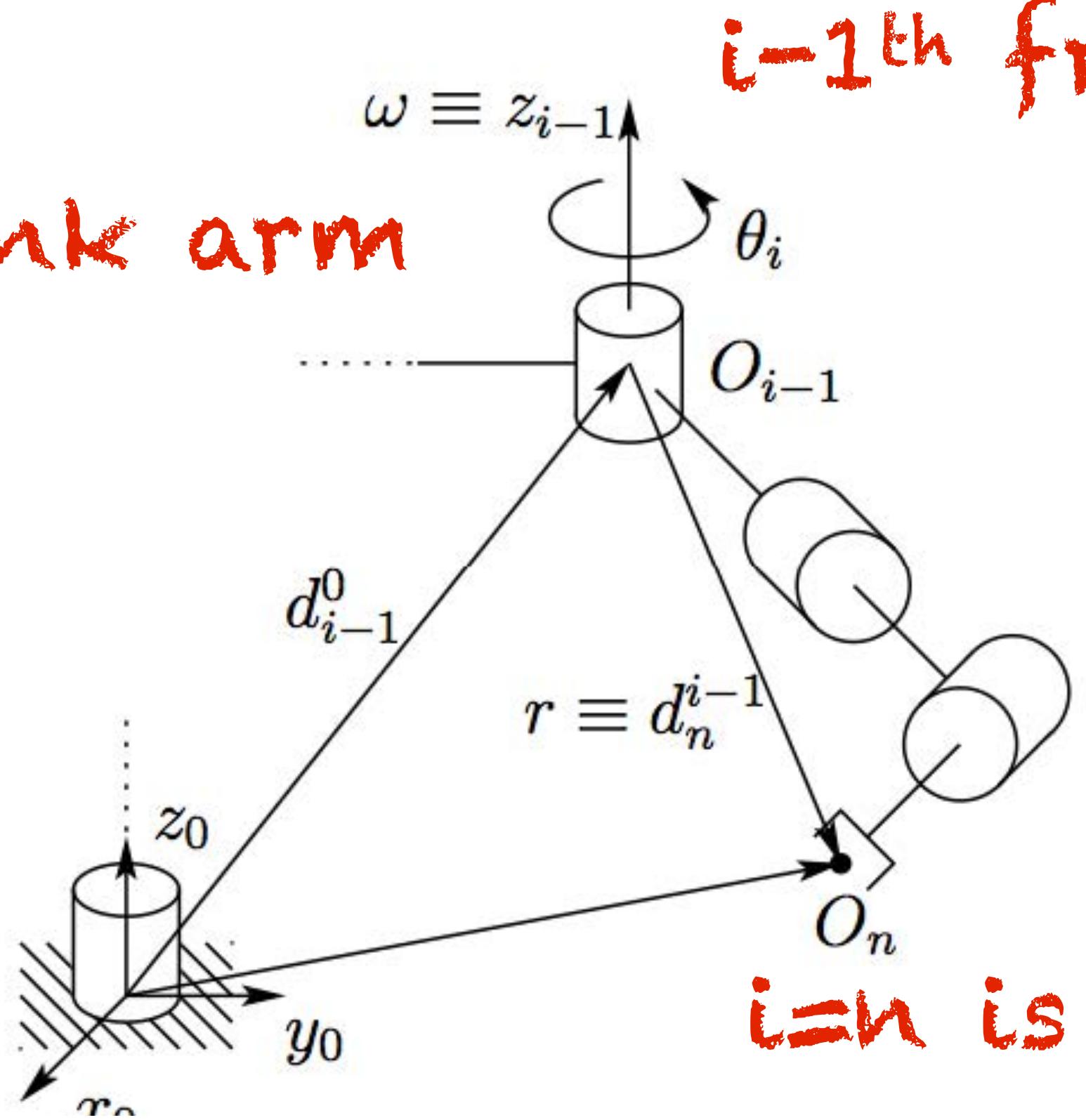
$$\frac{\partial \mathbf{x}}{\partial \mathbf{q}} = \frac{\partial f(\mathbf{q})}{\partial \mathbf{q}} = J(\mathbf{q})$$

Figure 5.1: Motion of the end-effector due to link i .

Each column transforms
velocity at the endeffector
to velocity at a DOF



3D N-Link arm



i=0 is base frame

effector due to link i .

i=n is endeffector frame

i-1th frame maps to ith column

The Jacobian

A 6xN matrix

$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

$$T_n^0(\mathbf{q}) = \begin{bmatrix} R_n^0(\mathbf{q}) & o_n^0(\mathbf{q}) \\ 0 & 1 \end{bmatrix}$$

Note: figure taken from Spong et al.
textbook, which assumes D-H
parameters and offset column index

3D N-Link arm

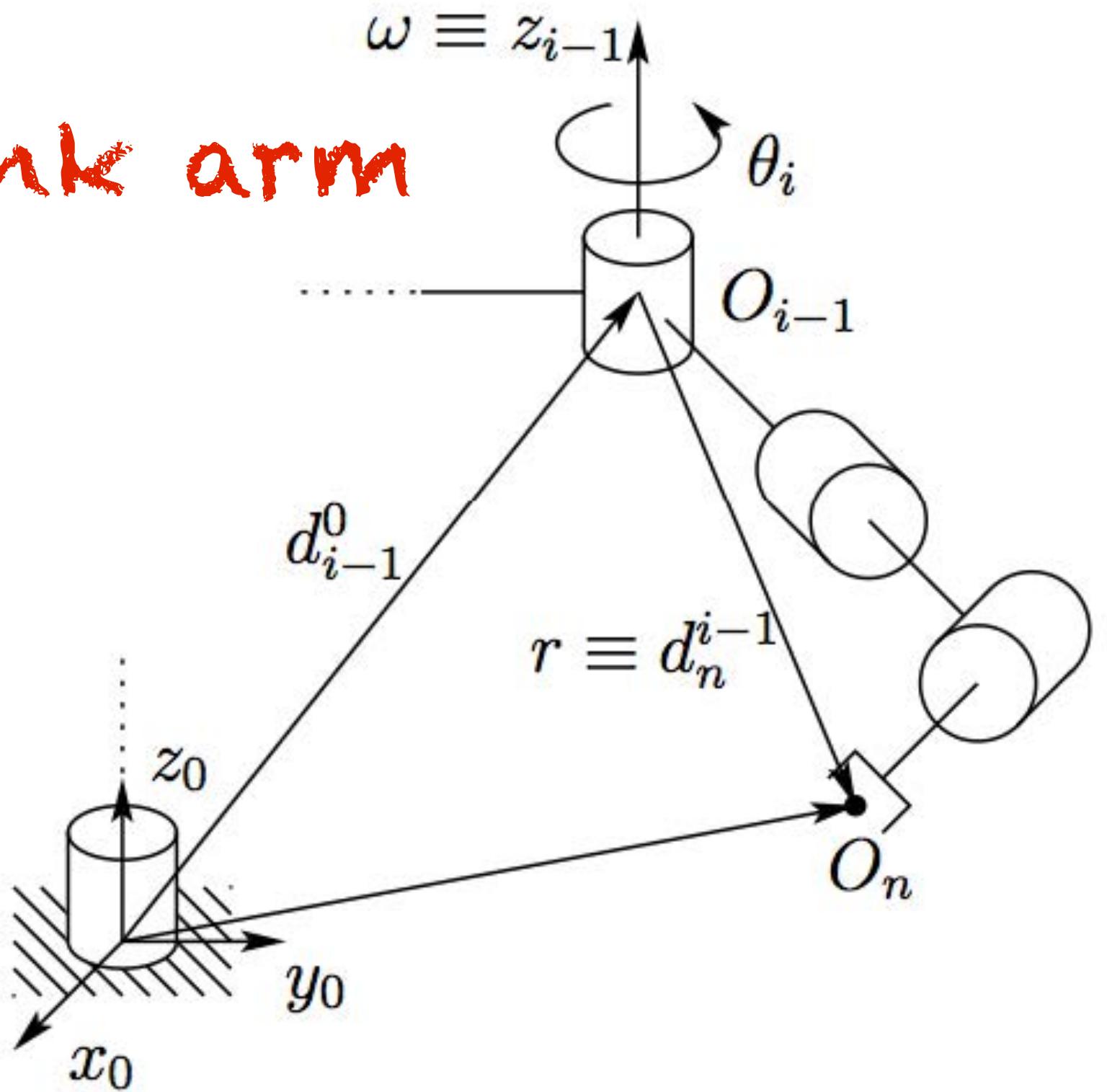


Figure 5.1: Motion of the end-effector due to link i .

$$\frac{\partial F_1}{\partial x_1}$$

Change in an endeffector
variable wrt. change in a
joint variable

The Jacobian

A $6 \times N$ matrix

$$J = [J_1 \ J_2 \ \dots \ J_n]$$

consisting of two $3 \times N$ matrices

$$J = \begin{bmatrix} J_v \\ J_\omega \end{bmatrix}$$

with overall form

$$J = \begin{bmatrix} \frac{\partial F_1}{\partial x_1} & \dots & \frac{\partial F_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial F_m}{\partial x_1} & \dots & \frac{\partial F_m}{\partial x_n} \end{bmatrix}$$

