

# Animation and Kinematics



Some Slides/Images adapted from Marschner and Shirley and David Levin

# Animation and Kinematics

## Agenda:

- Animation in Computer Graphics
- Forward Kinematics
- Skinning for Mesh Deformation
- Keyframe Animation + Splines
- Inverse Kinematics

# **“Core” Areas of Computer Graphics**

Modeling/Geometry

Rendering

Animation

# Animation Timeline

1908: Emile Cohl (1857-1938) France, makes his first film, FANTASMAGORIE, arguably the first animated film.

1911: Winsor McCay (1867-1934) makes his first film, LITTLE NEMO. McCay, already famous for comic strips, used the film in his vaudeville act. His advice on animation:

*Any idiot that wants to make a couple of thousand drawings for a hundred feet of film is welcome to join the club.*

1928: Walter Disney (1901-1966) working at the Kansas City Slide Company creates Mickey Mouse.

1974: First Computer animated film “Faim” from NFB nominated for an Oscar.

# Animation Principles

Squash & Stretch

Timing

Ease-In & Ease-Out

Arcs

Anticipation

Follow-through & Secondary  
Motion

Overlapping Action & Asymmetry

Exaggeration

Staging

Appeal

Straight-Ahead vs. Pose-to-Pose

# Case Study: Squash and Stretch

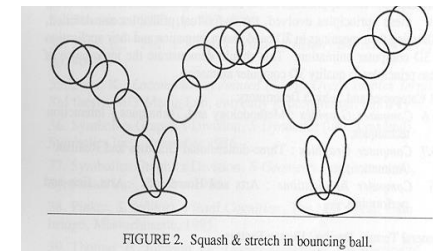
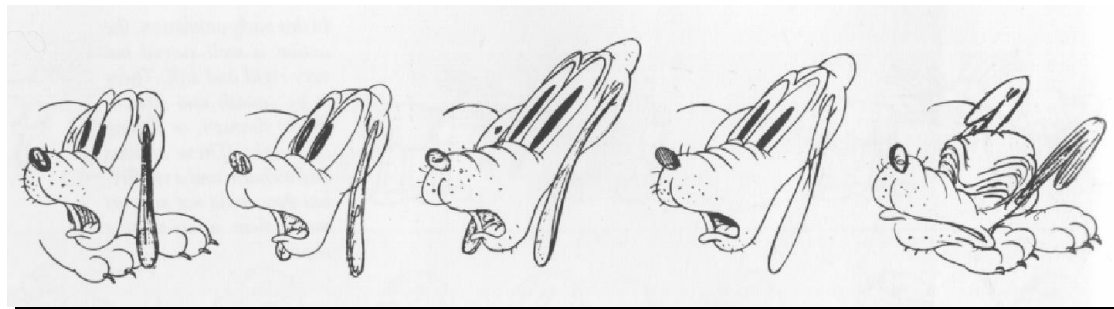
Rigid objects look robotic: deformations make motion natural

Accounts for physics of deformation

- Think squishy ball...
- Communicates to viewer what the object is made of, how heavy it is, ...
- Usually large deformations conserve volume: if you squash one dimension, stretch in another to keep mass constant

Also accounts for persistence of vision

- Fast moving objects leave an elongated streak on our retinas



# What can be animated?

- Lights
- Camera
- **Jointed figures**
- **Skin/muscles**
- **Deformable objects**
- Clothing
- Wind/water/fire/smoke
- Hair

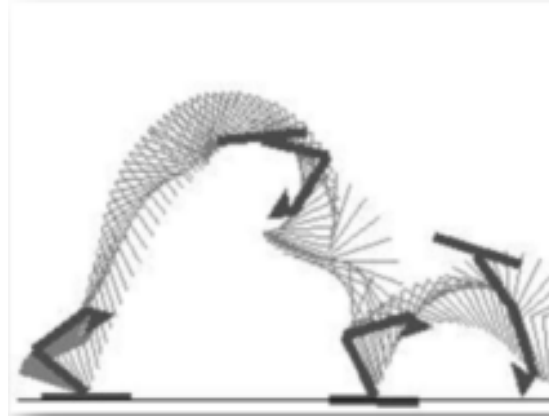
...any variable, Given the right time scale, almost anything...

# Approaches to Animation

How does one make digital models move?



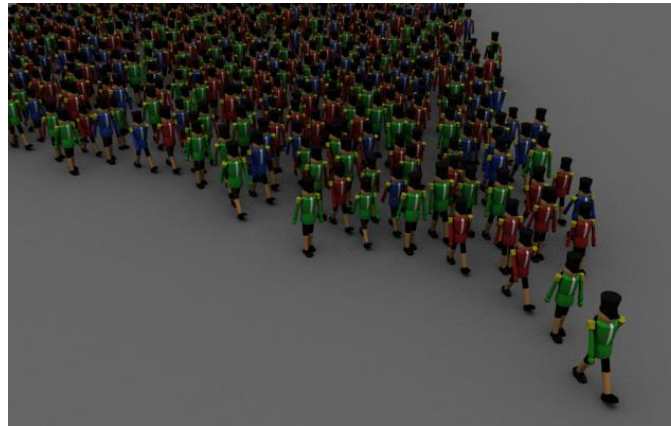
Keyframing



Physical simulation



Motion capture

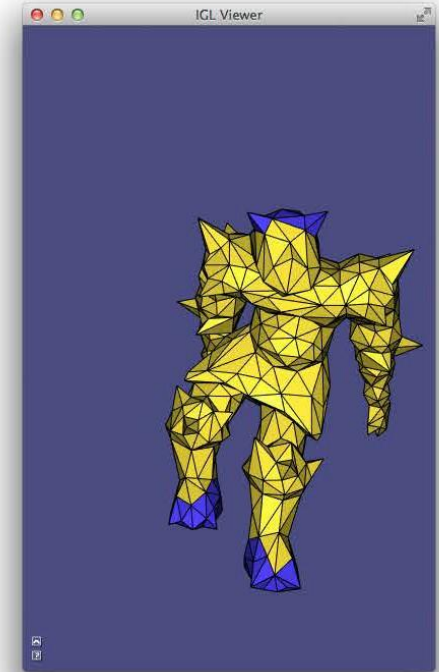
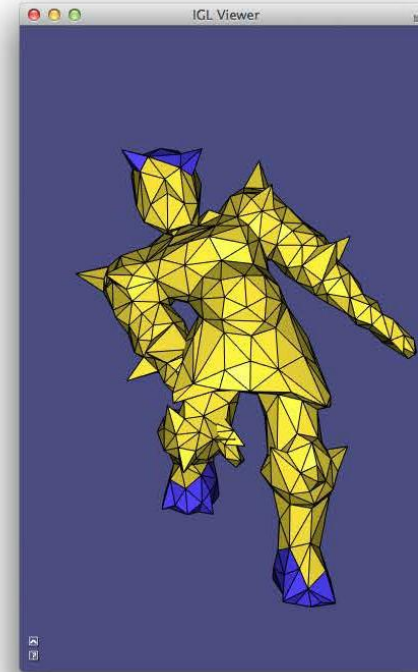
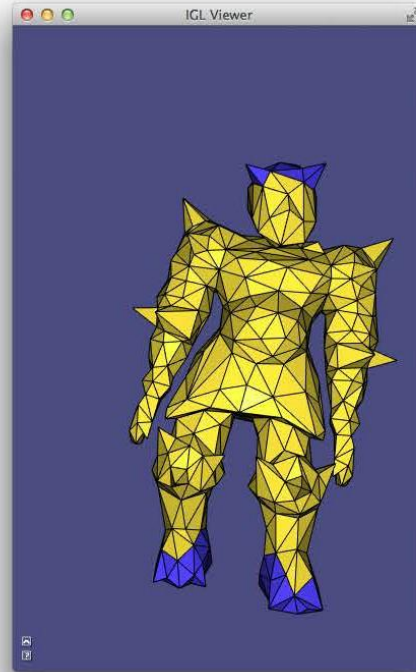
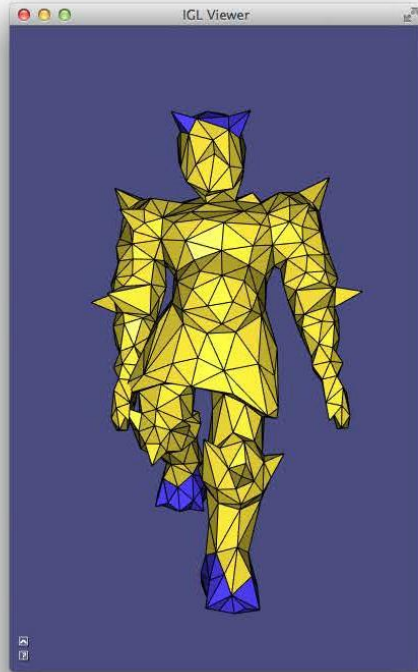
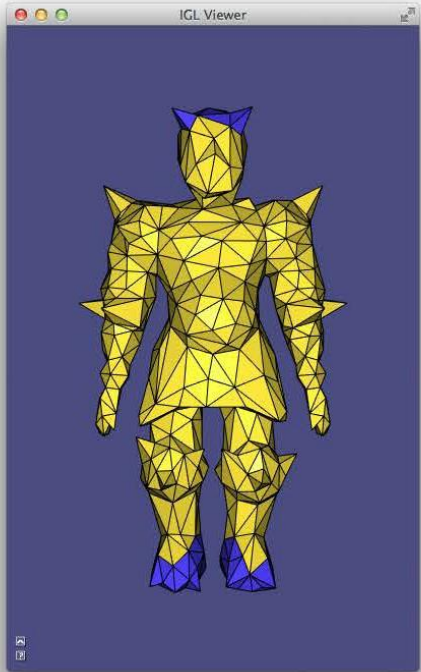


Behavior rules



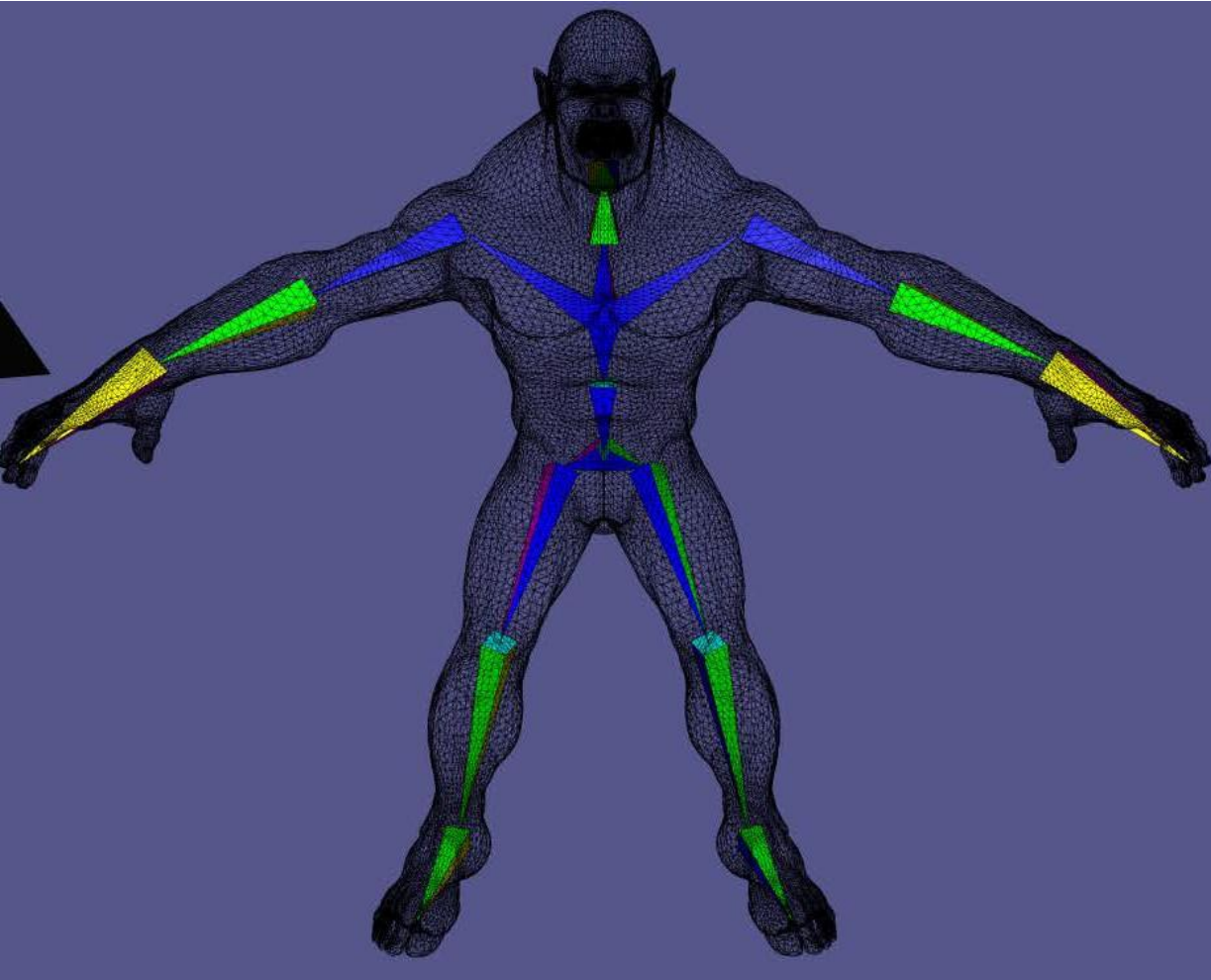
# How do articulated characters move?

## What to keyframe/simulate/animate?



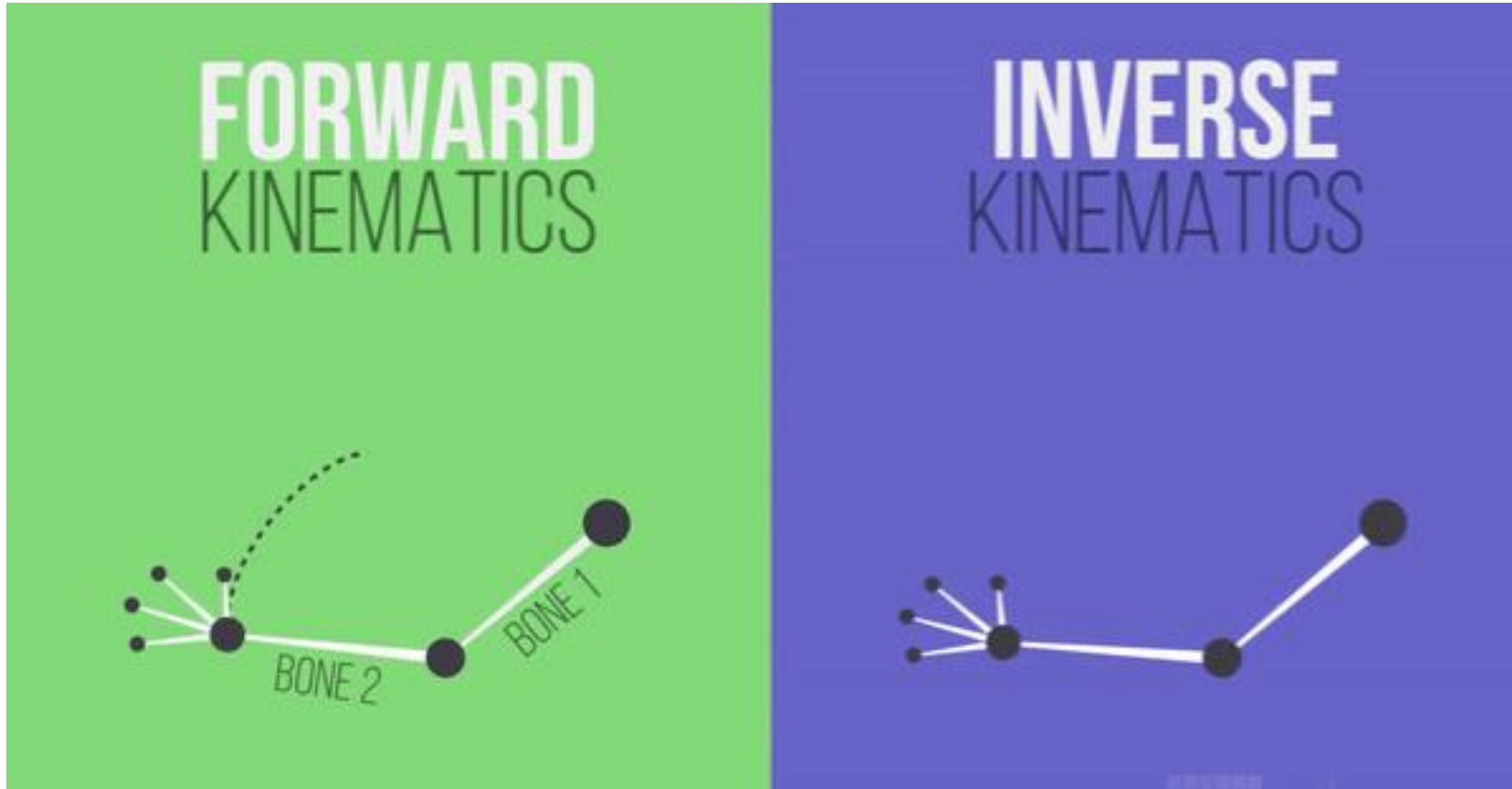
Per-Vertex?

# Skeleton: hierarchy of bones/joints

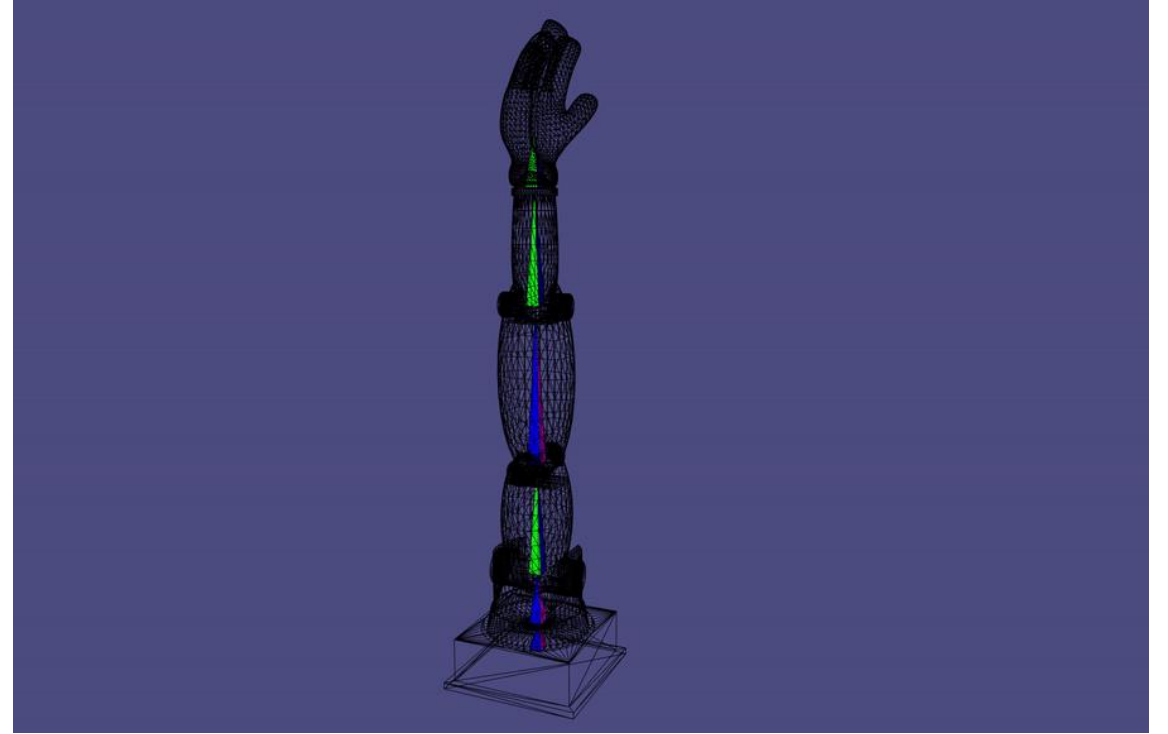


# Posing a skeleton:

## Forward Kinematics vs. Inverse Kinematics

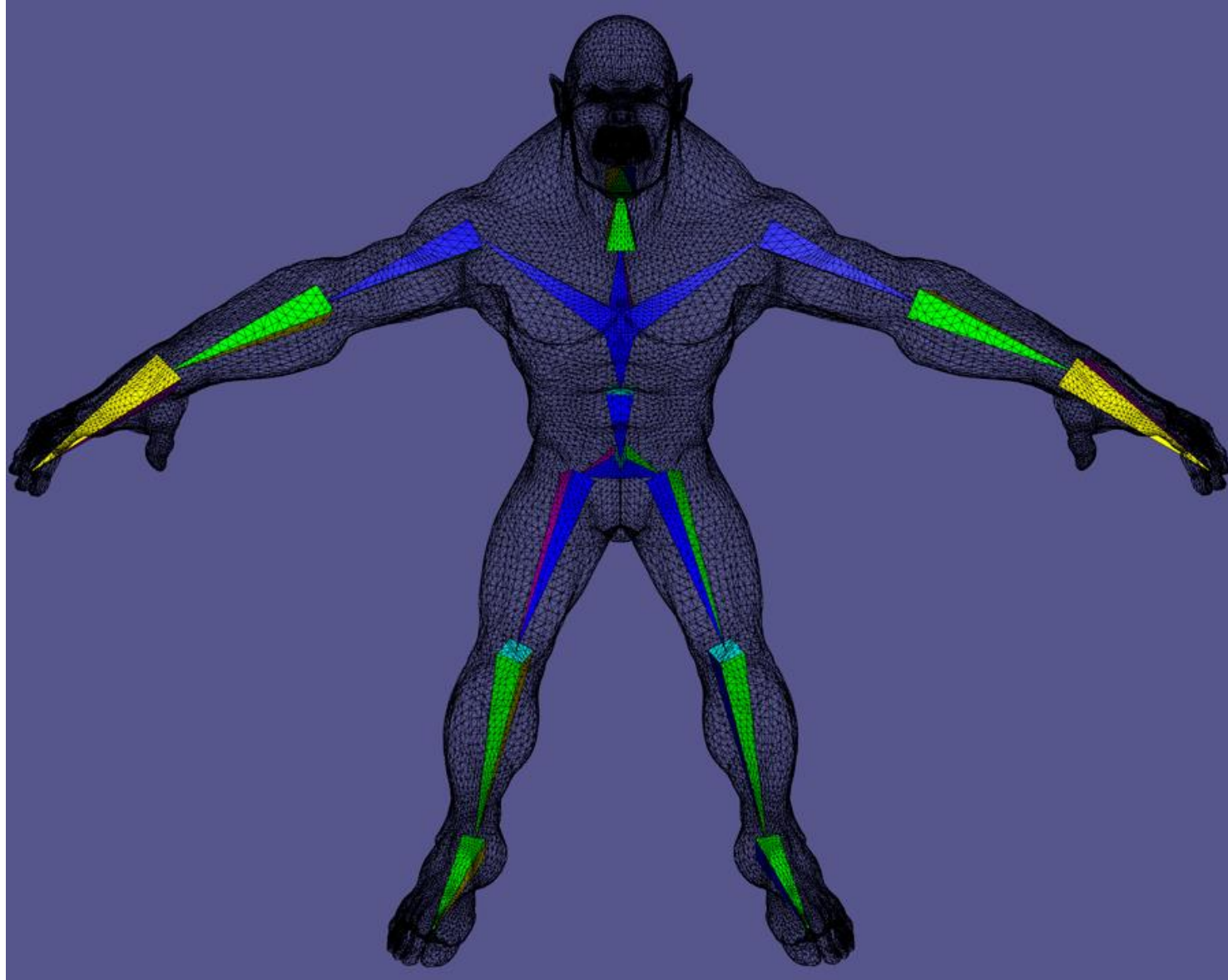


# Skeletons





# Deforming the object: Skinning



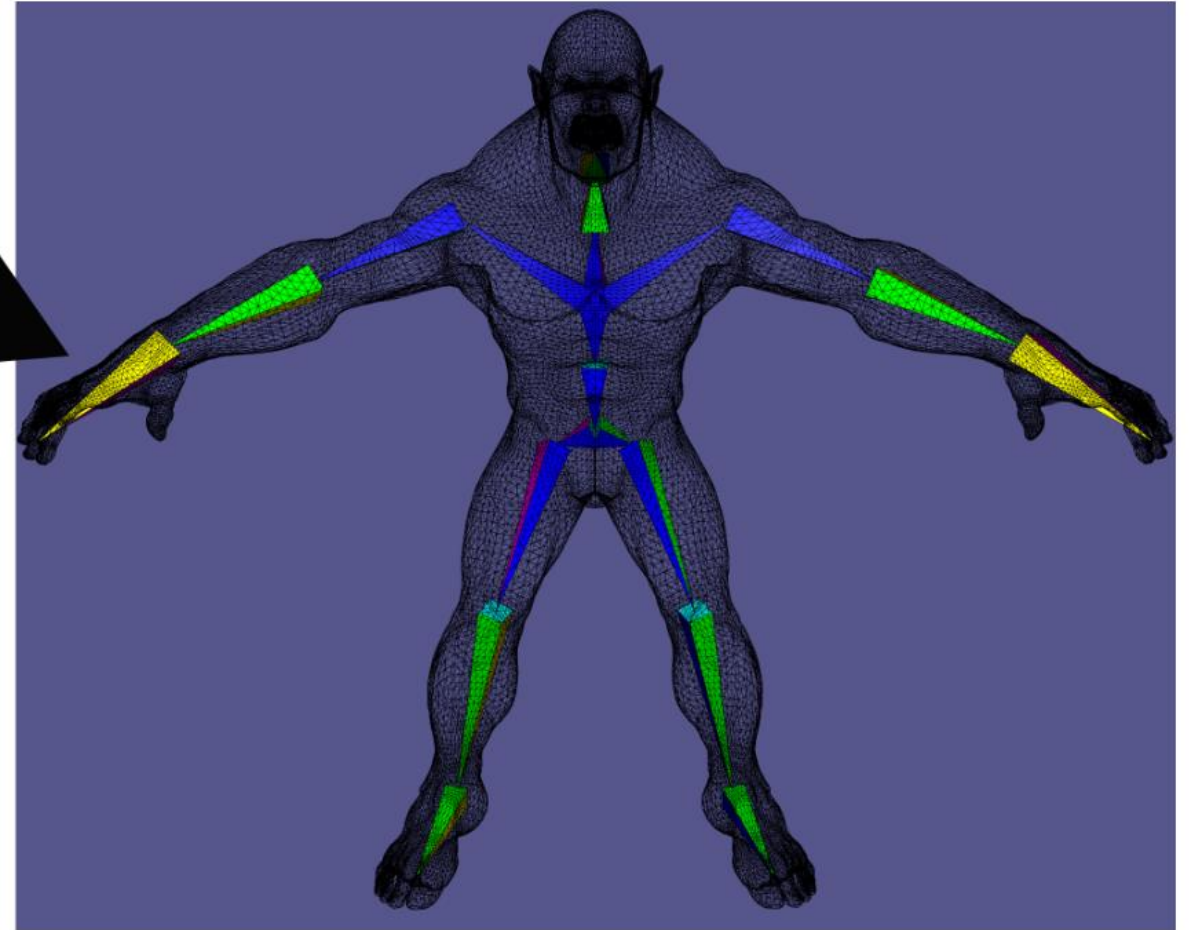
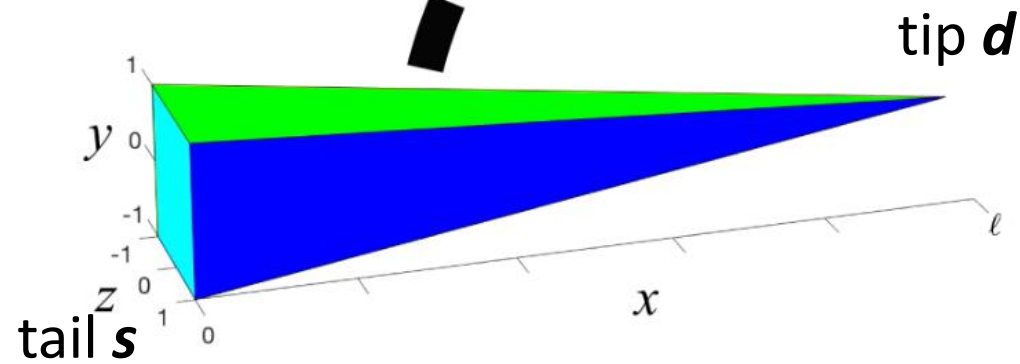
# Skeletons: Rest Bone