3D N-Link arm

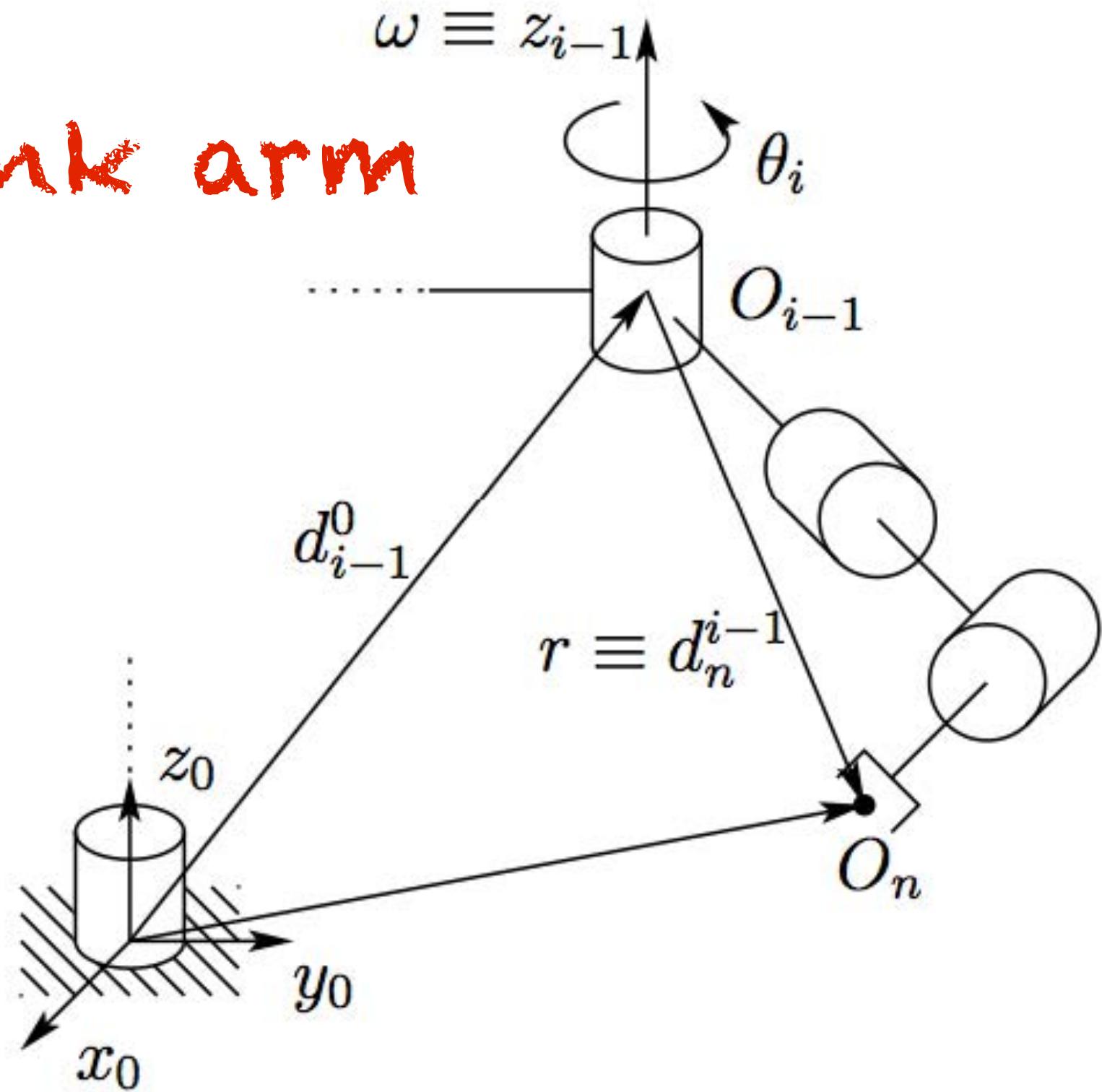


Figure 5.1: Motion of the end-effector due to link i .

The Jacobian

A $6 \times N$ matrix

$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

consisting of two $3 \times N$ matrices

$$J = \begin{bmatrix} J_v \\ J_\omega \end{bmatrix}$$

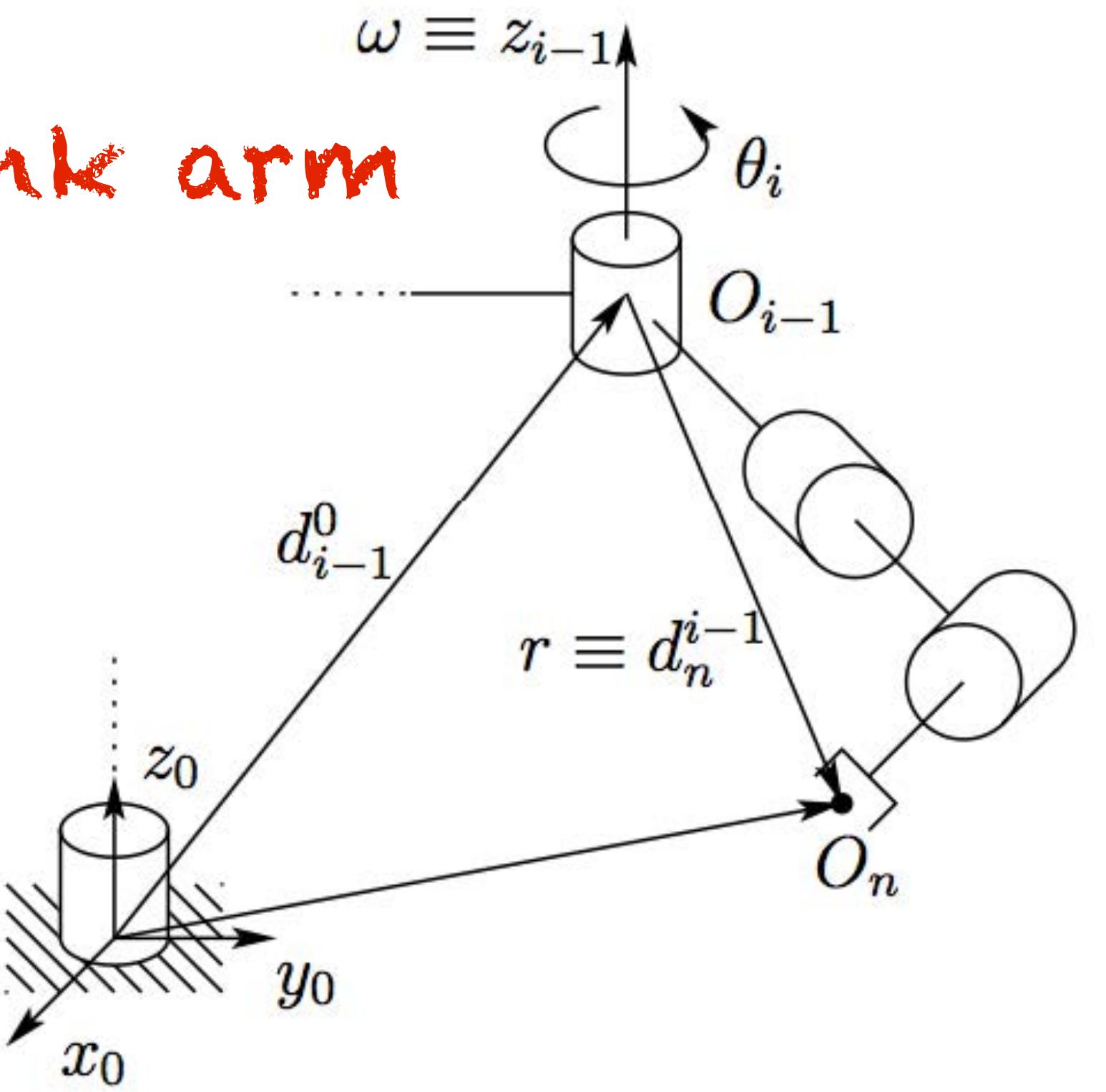
Linear
angular

linear velocity of endeffector $\rightarrow v_n^0 = J_v \dot{q}$

angular velocity of endeffector $\rightarrow \omega_n^0 = J_\omega \dot{q}$

vector of
of joint
angle
velocities

3D N-link arm



The Jacobian

A $6 \times N$ matrix

$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

consisting of two $3 \times N$ matrices

$$J = \begin{bmatrix} J_v \\ J_\omega \end{bmatrix}$$

Linear
angular

Figure 5.1: Motion of the

J_i is a single column
of the Jacobian matrix

J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$

Z is joint axis in
world coordinates
(overloaded notation)