$$t=s=d_p$$

$$d=R[l \ 0 \ 0]^T + d_p$$

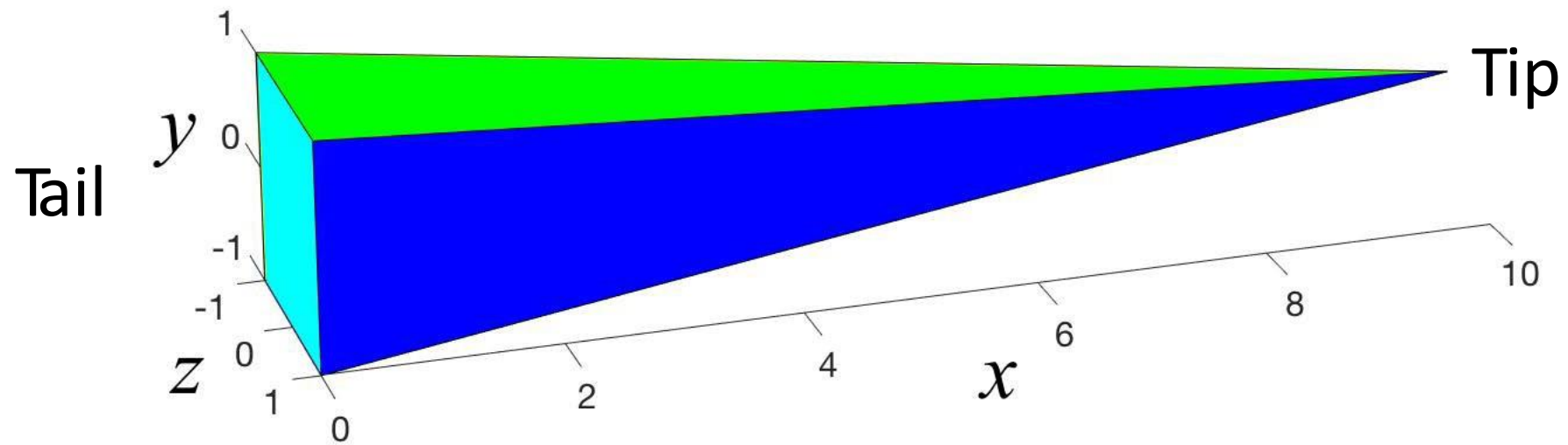
$$\hat{\mathbf{T}} = (\hat{\mathbf{R}} \quad \hat{\mathbf{t}}) \in \mathbb{R}^{3 \times 4}$$

Bone of length  $\ell$ :

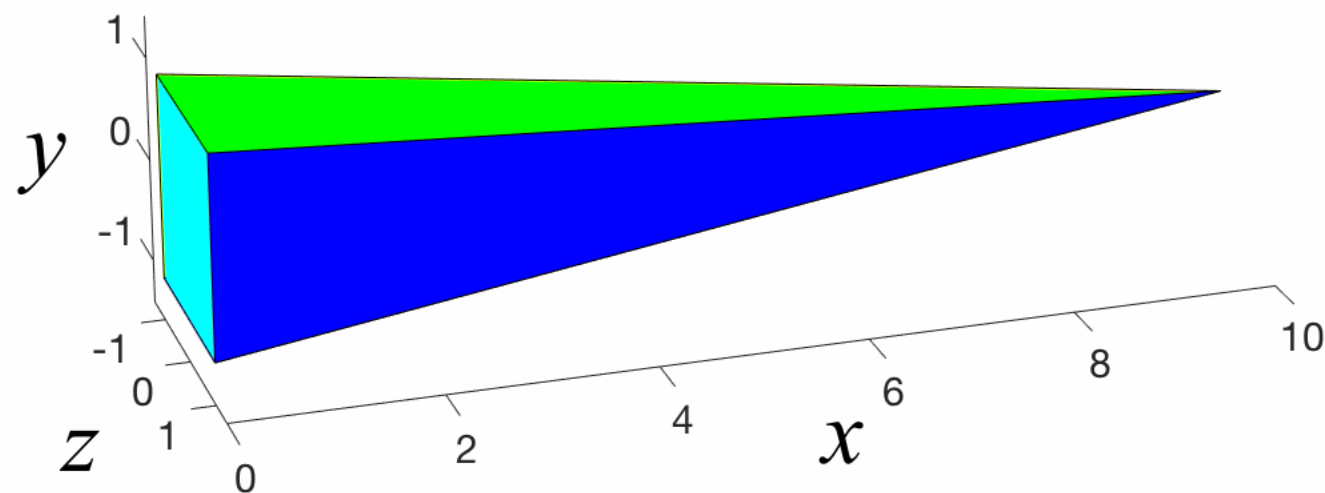


# Skeletons: Specifying Rotations

Bone of length  $\ell = 10$

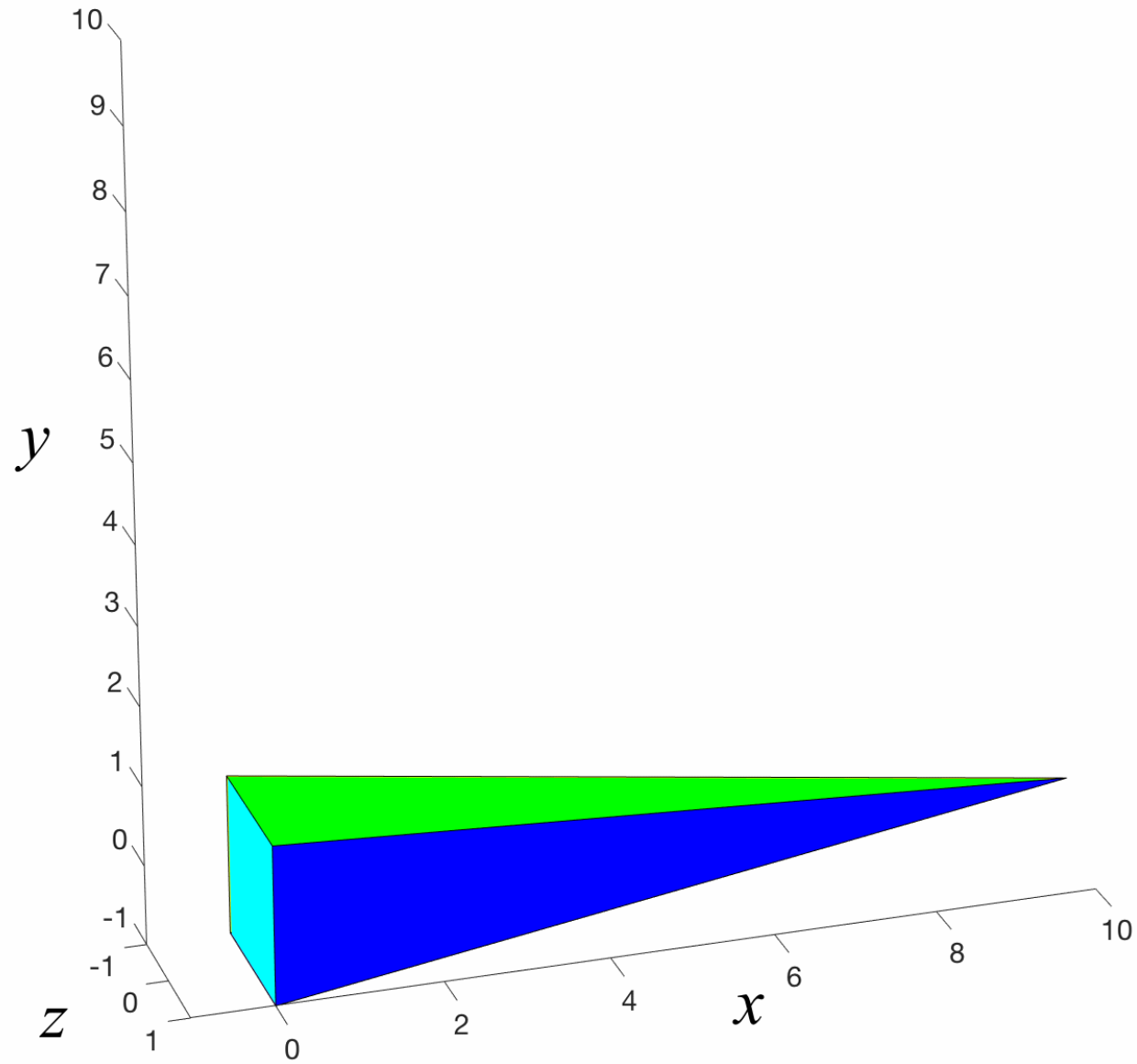


Twisting around  $x$  axis:  $\theta_1 = 0^\circ$

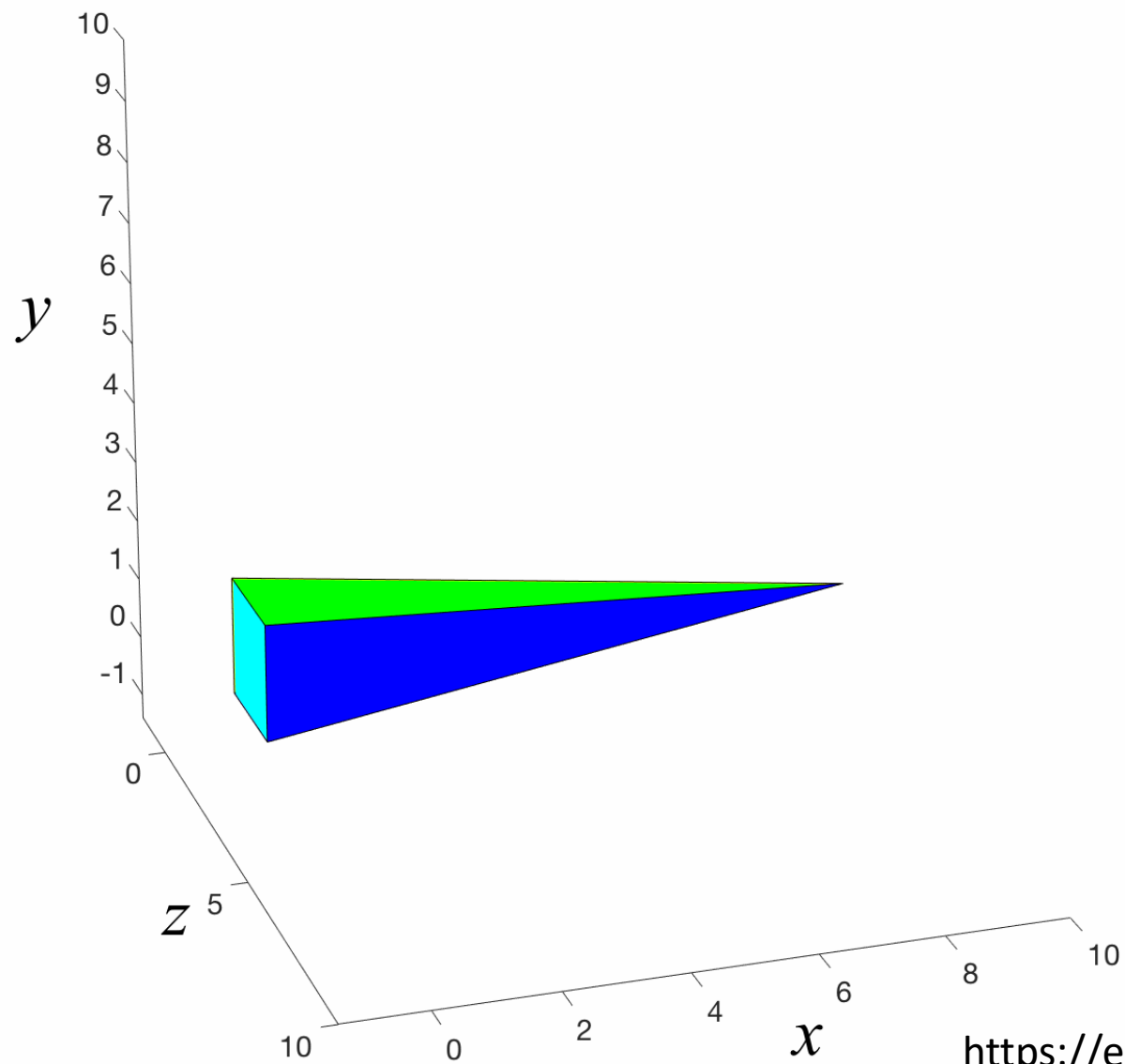




Bending around  $z$  axis:  $\theta_2 = 0^\circ$



Twist-bend-twist:  $(\theta_1, \theta_2, \theta_3) = (0^\circ, 0^\circ, 0^\circ)$