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

vectors in
base frame

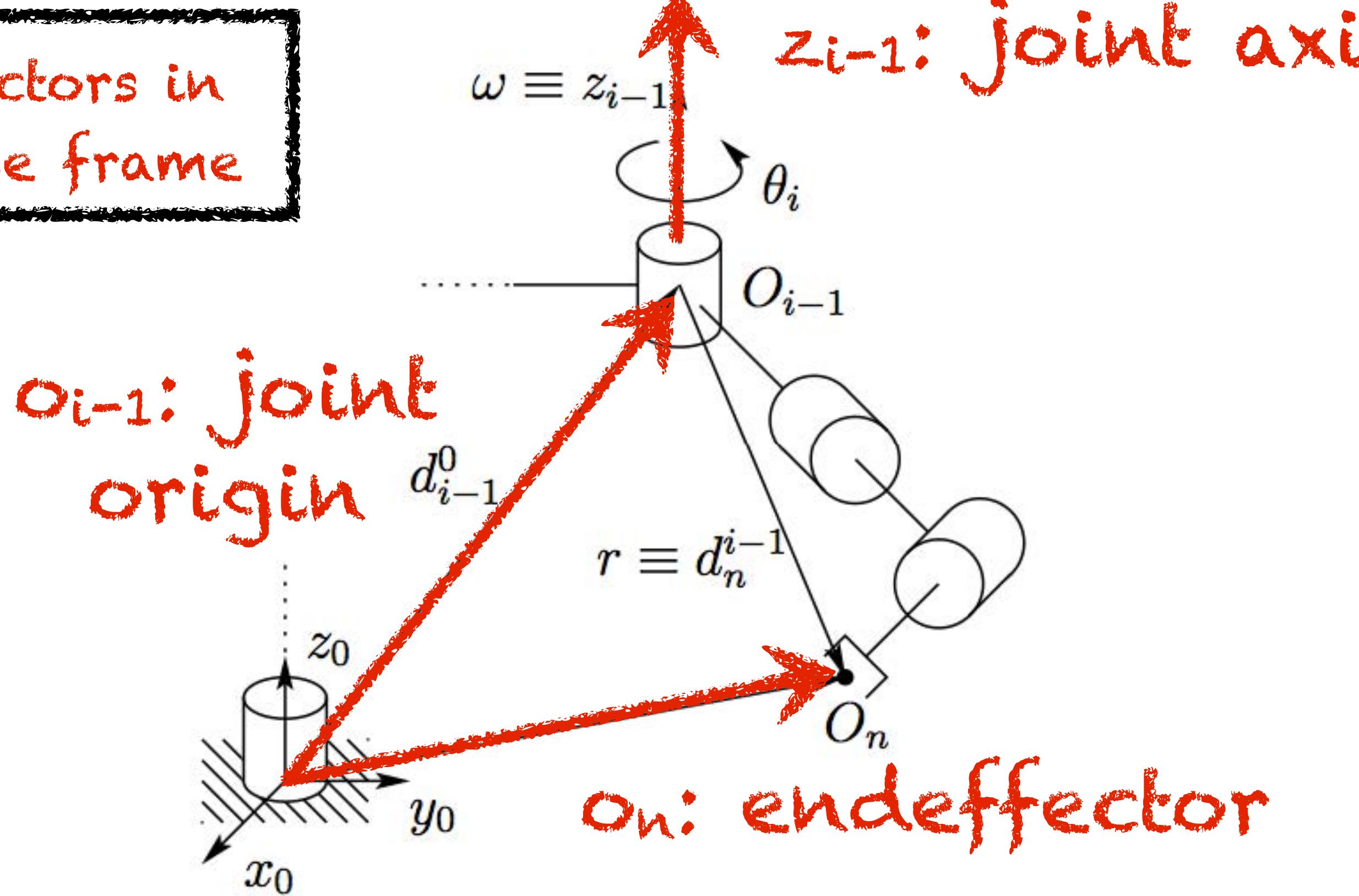


Figure 5.1: Motion of the end-effector due to link i .

The Jacobian

A $6 \times N$ matrix

$$J = [J_1 \ J_2 \ \dots \ J_n]$$

$$J = \begin{bmatrix} J_v \\ J_\omega \end{bmatrix}$$

Linear ←———— angular ←————

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$

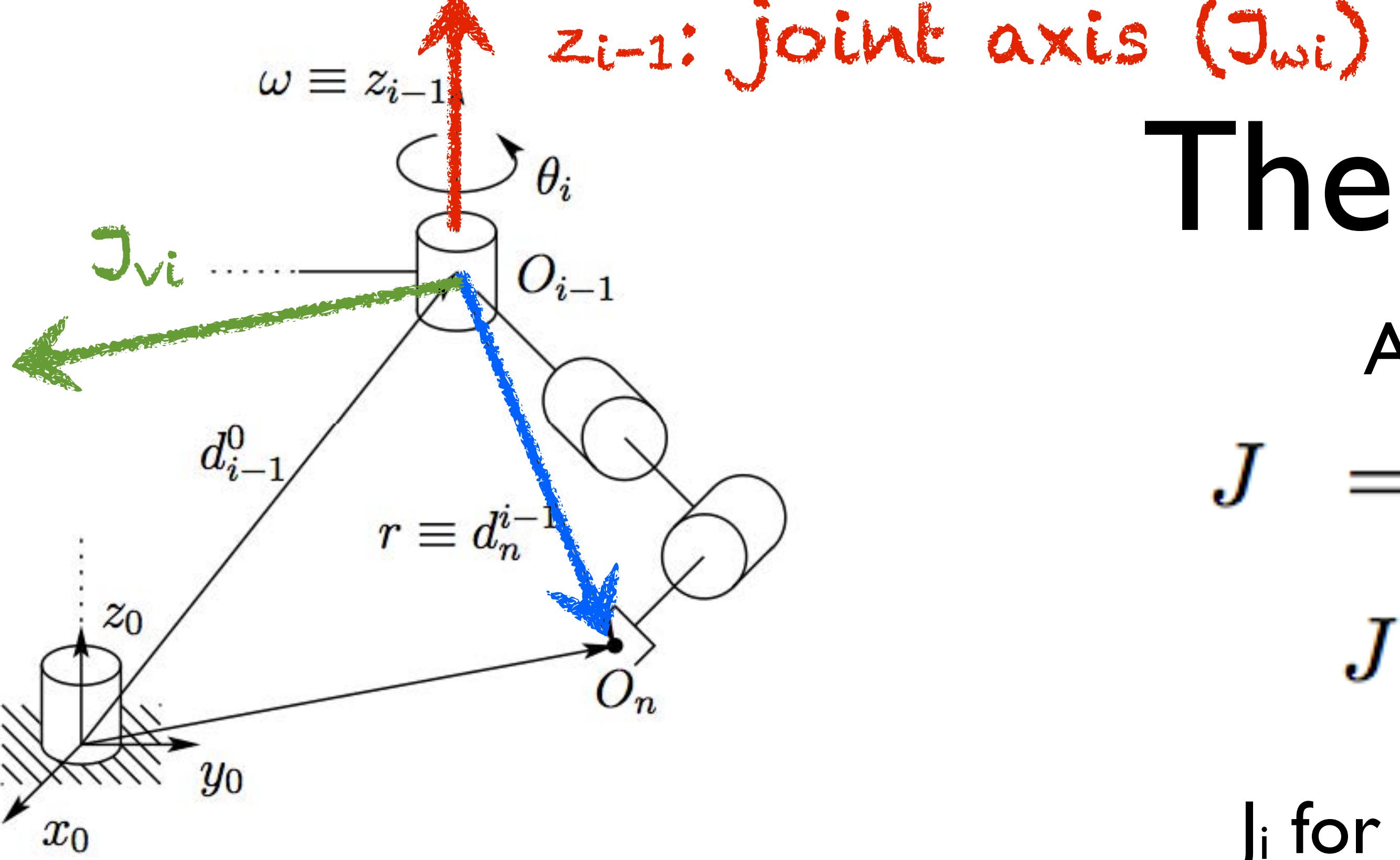


Figure 5.1: Motion of the end-effector due to link i .

The Jacobian

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$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

$$J = \begin{bmatrix} J_v \\ J_\omega \end{bmatrix}$$

Linear

Angular

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$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$

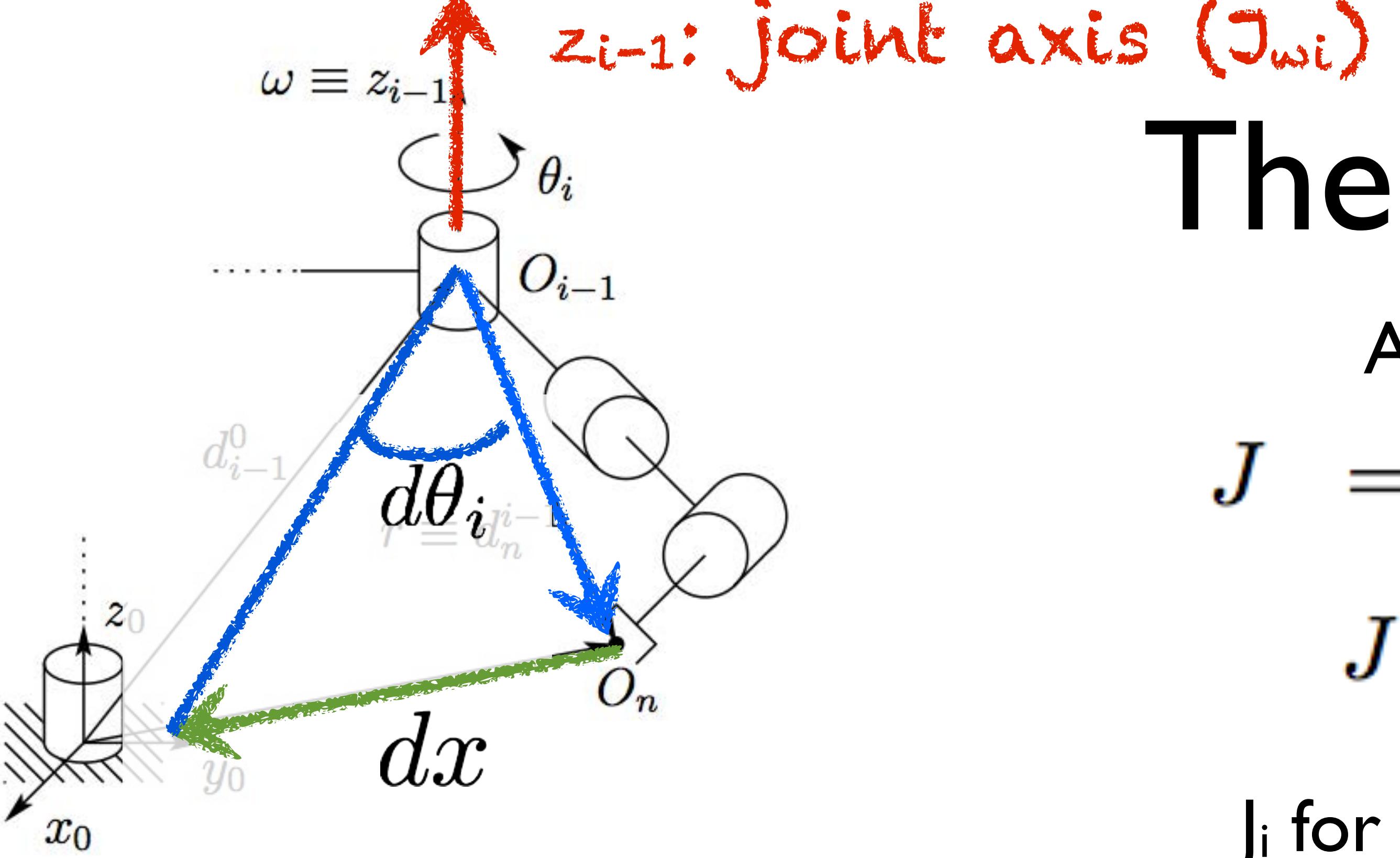


Figure 5.1: Motion of the end-effector due to link i .

The Jacobian

A $6 \times N$ matrix

$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

$$J = \begin{bmatrix} J_v \\ J_\omega \end{bmatrix}$$

← linear
← angular

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$

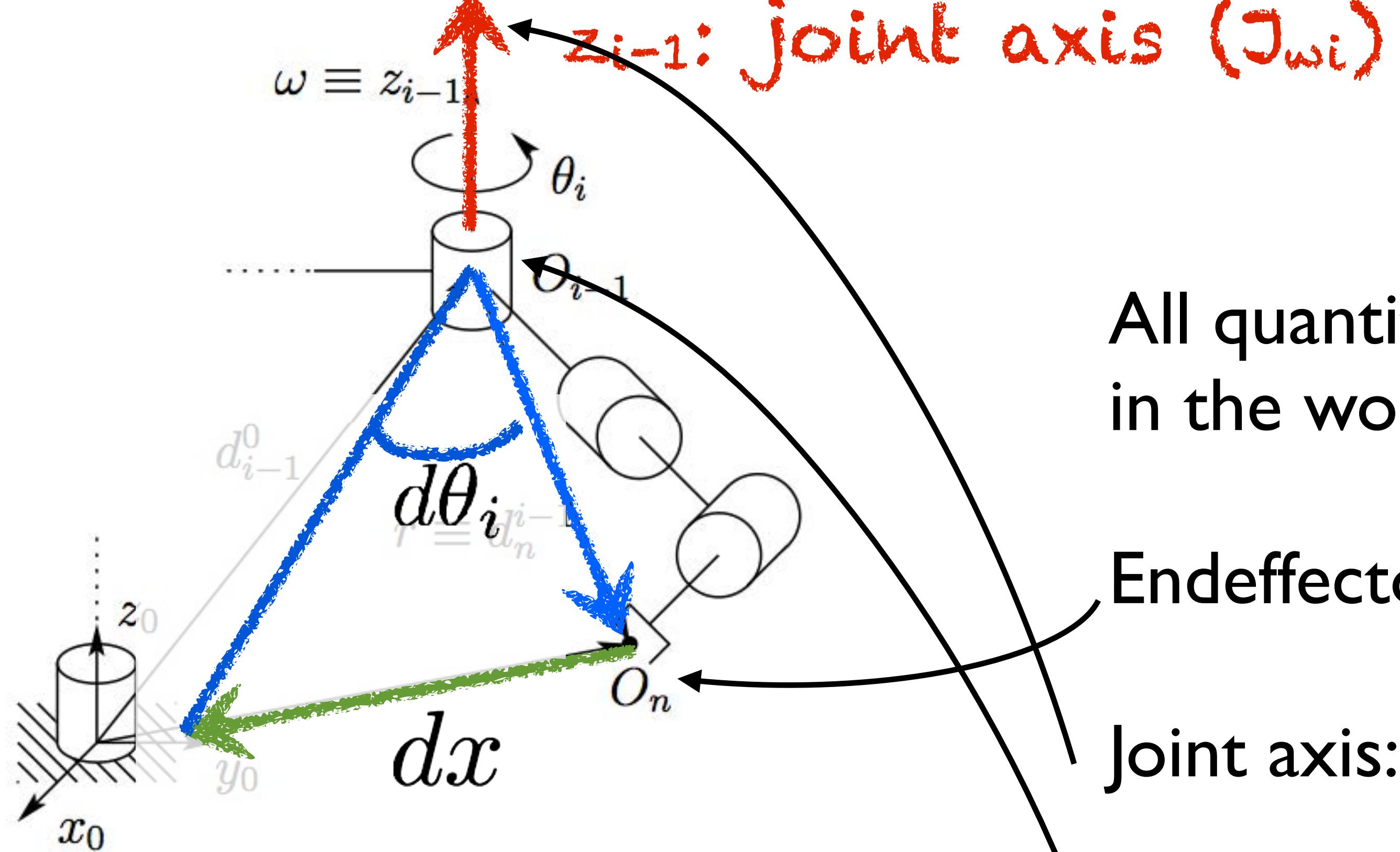


Figure 5.1: Motion of the end-effector due to link i .

Important

All quantities must be expressed in the world frame

$$\text{Endeffector: } \{\mathbf{p}_{\text{tool}}\}^w = T_n^w \{\mathbf{p}_{\text{tool}}\}^n$$

$$\text{Joint axis: } \{\mathbf{k}_i\}^w = T_i^w \{\mathbf{k}_i\}^i$$

$$\text{Joint origin: } \{\mathbf{o}_i\}^w = T_i^w \{\mathbf{o}_i\}^i$$

$$J_{vi} = (\{\mathbf{k}_i\}^w - \{\mathbf{o}_i\}^w) \times (\{\mathbf{p}_{\text{tool}}\}^w - \{\mathbf{o}_i\}^w)$$

$$J_{\omega i} = \{\mathbf{k}_i\}^w - \{\mathbf{o}_i\}^w$$

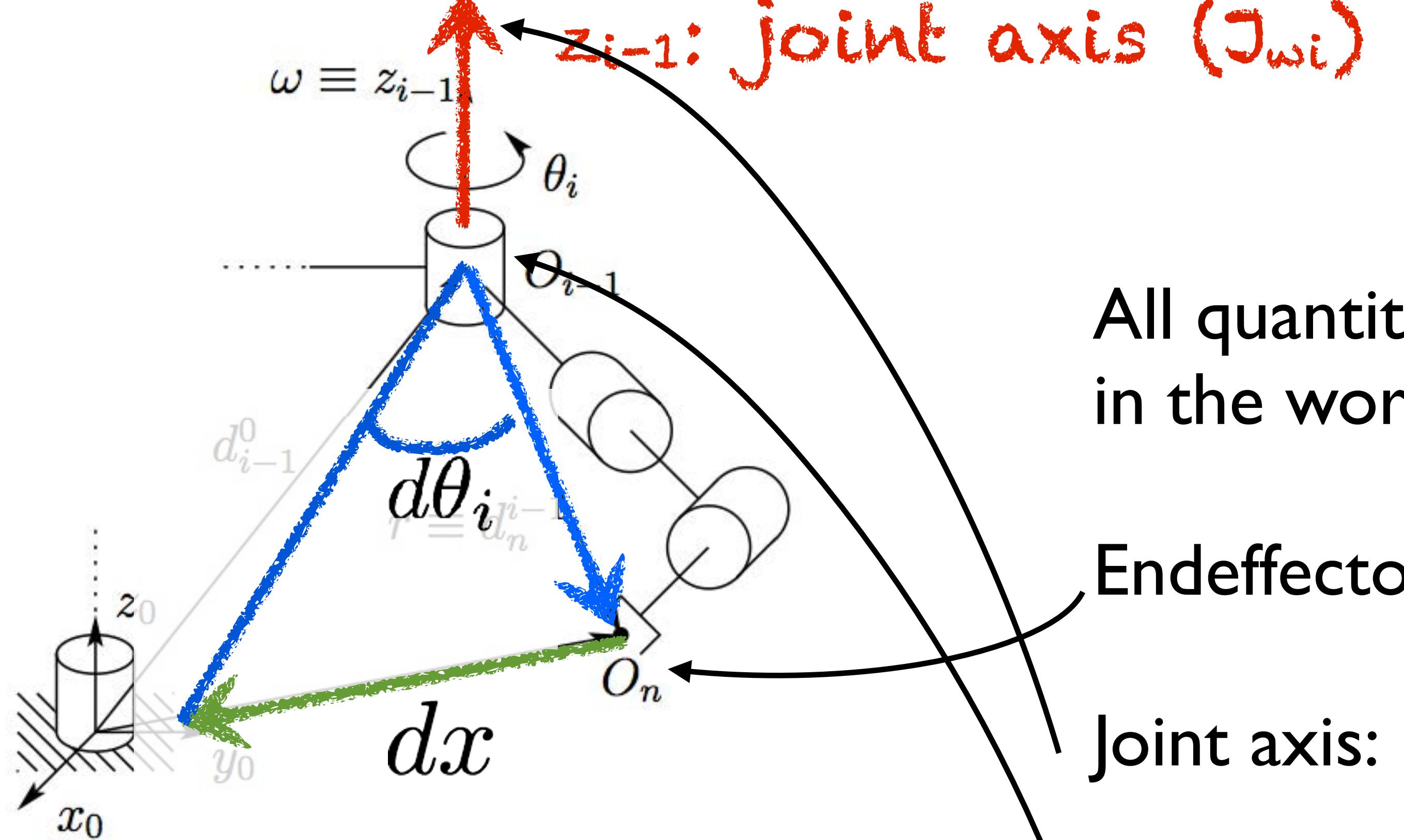


Figure 5.1: Motion of the end-effector due to link i .