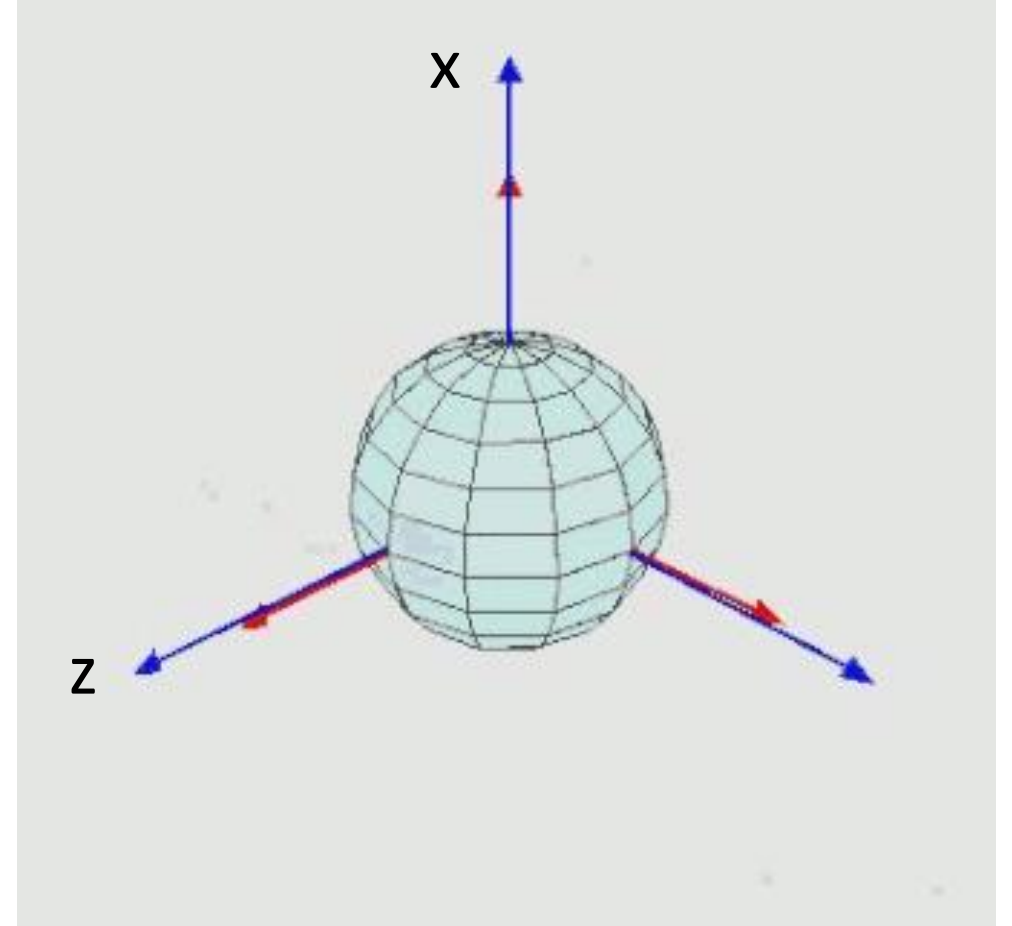


[https://en.wikipedia.org/wiki/Euler\\_angles](https://en.wikipedia.org/wiki/Euler_angles)

<https://mathworld.wolfram.com/EulerAngles.html>

# Euler Angle Rotations

$$R_x(\theta_3) * R_z(\theta_2) * R_x(\theta_1)$$



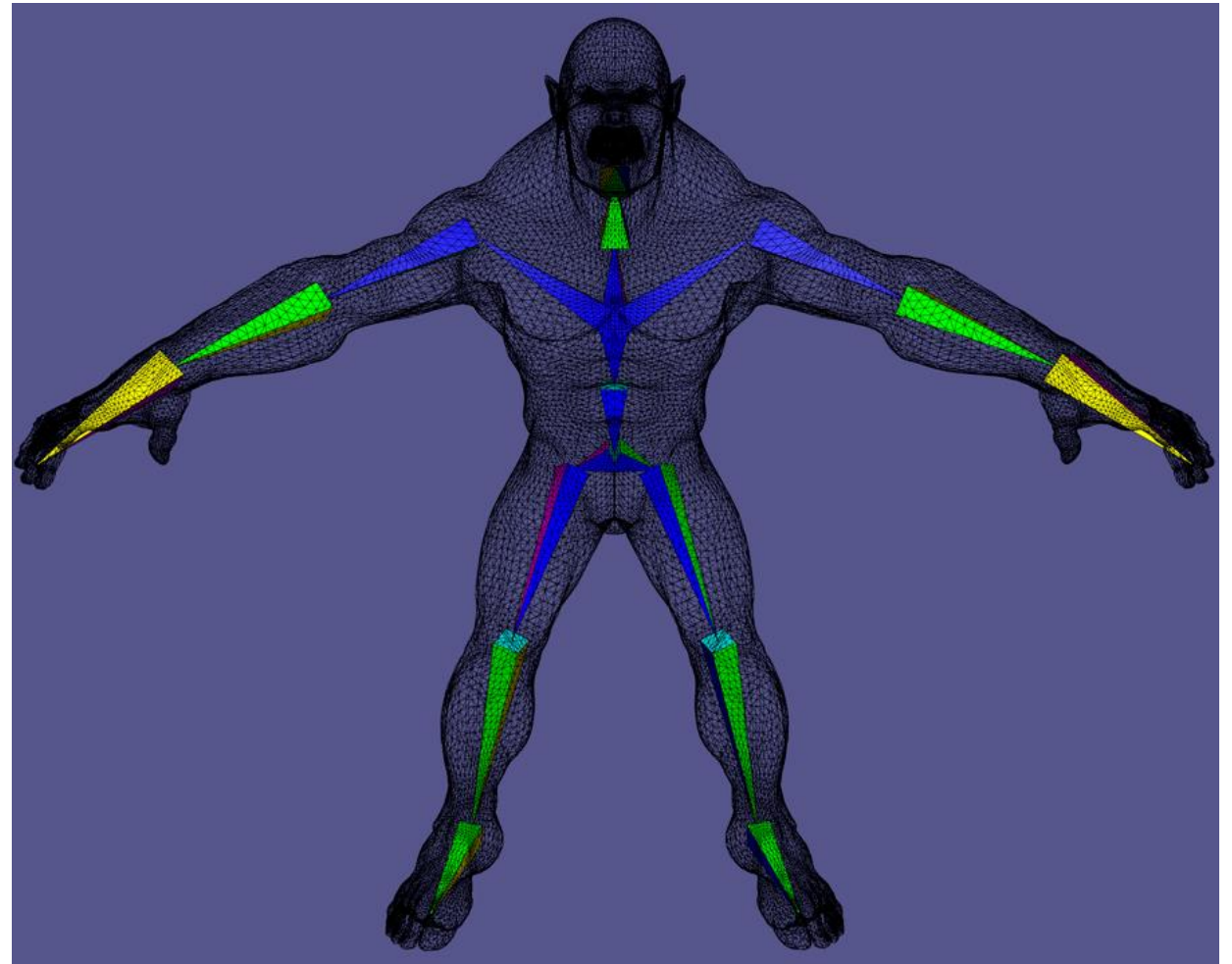
[https://en.wikipedia.org/wiki/Euler\\_angles](https://en.wikipedia.org/wiki/Euler_angles)

<https://mathworld.wolfram.com/EulerAngles.html>

# Skeletons: Forward Kinematics

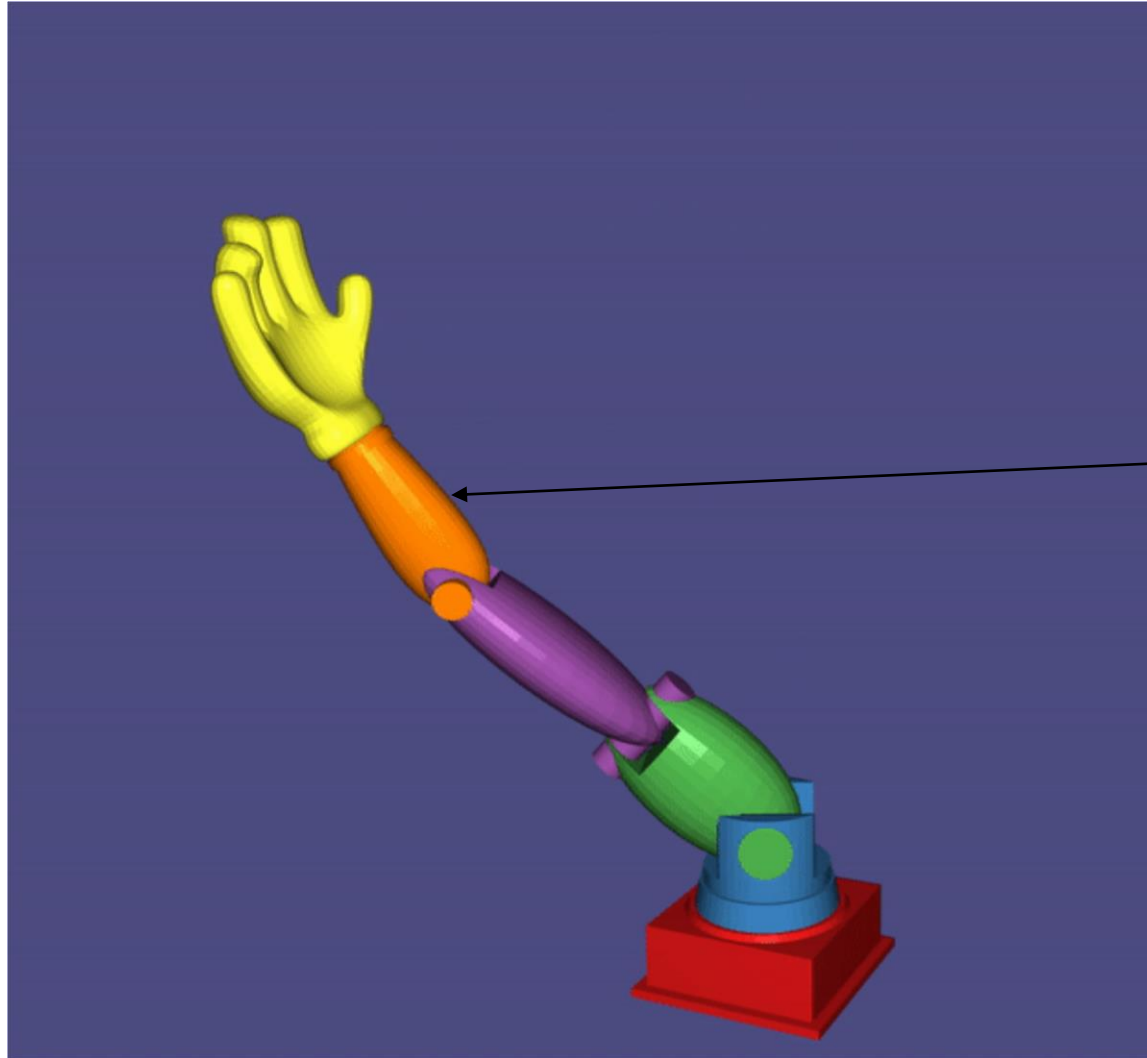
$$\mathbf{T}_i = \mathbf{T}_{p_i} \begin{pmatrix} \hat{\mathbf{T}}_i & \\ & 0 \ 0 \ 0 \ 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{R}}_i & 0 \\ & 0 \ 0 \ 0 \ 1 \end{pmatrix} \begin{pmatrix} \hat{\mathbf{T}}_i & \\ & 0 \ 0 \ 0 \ 1 \end{pmatrix}^{-1}$$

$$\mathbf{T}_i = \mathbf{T}_{p_i} \hat{\mathbf{T}}_i \begin{pmatrix} & 0 \\ \mathbf{R}_x(\theta_{i3}) & 0 \\ & 0 \\ 0 \ 0 \ 0 & 1 \end{pmatrix} \begin{pmatrix} & 0 \\ \mathbf{R}_z(\theta_{i2}) & 0 \\ & 0 \\ 0 \ 0 \ 0 & 1 \end{pmatrix} \begin{pmatrix} & 0 \\ \mathbf{R}_x(\theta_{i1}) & 0 \\ & 0 \\ 0 \ 0 \ 0 & 1 \end{pmatrix} \hat{\mathbf{T}}_i^{-1}$$