Impo

All quantities must be
in the world frame

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$

Endeffector: $\{p_{\text{tool}}\}^w = T_n^w \{p_{\text{tool}}\}^n$

Joint axis: $\{k_i\}^w = T_i^w \{k_i\}^i$

Joint origin: $\{o_i\}^w = T_i^w \{o_i\}^i$

$$J_{vi} = (\{k_i\}^w - \{o_i\}^w) \times (\{p_{\text{tool}}\}^w - \{o_i\}^w)$$

$$J_{\omega i} = \{k_i\}^w - \{o_i\}^w$$

How did we get the Geometric Jacobian?

Velocity of Point Rotating on N-link Arm

$$T_n^0(\mathbf{q}) = \begin{bmatrix} R_n^0(\mathbf{q}) & o_n^0(\mathbf{q}) \\ \mathbf{0} & 1 \end{bmatrix}$$

Angular Velocity

$$R_n^0 = R_1^0 R_2^1 \cdots R_{n-1}^{n-1}$$

assuming velocities expressed in the same frame

$$\omega_n^0 = \omega_1^0 + R_1^0 \omega_2^1 + R_2^0 \omega_3^2 + R_3^0 \omega_4^3 + \cdots + R_{n-1}^0 \omega_n^{n-1}$$

$$z_{i-1}^0 = R_{i-1}^0 \mathbf{k}$$

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

$$J_\omega = [z_0^0 \ z_1^0 \ \dots \ z_{n-1}^0]$$



Velocity of Point Rotating on N-link Arm

$$\begin{aligned} T_n^0(\mathbf{q}) &= \begin{bmatrix} R_n^0(\mathbf{q}) & o_n^0(\mathbf{q}) \\ \mathbf{0} & 1 \end{bmatrix} \\ &= T_n^0 \\ &= T_{i-1}^0 T_i^{i-1} T_n^i \\ &= \begin{bmatrix} R_{i-1}^0 & o_{i-1}^0 \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} R_i^{i-1} & o_i^{i-1} \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} R_n^i & o_n^i \\ \mathbf{0} & 1 \end{bmatrix} \\ &= \begin{bmatrix} R_n^0 & R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1} + o_{i-1}^0 \\ \mathbf{0} & 1 \end{bmatrix} \end{aligned}$$

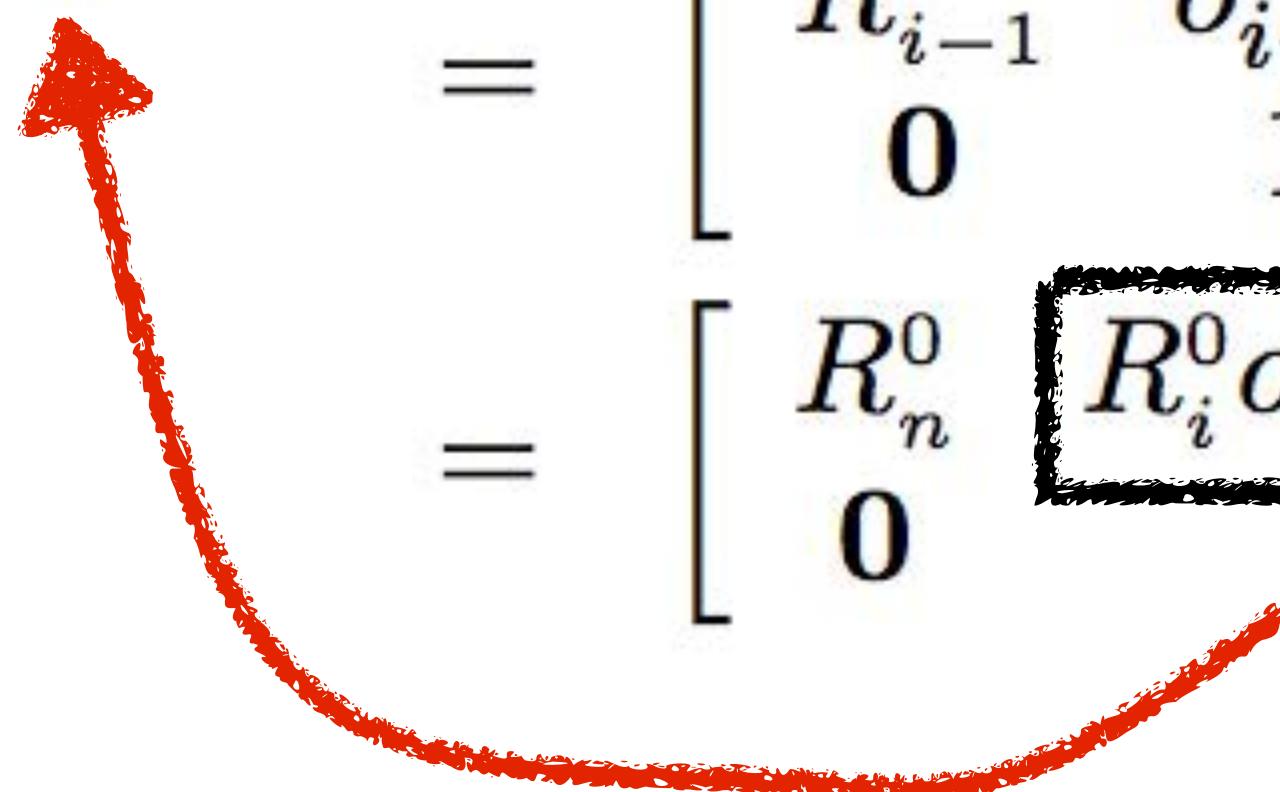
consider effect of all frames
(o..n) on endeffector

Velocity of Point Rotating on N-link Arm

$$T_n^0(\mathbf{q}) = \begin{bmatrix} R_n^0(\mathbf{q}) & o_n^0(\mathbf{q}) \\ \mathbf{0} & 1 \end{bmatrix}$$
$$= T_n^0$$

Linear Velocity for Rotational Joint

$$o_n^0 = R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1} + o_{i-1}^0$$



$$= T_{i-1}^0 T_i^{i-1} T_n^i$$
$$= \begin{bmatrix} R_{i-1}^0 & o_{i-1}^0 \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} R_i^{i-1} & o_i^{i-1} \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} R_n^i & o_n^i \\ \mathbf{0} & 1 \end{bmatrix}$$
$$= \begin{bmatrix} R_n^0 & R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1} + o_{i-1}^0 \\ \mathbf{0} & 1 \end{bmatrix}$$

position of endeffector frame

Velocity of Point Rotating on N-link Arm

$$T_n^0(\boldsymbol{q}) = \begin{bmatrix} R_n^0(\boldsymbol{q}) & o_n^0(\boldsymbol{q}) \\ 0 & 1 \end{bmatrix}$$
$$= T_n^0$$

Linear Velocity for Rotational Joint

$$o_n^0 = R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1} + o_{i-1}^0$$

$$\frac{\partial}{\partial \theta_i} o_n^0 = \frac{\partial}{\partial \theta_i} [R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1}]$$

take derivative wrt.
ith joint angle

$$= T_{i-1}^0 T_i^{i-1} T_n^i$$
$$= \begin{bmatrix} R_{i-1}^0 & o_{i-1}^0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} R_i^{i-1} & o_i^{i-1} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} R_n^i & o_n^i \\ 0 & 1 \end{bmatrix}$$
$$= \begin{bmatrix} R_n^0 & R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1} + o_{i-1}^0 \\ 0 & 1 \end{bmatrix}$$

Velocity of Point Rotating on N-link Arm

$$T_n^0(\boldsymbol{q}) = \begin{bmatrix} R_n^0(\boldsymbol{q}) & o_n^0(\boldsymbol{q}) \\ 0 & 1 \end{bmatrix}$$
$$= T_n^0$$