# Rigid “Skinning”

Idea: Attach each vertex to a single bone



$$\mathbf{v}_j = \mathbf{T}_i \hat{\mathbf{v}}_j$$

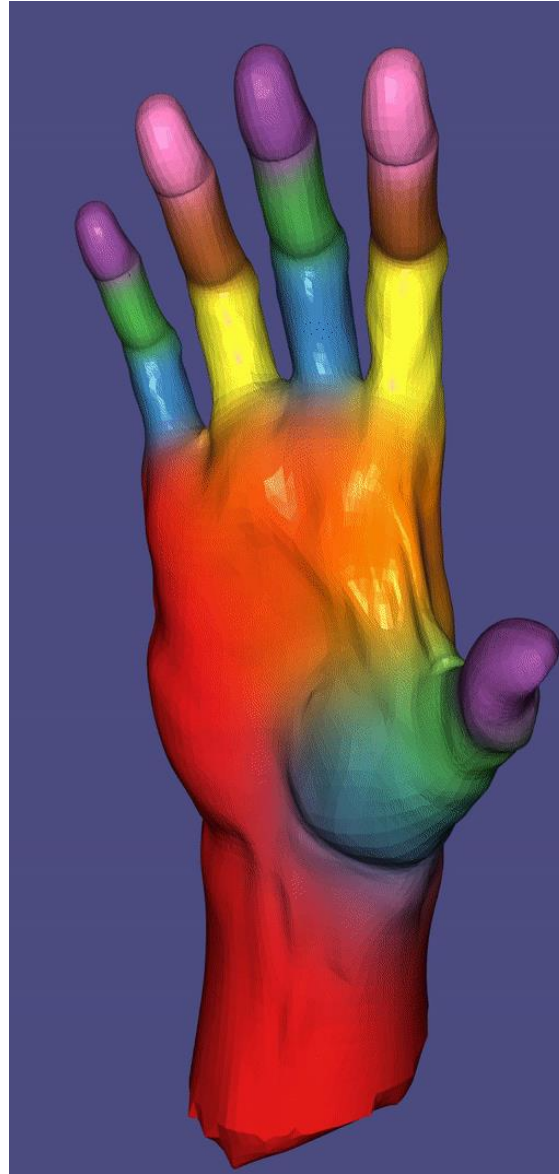
# Deformable Skinning

Rigid Skinning is fine for mechanical things, but for smoother deformations we need to try something else

Rather than attach each vertex to a single bone, we attach each vertex to multiple bones and ***blend*** their transformations

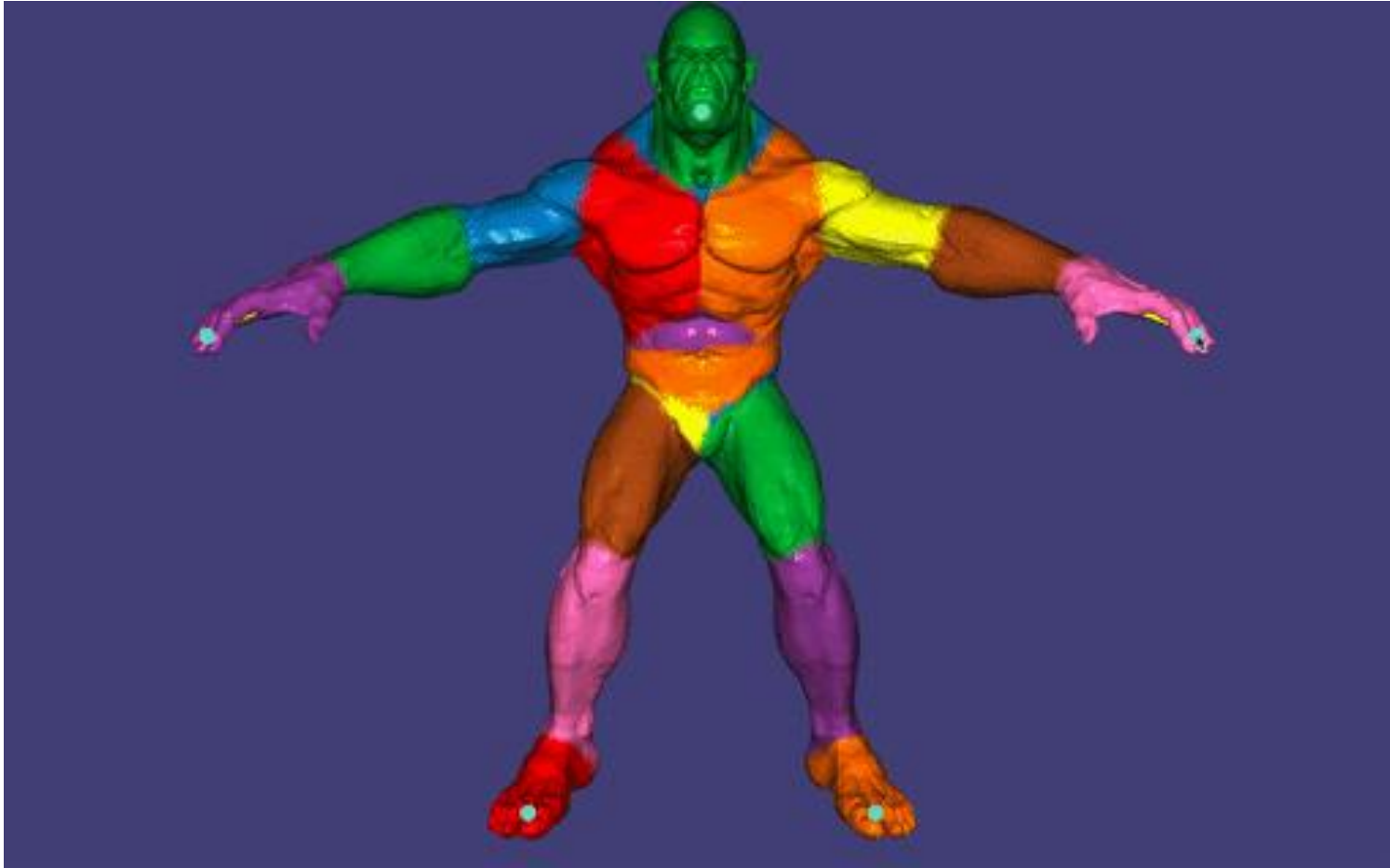
If this blending is linear in the transformations, we call it linear blend skinning.

# Linear Blend Skinning



$$\mathbf{v}_j = \sum_{i=1}^{\text{\#bones}} w_{ij} \mathbf{T}_i \hat{\mathbf{v}}_j$$