Linear Velocity for Rotational Joint

$$o_n^0 = R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1} + o_{i-1}^0$$

$$\frac{\partial}{\partial \theta_i} o_n^0 = \frac{\partial}{\partial \theta_i} [R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1}]$$

$$= \dot{\theta}_i S(z_{i-1}^0) R_i^0 o_n^i + \dot{\theta}_i S(z_{i-1}^0) R_{i-1}^0 o_i^{i-1}$$

$$= \dot{\theta}_i z_{i-1}^0 \times (o_n^0 - o_{i-1}^0)$$

$$Jv_i = z_{i-1} \times (o_n - o_{i-1})$$

$$= T_{i-1}^0 T_i^{i-1} T_n^i$$
$$= \begin{bmatrix} R_{i-1}^0 & o_{i-1}^0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} R_i^{i-1} & o_i^{i-1} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} R_n^i & o_n^i \\ 0 & 1 \end{bmatrix}$$
$$= \begin{bmatrix} R_n^0 & R_i^0 o_n^i + R_{i-1}^0 o_i^{i-1} + o_{i-1}^0 \\ 0 & 1 \end{bmatrix}$$

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

IK Procedure Restated



Geometric The ~~J~~Jacobian

A $6 \times N$ matrix

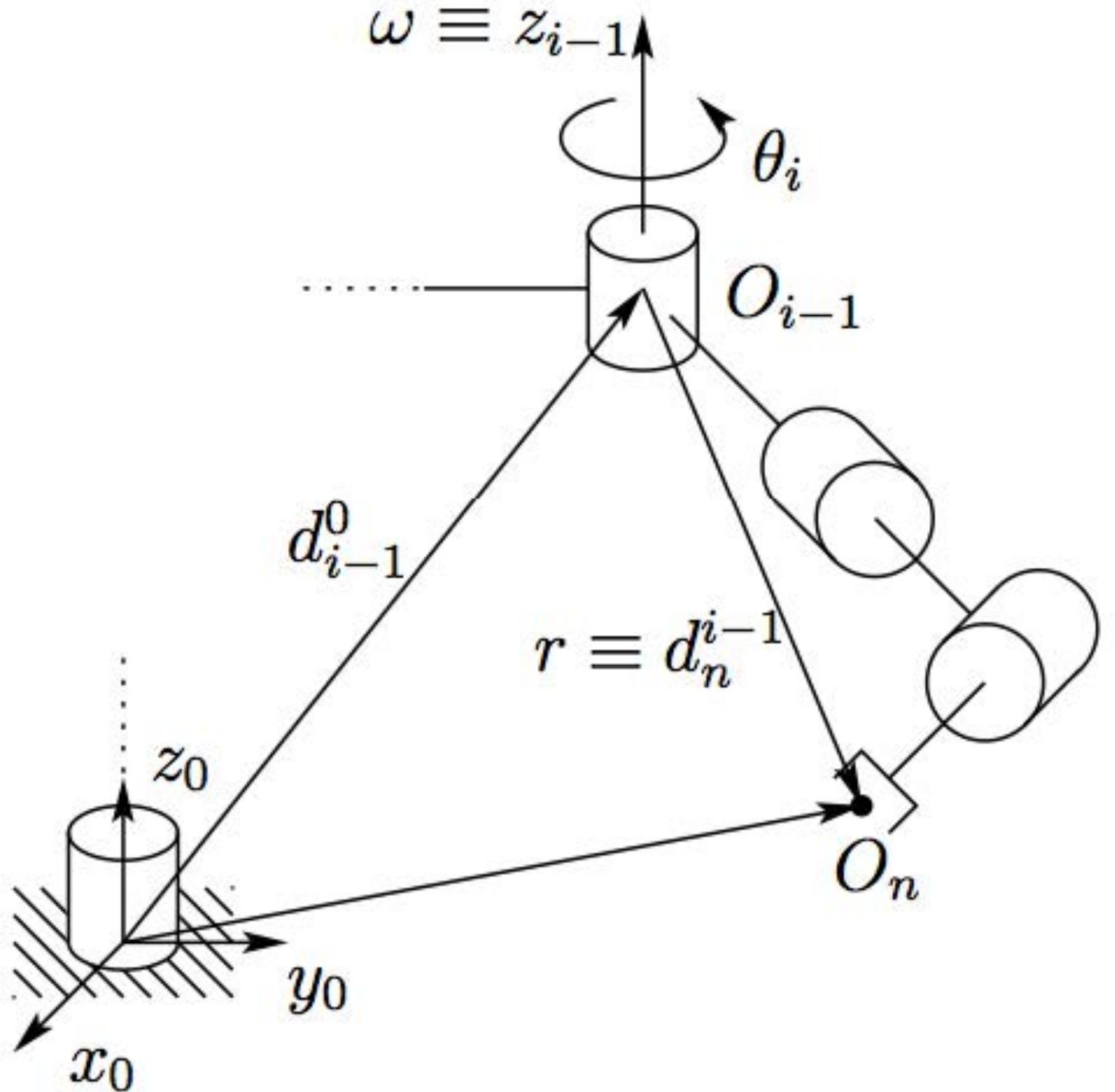
$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

J_i for a rotational joint

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J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$



IK Procedure restated:

$$\Delta \mathbf{x}_n = \mathbf{x}_d - \mathbf{x}_n$$

$$\Delta \mathbf{q}_n = J(\mathbf{q}_n)^{-1} \Delta \mathbf{x}_n$$

$$\mathbf{q}_{n+1} = \mathbf{q}_n + \gamma \Delta \mathbf{q}_n$$

Geometric The ~~J~~Jacobian

A $6 \times N$ matrix

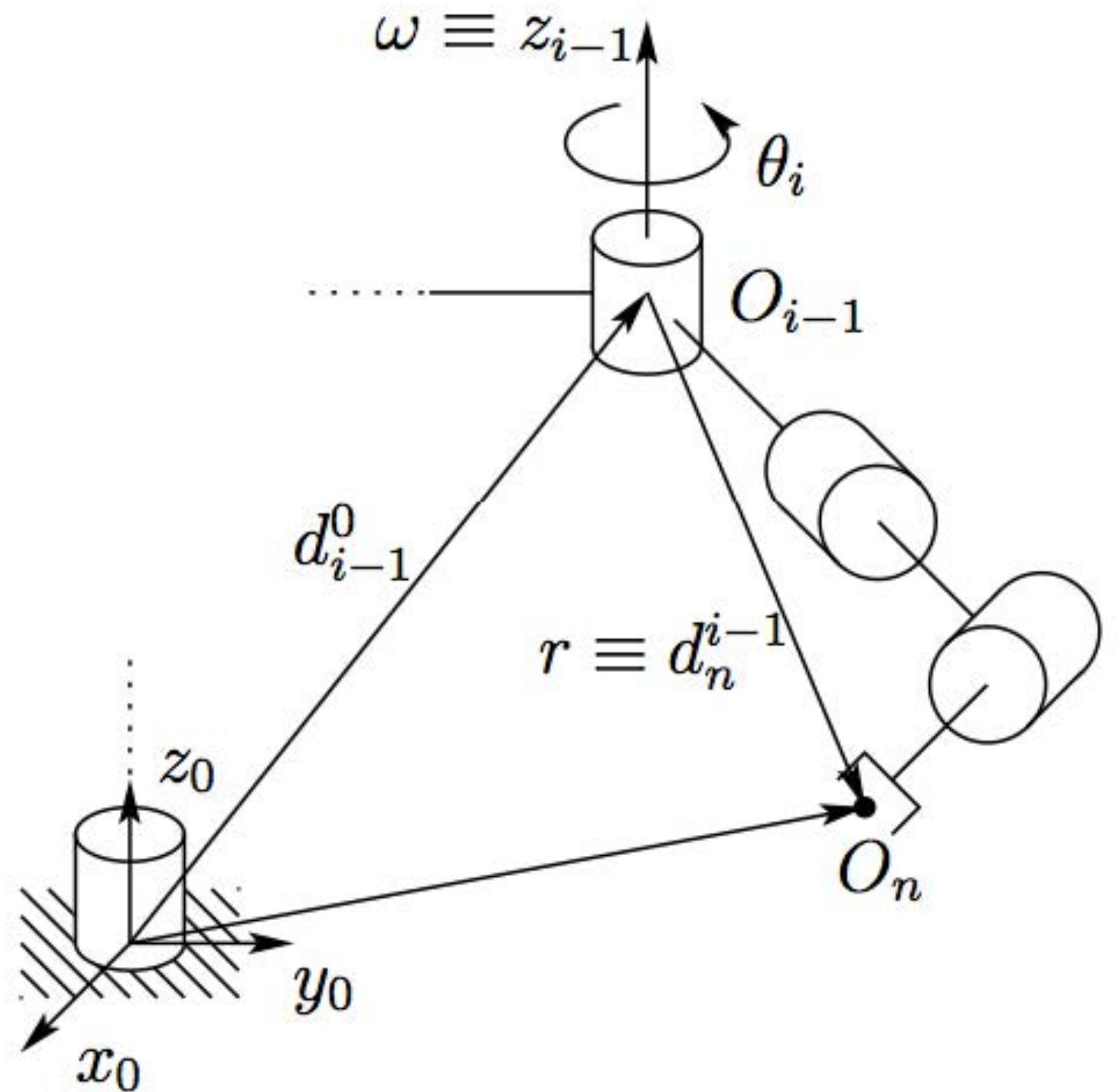
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$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