# Rigid vs Linear Blend Skinning

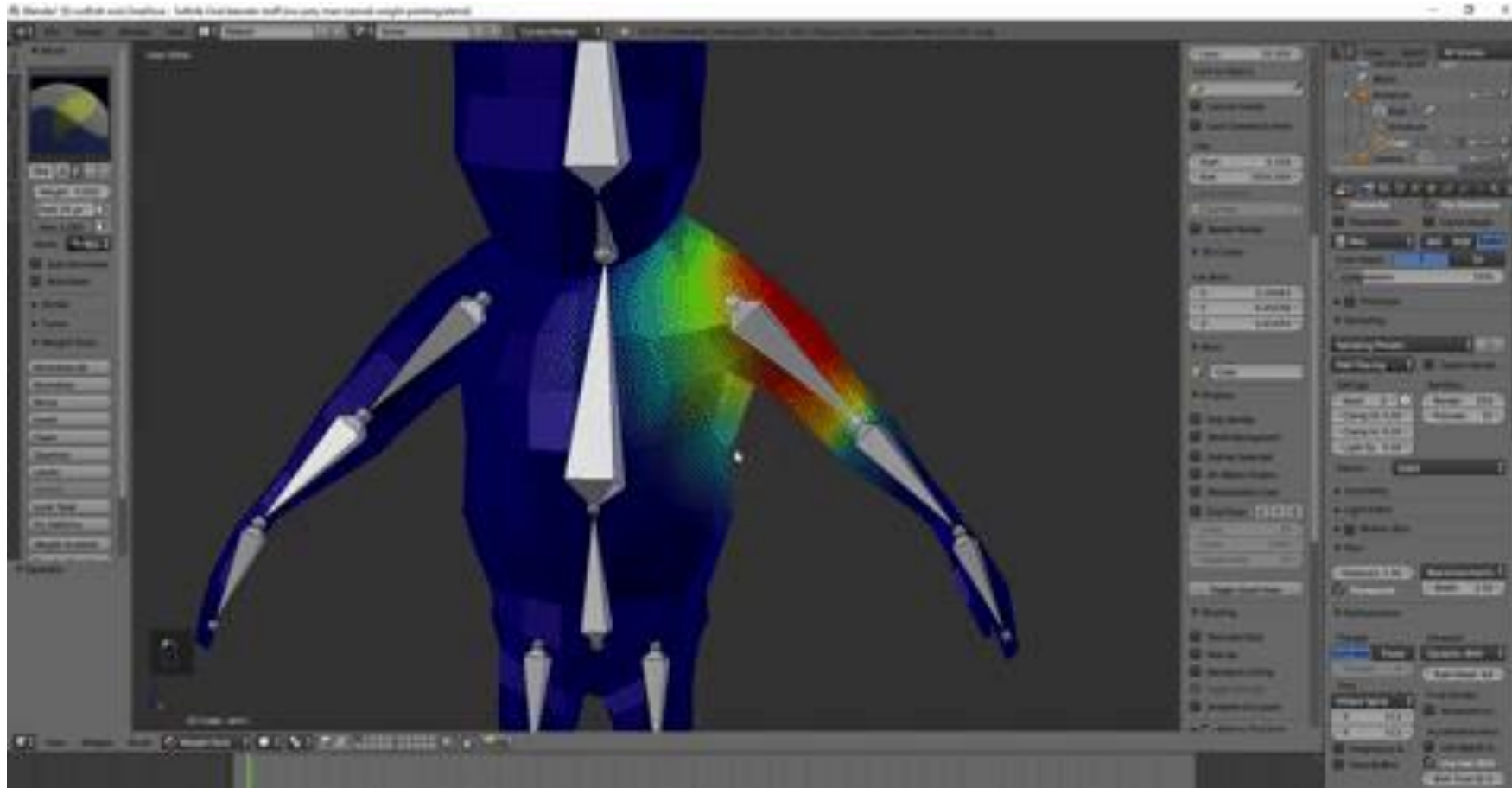




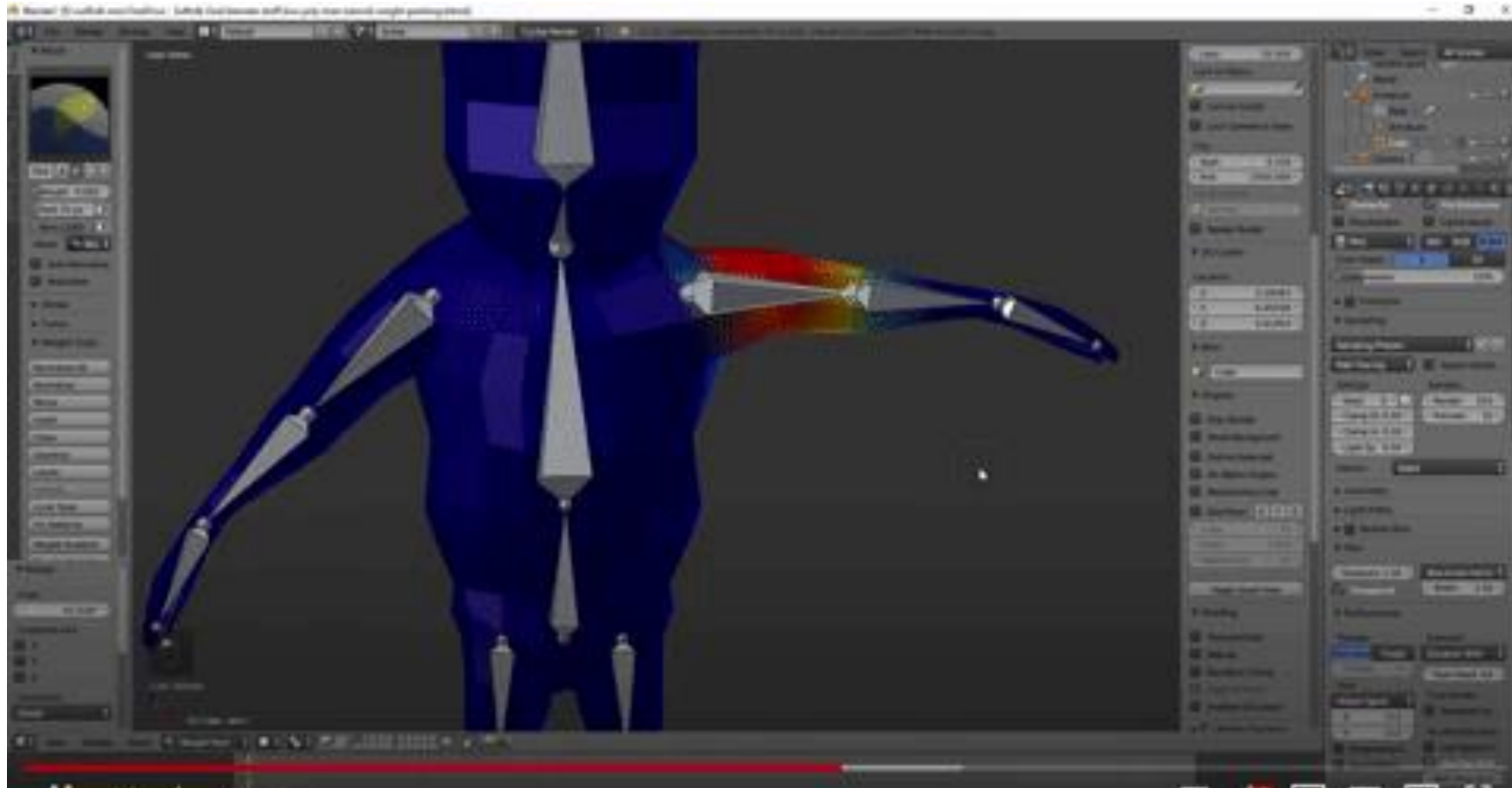
# Weight Painting



# Weight Painting



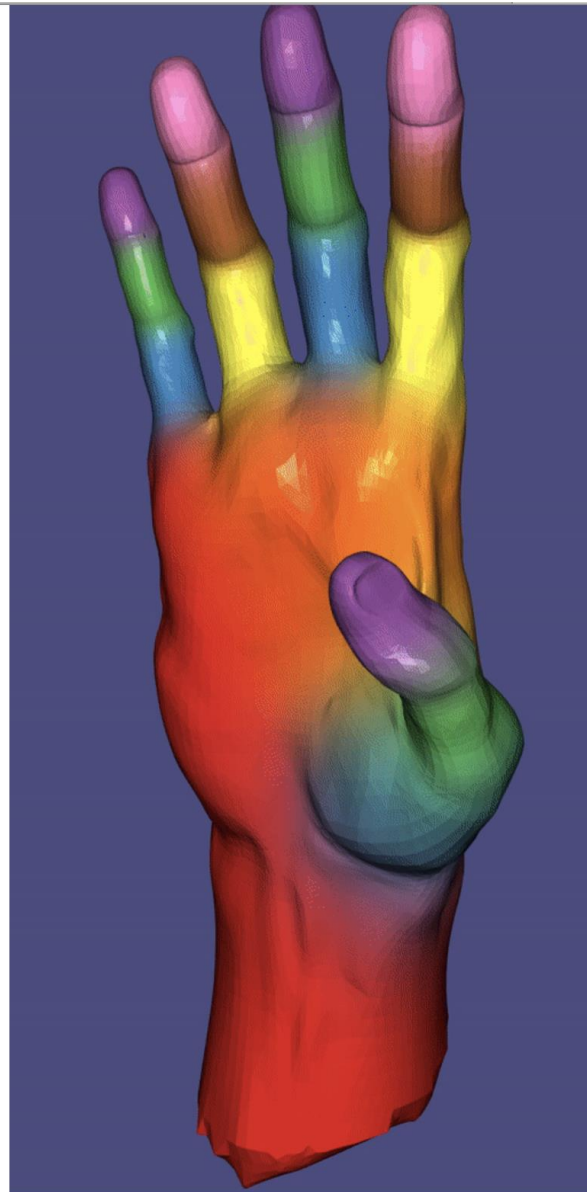
# Weight Painting



# Specifying Keyframes



Time = 0



Time = 10s

Poses are generated by specifying rotations of bones

Each pose can be represented as

$$\theta = \begin{pmatrix} \theta_{11} \\ \theta_{11} \\ \theta_{11} \\ \vdots \\ \theta_{n1} \\ \theta_{n2} \\ \theta_{n3} \end{pmatrix}$$

# Specifying Keyframes



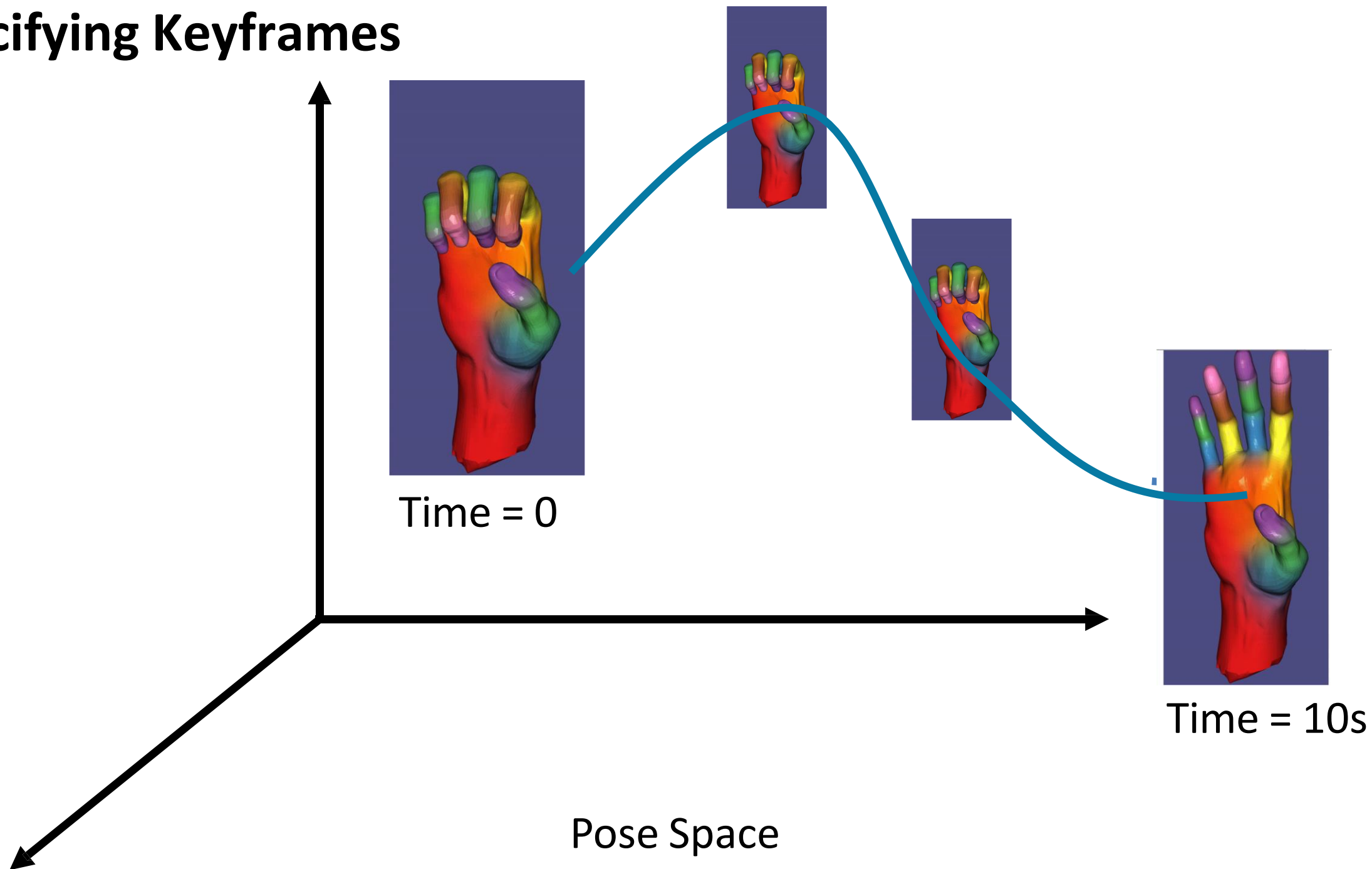
Time = 0

????????????



Time = 10s

# Specifying Keyframes





# Specifying Keyframes



[https://en.wikipedia.org/wiki/Twelve basic principles of animation#Slow in and slow out](https://en.wikipedia.org/wiki/Twelve_basic_principles_of_animation#Slow_in_and_slow_out)

<https://www.youtube.com/watch?v=fQBFsTqbKhY>

# Interpolating Keyframes

$$\theta = \mathbf{c}(t)$$

is a curve in the pose space

How could we construct such a curve from keyframes ?

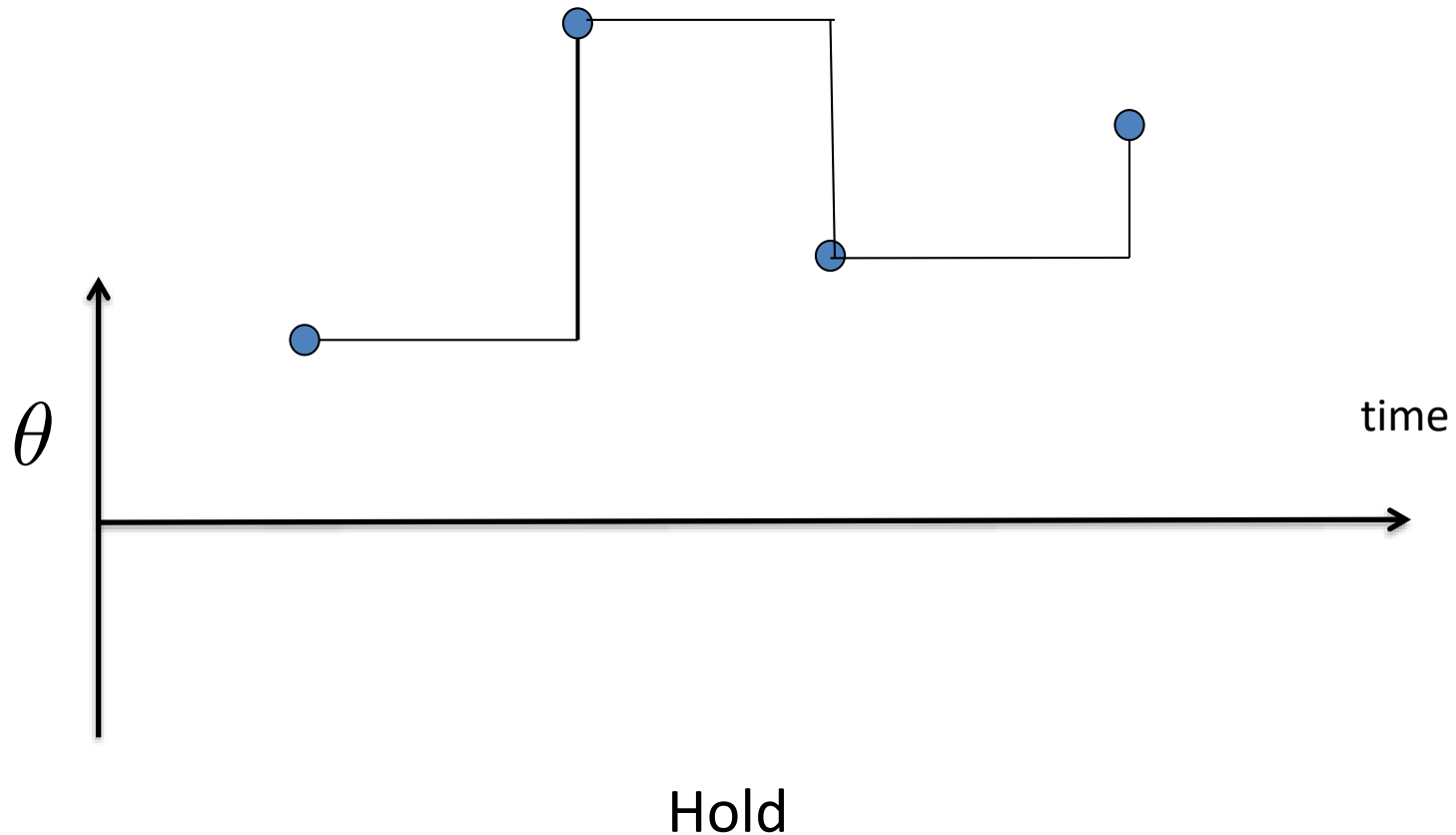


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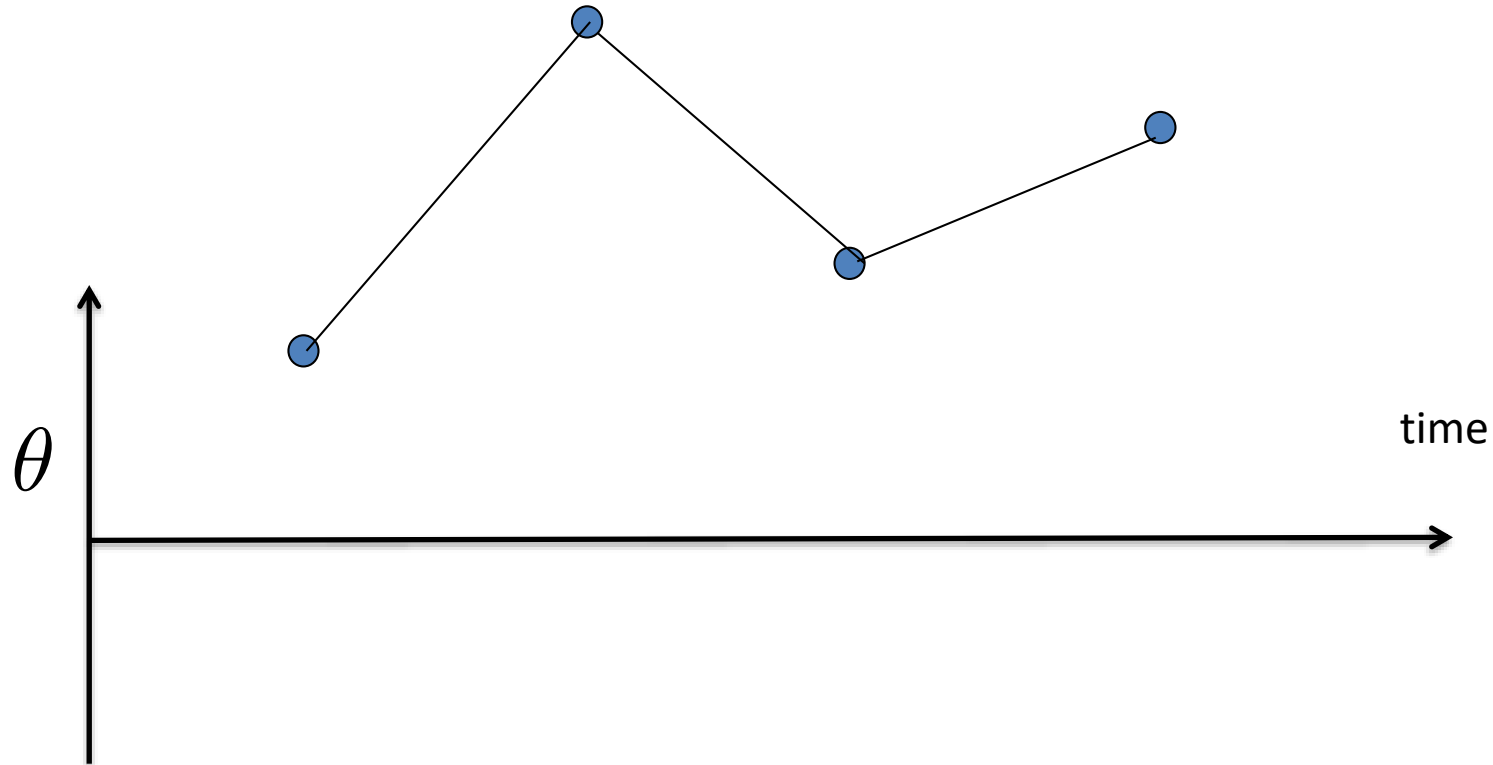