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compute endpoint error

IK Procedure restated:

$$\Delta \mathbf{x}_n = \mathbf{x}_d - \mathbf{x}_n$$

$$\Delta \mathbf{q}_n = J(\mathbf{q}_n)^{-1} \Delta \mathbf{x}_n$$

$$\mathbf{q}_{n+1} = \mathbf{q}_n + \gamma \Delta \mathbf{q}_n$$

Geometric The \checkmark Jacobian

A $6 \times N$ matrix

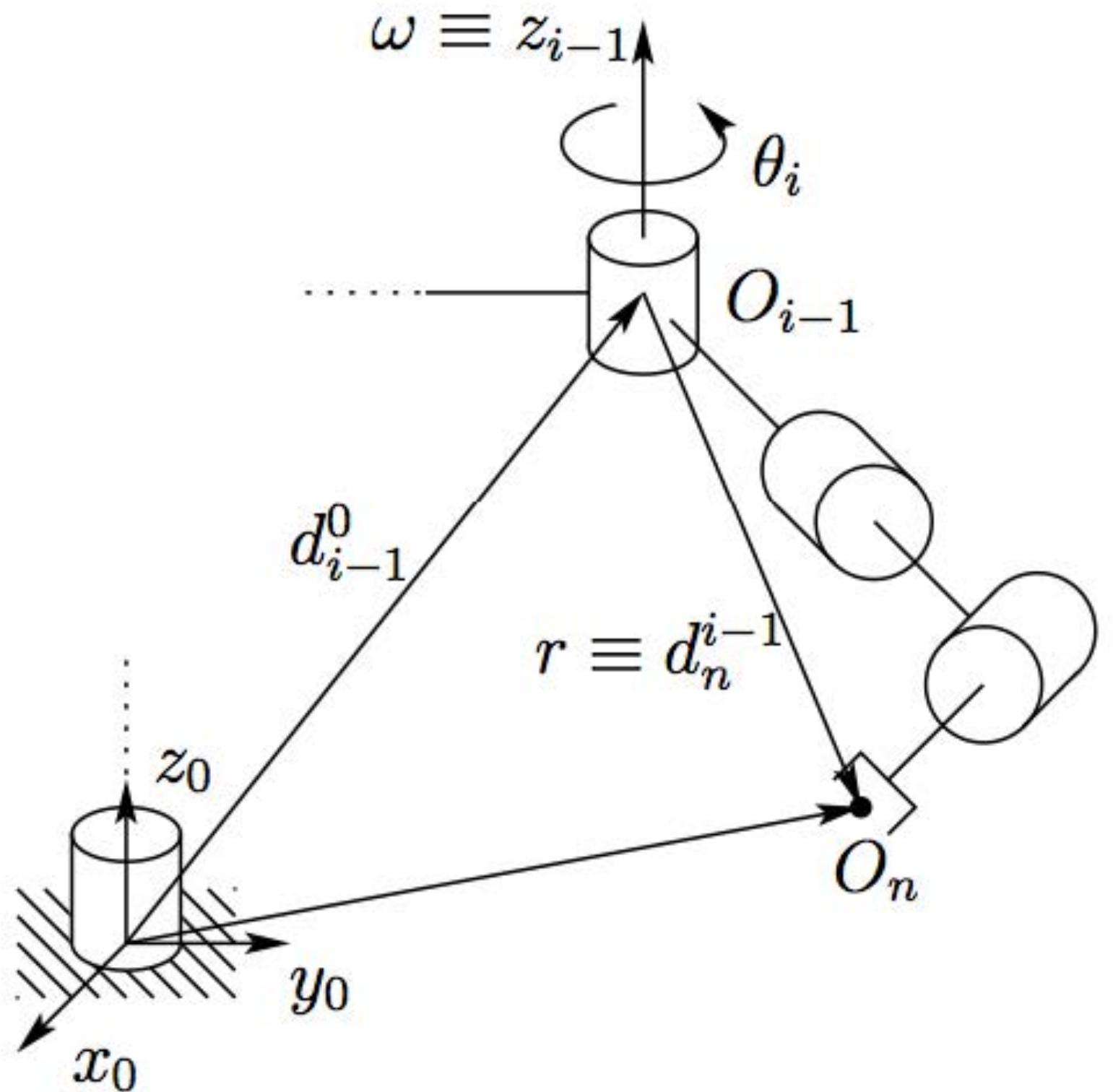
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J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$



compute endpoint error

IK Procedure restated:

compute step direction $\Delta \mathbf{x}_n = \mathbf{x}_d - \mathbf{x}_n$

$$\Delta \mathbf{q}_n = J(\mathbf{q}_n)^{-1} \Delta \mathbf{x}_n$$

$$\mathbf{q}_{n+1} = \mathbf{q}_n + \gamma \Delta \mathbf{q}_n$$

Geometric The \checkmark Jacobian

A $6 \times N$ matrix

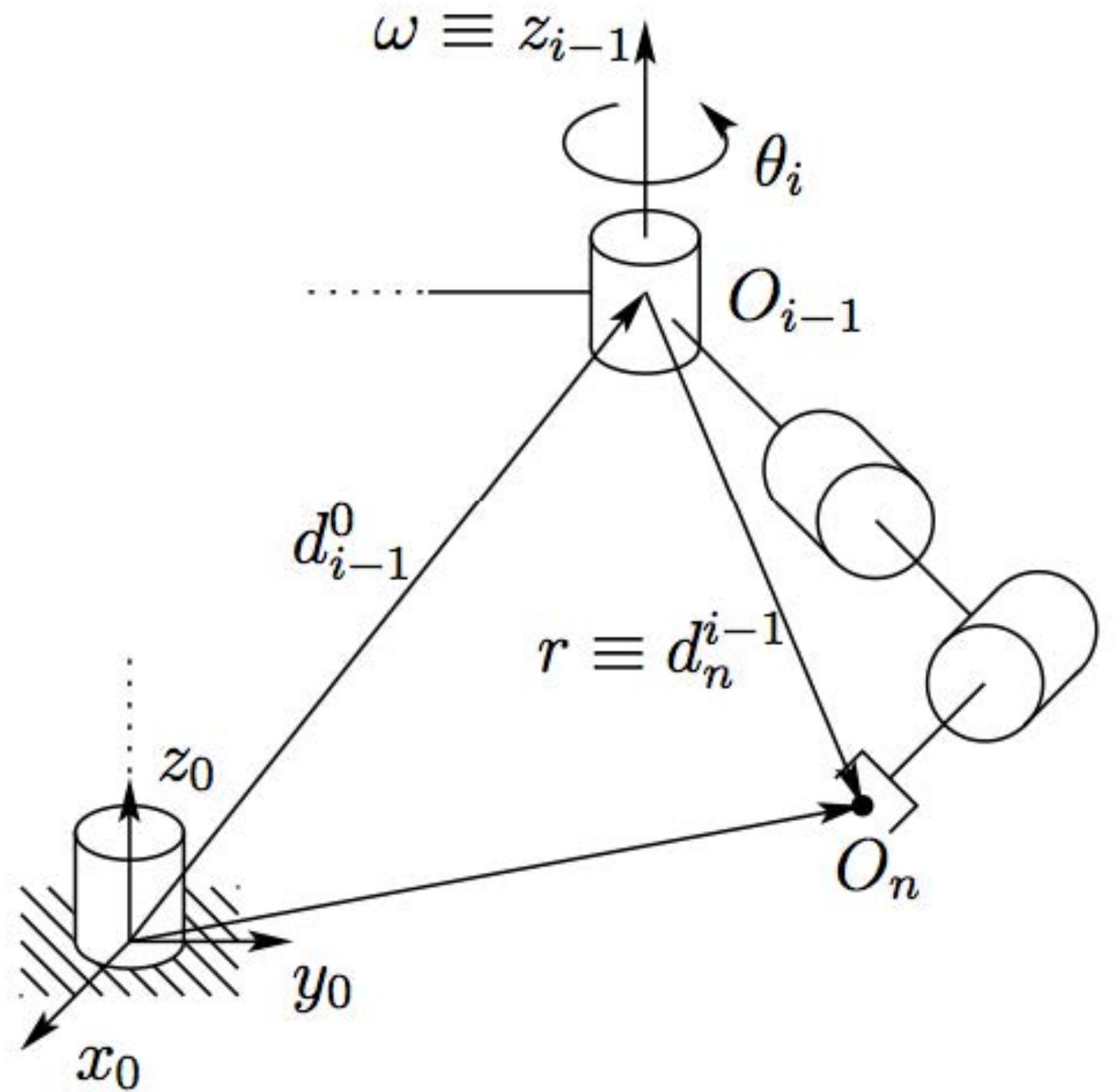
$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$



compute endpoint error

IK Procedure restated:

compute step direction $\Delta \mathbf{x}_n = \mathbf{x}_d - \mathbf{x}_n$

$$\Delta \mathbf{q}_n = J(\mathbf{q}_n)^{-1} \Delta \mathbf{x}_n$$

perform step direction $\mathbf{q}_{n+1} = \mathbf{q}_n + \gamma \Delta \mathbf{q}_n$