# Interpolating Keyframes

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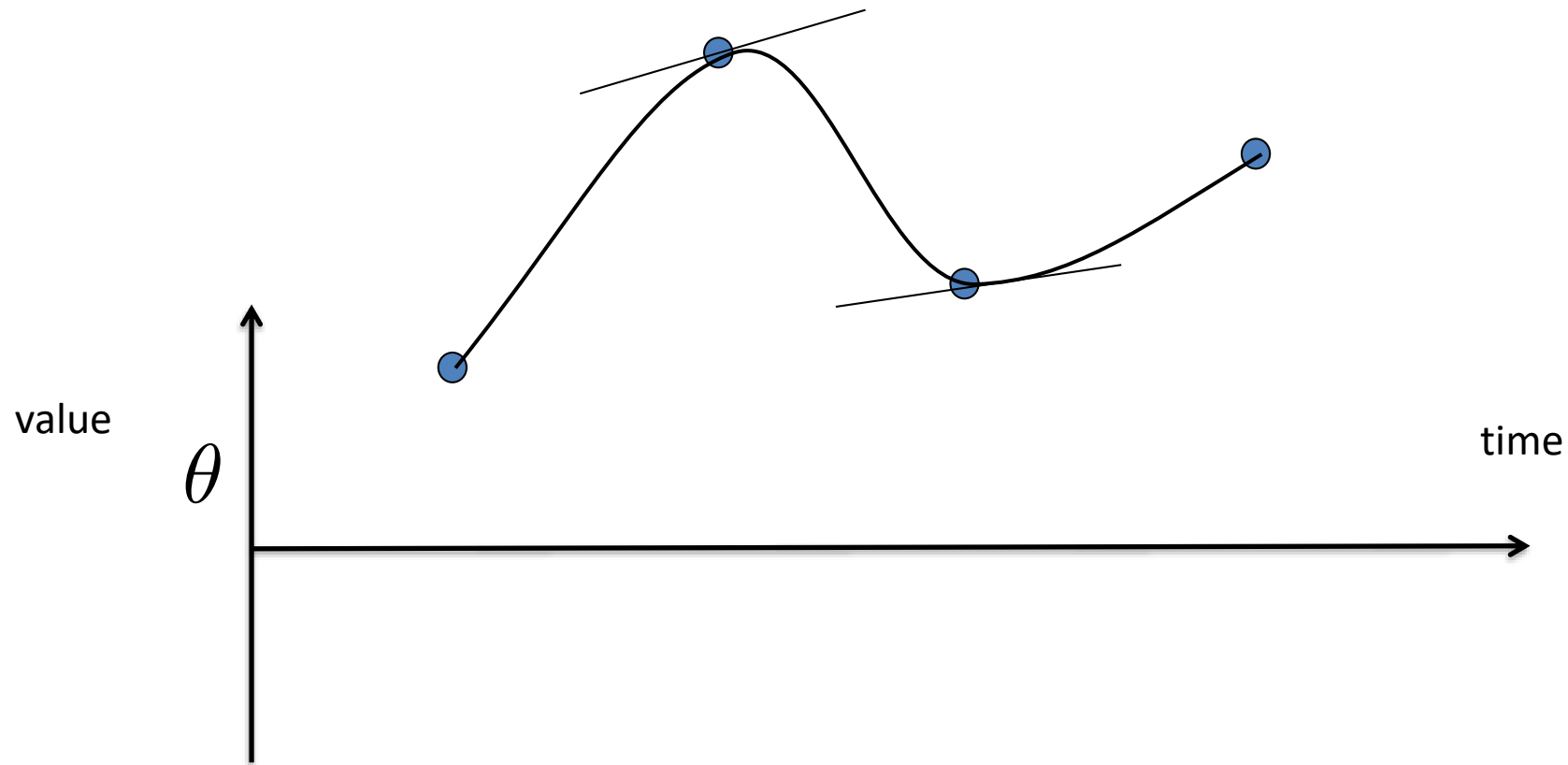
Linear

# Interpolating Keyframes

$$\theta = \mathbf{c}(t)$$

is a curve in the pose space

How could we construct such a curve from keyframes ?



Spline

# Catmull-Rom Spline

A **cubic** curve created by specifying the end points and the tangents of the curve.

$$\mathbf{c}(t) = at^3 + bt^2 + ct + d$$
$$\mathbf{c}'(t) = 3at^2 + 2bt + c$$



# Catmull-Rom Spline

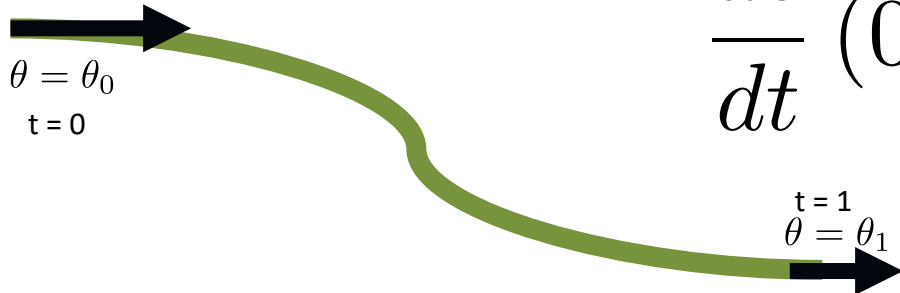
A **cubic** curve created by specifying the end points and the tangents of the curve.

$$c(0) = d$$

$$c(1) = a + b + c + d$$

$$\frac{dc}{dt}(1) = 3a + 2b + 1c$$

$$\frac{dc}{dt}(0) = 1c$$



## Catmull-Rom Spline

$$\begin{pmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \begin{pmatrix} \mathbf{a}^T \\ \mathbf{b}^T \\ \mathbf{c}^T \\ \mathbf{d}^T \end{pmatrix} = \begin{pmatrix} \theta_0^T \\ \theta_1^T \\ \mathbf{m}_0^T \\ \mathbf{m}_1^T \end{pmatrix}$$

# Catmull-Rom Spline

After solving and rearranging we end up with

$$\mathbf{c}(t) = (2t^3 - 3t^2 + 1)\theta_0 + (t^3 - 2t^2 + t)\mathbf{m}_0 + (-2t^3 + 3t^2)\theta_1 + (t^3 - t^2)\mathbf{m}_1$$

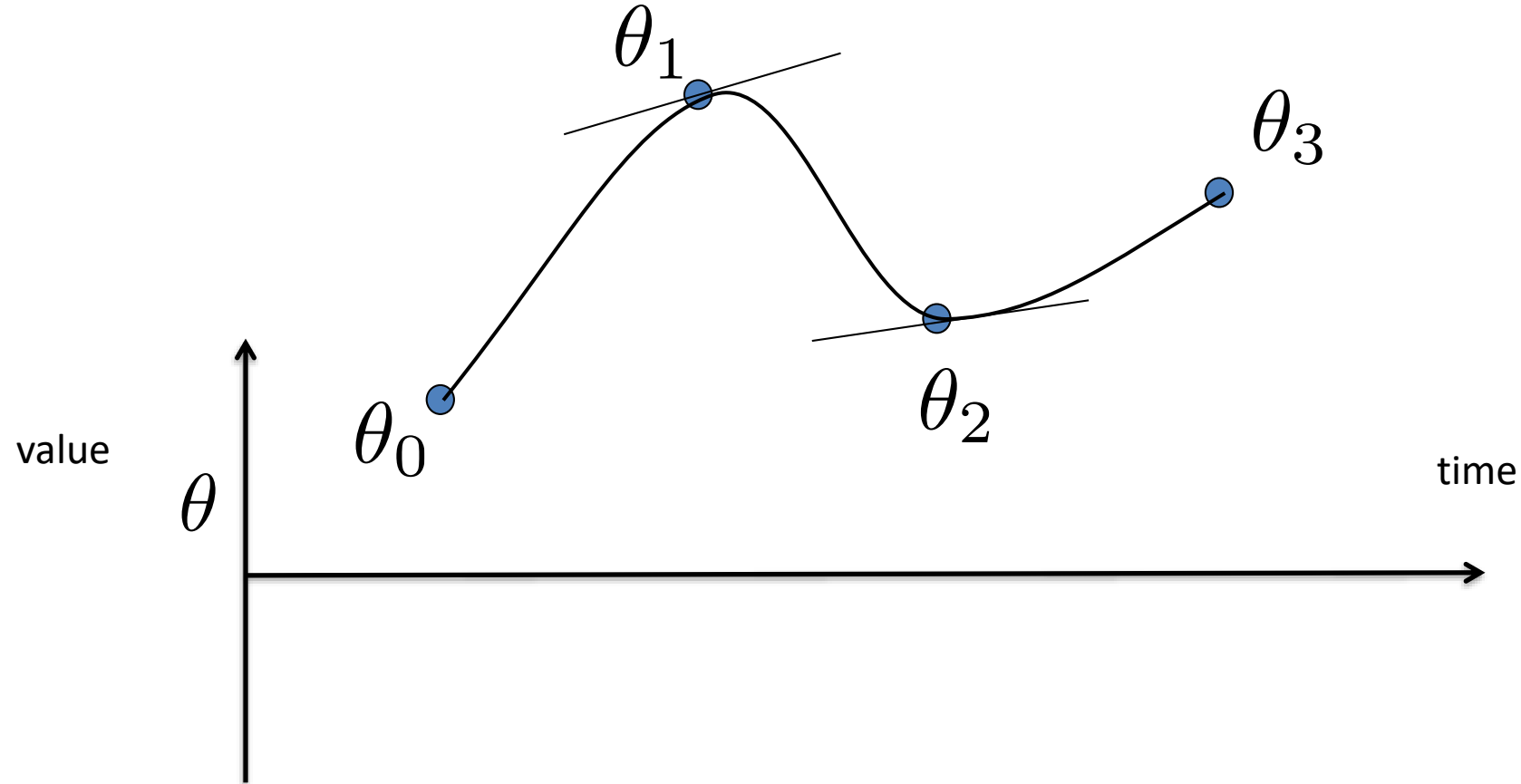
Remember this is for  $t = 0$  to  $t = 1$ .