Geometric The Jacobian

A $6 \times N$ matrix

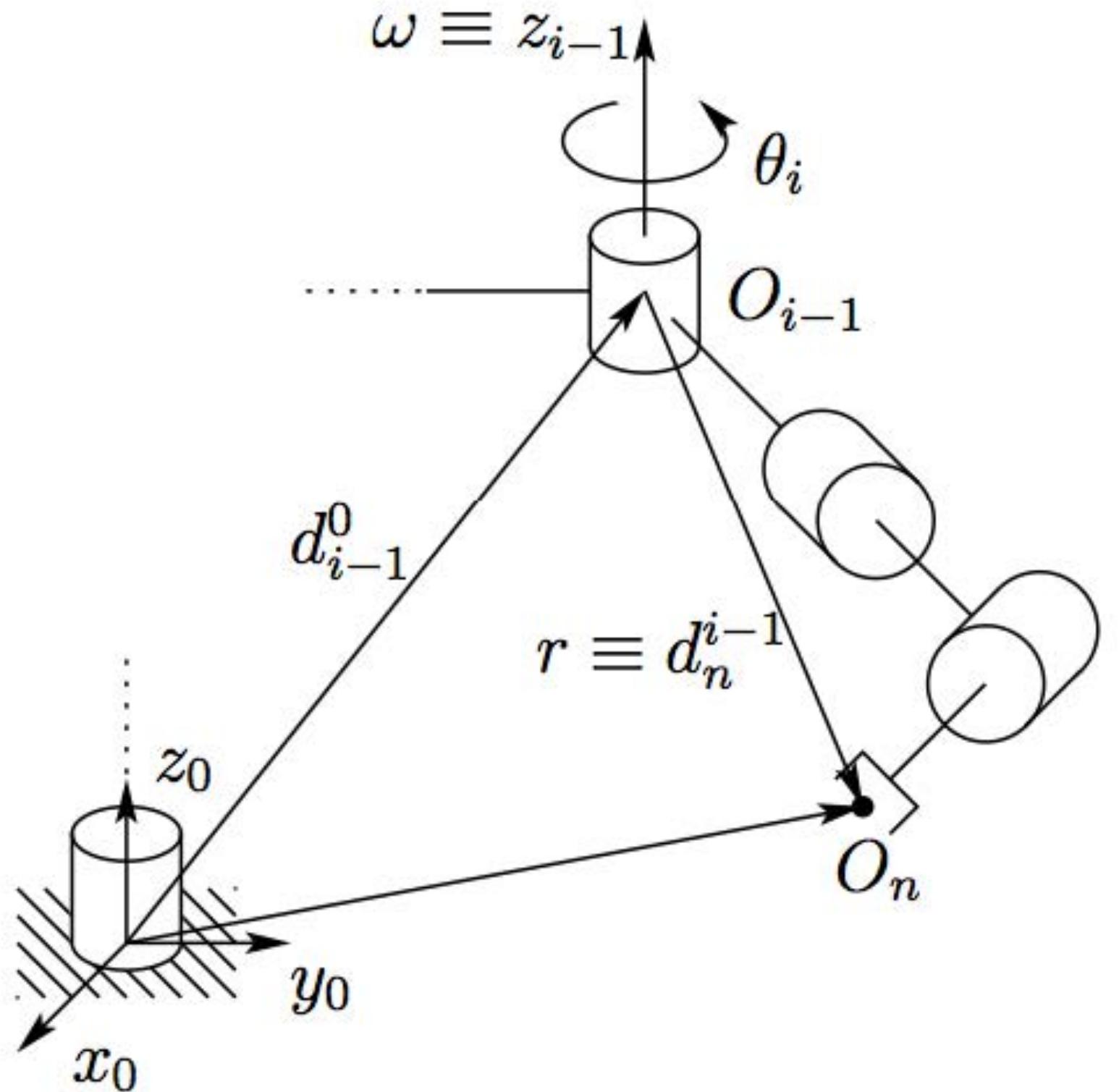
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J_i for a prismatic joint

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compute endpoint error

IK Procedure restated:

compute step direction

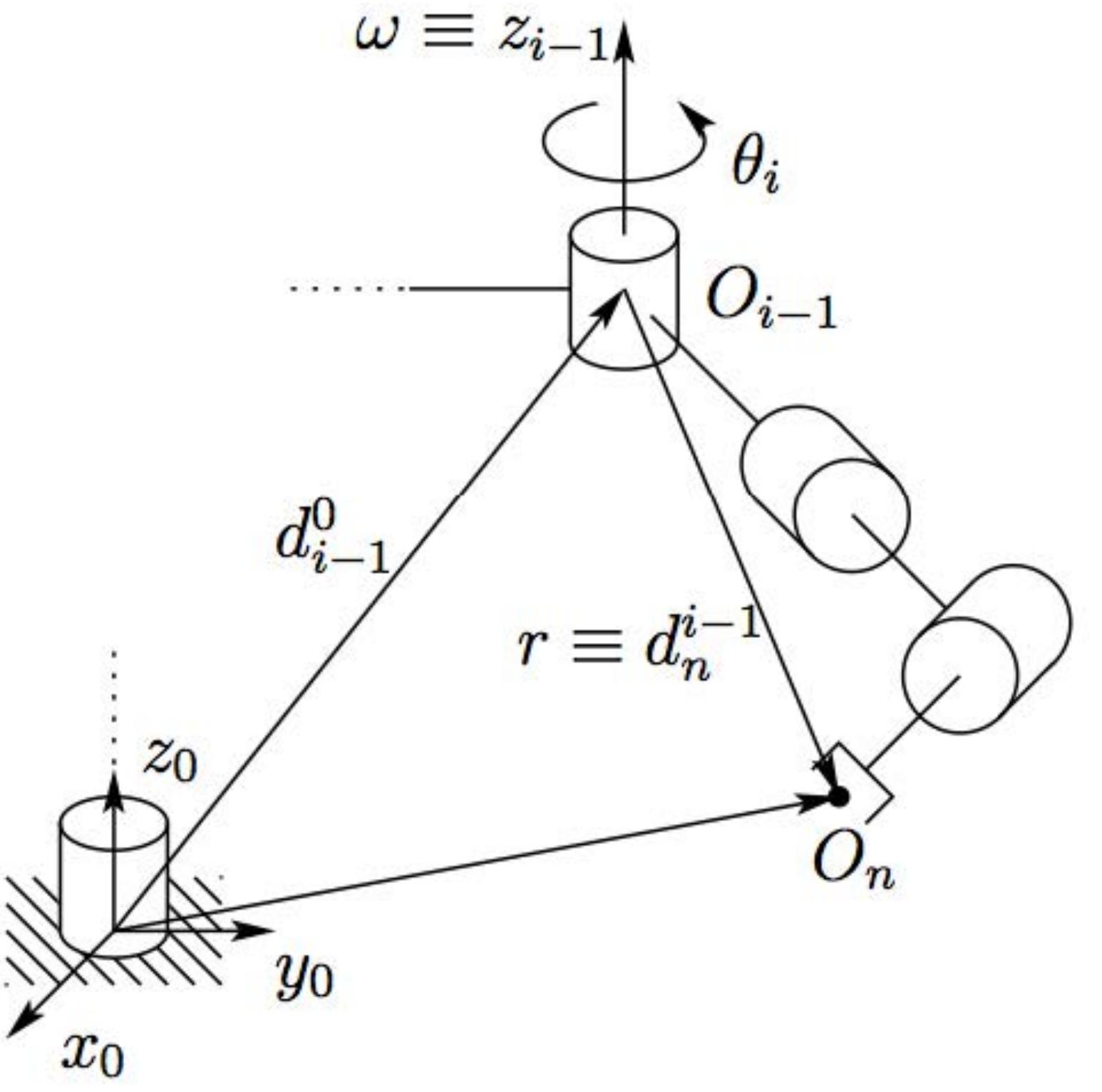
$$\Delta \mathbf{x}_n = \mathbf{x}_d - \mathbf{x}_n$$

$$\Delta \mathbf{q}_n = J(\mathbf{q}_n)^{-1} \Delta \mathbf{x}_n$$

perform step direction

$$\mathbf{q}_{n+1} = \mathbf{q}_n + \gamma \Delta \mathbf{q}_n$$

repeat



The Jacobian

A $6 \times N$ matrix

$$J = [J_1 \ J_2 \ \cdots \ J_n]$$

J_i for a rotational joint

$$J_i = \begin{bmatrix} z_{i-1} \times (o_n - o_{i-1}) \\ z_{i-1} \end{bmatrix}$$

J_i for a prismatic joint

$$J_i = \begin{bmatrix} z_{i-1} \\ 0 \end{bmatrix}$$

when can we invert $J(q)$?

$$\Delta \mathbf{x}_n = \mathbf{x}_d - \mathbf{x}_n$$

$$\Delta \mathbf{q}_n = J(\mathbf{q}_n)^{-1} \Delta \mathbf{x}_n$$

$$\mathbf{q}_{n+1} = \mathbf{q}_n + \gamma \Delta \mathbf{q}_n$$

Which Pseudoinverse

- For matrix A with dimensions $N \times M$ with full rank
- Left pseudoinverse, for when $N > M$, (i.e., “tall”, less than than 6 DoFs)

$$A_{\text{left}}^{-1} = (A^T A)^{-1} A^T \quad \text{s.t.} \quad A_{\text{left}}^{-1} A = I_n$$

- Right pseudoinverse, for when $N < M$, (i.e., “wide”, more than 6 DoFs)

$$A_{\text{right}}^{-1} = A^T (A A^T)^{-1} \quad \text{s.t.} \quad A A_{\text{right}}^{-1} = I_m$$



Maybe there is a simpler
approach to IK?

Next lecture:
IK continued ...
Manipulation New Frontiers