For arbitrary intervals substitute

$$t = (t' - t_0)/(t_1 - t_0)$$

# Catmull-Rom Spline

Catmull-Rom chooses the tangents using “Finite Differences”



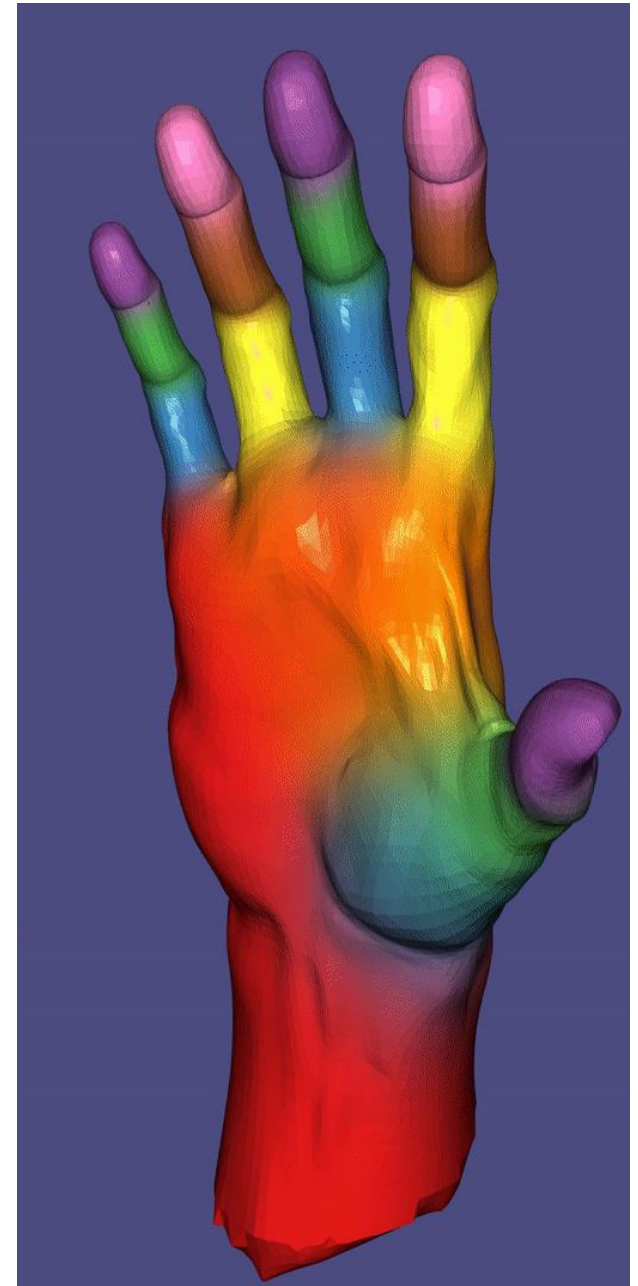
$$\mathbf{m}_k = \frac{\theta_{k+1} - \theta_k}{t_{k+1} - t_k}$$



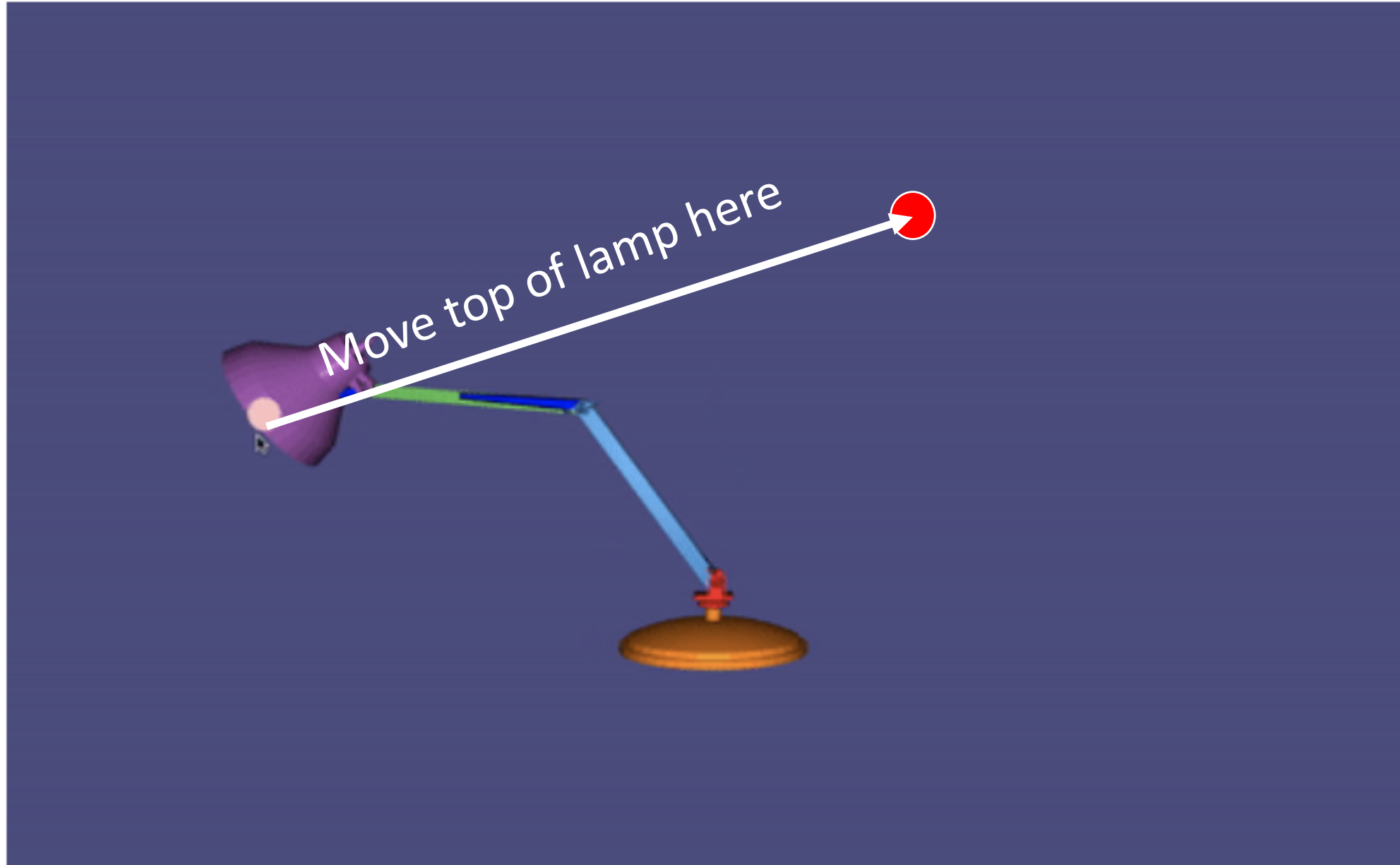
# Catmull-Rom Animation

For each time,  $t$ , create a new  $\theta(t)$  using your spline.

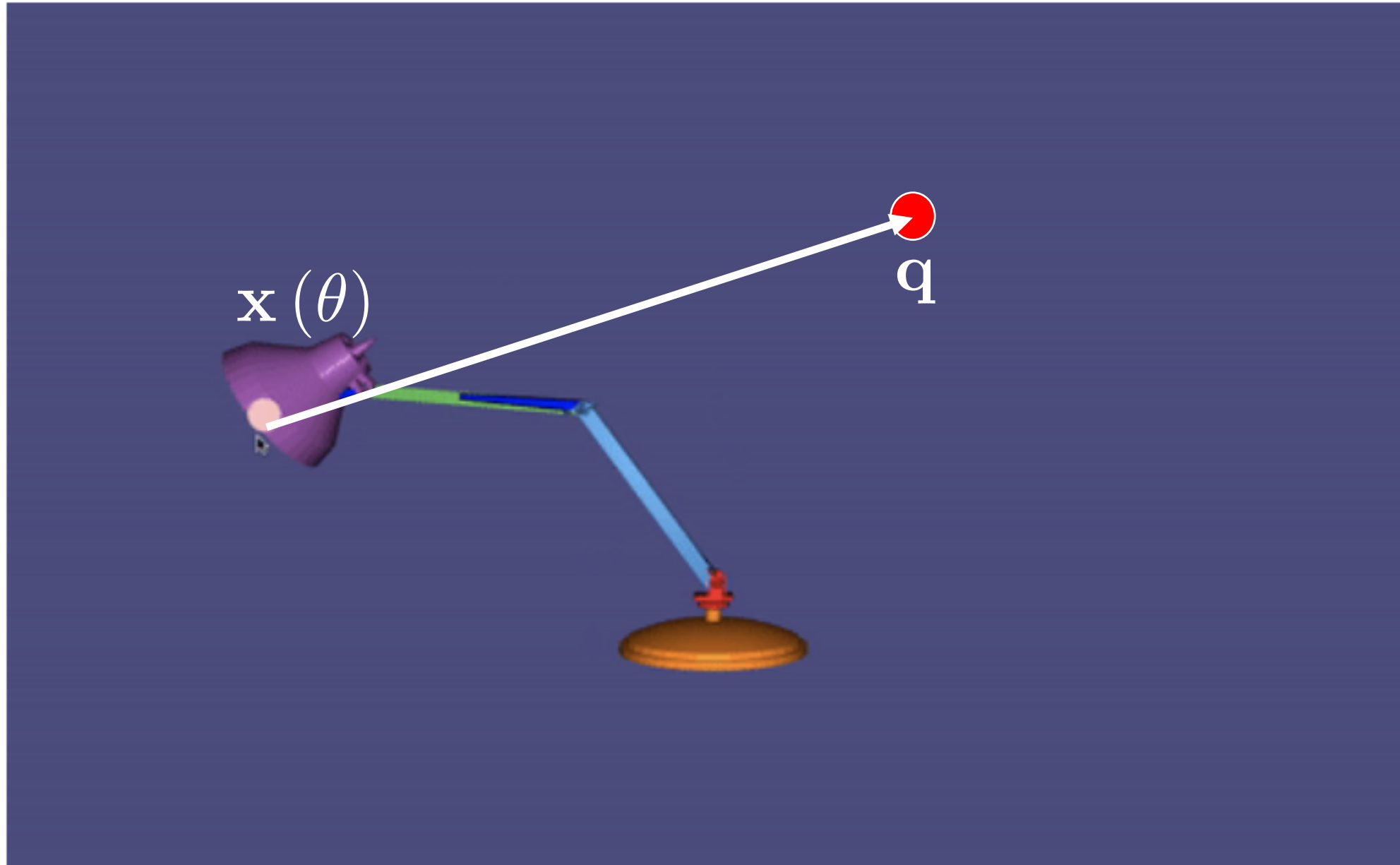
Re-pose your bones using  $\theta$ , deform skin and draw...



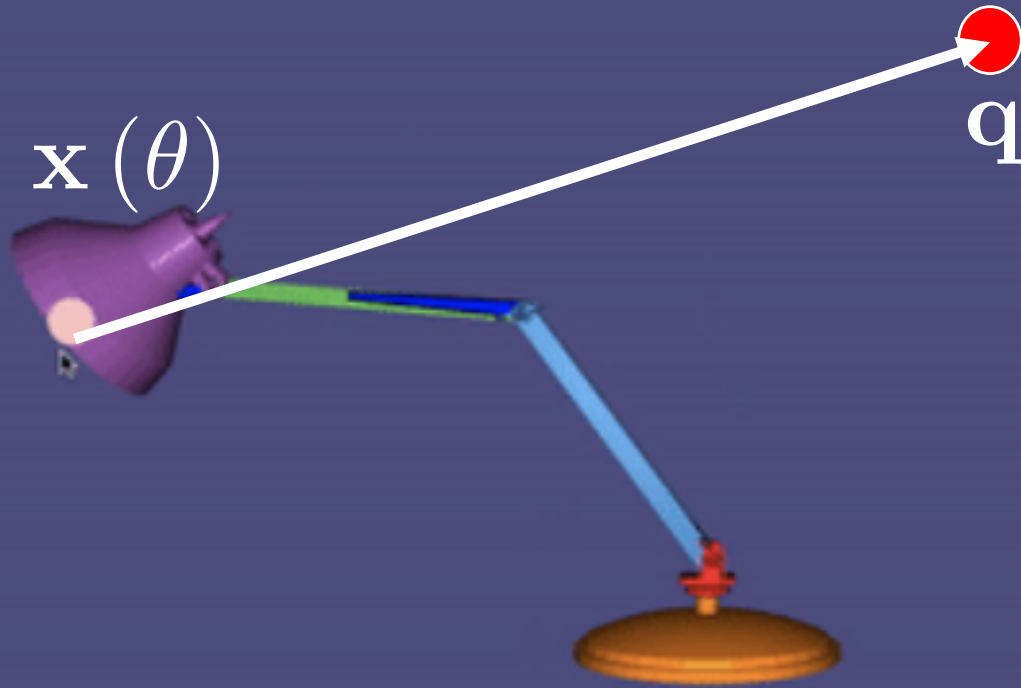
# Inverse Kinematics



# Inverse Kinematics



# Inverse Kinematics



$$\theta^* = \arg \min \|\mathbf{x}(\theta) - \mathbf{q}\|_2^2$$

# Skeletons: Inverse Kinematics

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What is the pose (set of joint angles  $a$ ) that lets us reach a given point (end-effector position).

$$\mathbf{a} = \begin{pmatrix} \theta_{11} \\ \theta_{12} \\ \theta_{13} \\ \theta_{21} \\ \theta_{22} \\ \theta_{23} \\ \vdots \\ \theta_{m1} \\ \theta_{m2} \\ \theta_{m3} \end{pmatrix}$$



What does it mean to reach (get as close as possible) to a point?

# Skeletons: Inverse Kinematics

Closeness energy can be measured the squared distance between the pose tip  $\mathbf{x}_b$  of some bone  $b$  and a desired goal location  $\mathbf{q} \in \mathbb{R}^3$ .

$$E(\mathbf{x}_b) = \|\mathbf{x}_b - \mathbf{q}\|^2.$$

Given pose vector  $\mathbf{a}$ , the bone tip  $\mathbf{x}_b$  is:

$$\mathbf{x}_b(\mathbf{a}) = \mathbf{T}_b \hat{\mathbf{d}}_b$$

Now given any number of end-effectors  $b_1 \dots b_k$ :

$$\min_{\mathbf{a}} \underbrace{\sum_{i=1}^k \|\mathbf{x}_{b_i}(\mathbf{a}) - \hat{\mathbf{x}}_{b_i}\|^2}_{E(\mathbf{x}_b(\mathbf{a}))}$$

And we impose some joint angle limits:  $\min_{\mathbf{a}^{\min} \leq \mathbf{a} \leq \mathbf{a}^{\max}} E(\mathbf{x}_b(\mathbf{a}))$

Minimizing this energy is a non-linear least squares problem.



# Inverse Kinematics: Energy

$$E(\mathbf{x}_b(\mathbf{a})) = \|\mathbf{x}_b(\mathbf{a}) - \mathbf{q}\|^2$$

the squared distance between the pose tip  $\mathbf{x}_b$  of some bone  $\mathbf{b}$ , and a desired goal location  $\mathbf{q}$ .

# Inverse Kinematics: Energy

$$E(\mathbf{x}_b(\mathbf{a})) = \|\mathbf{x}_b(\mathbf{a}) - \mathbf{q}\|^2$$

list of constrained end  
effectors

→  $b = \{b_1, b_2, \dots, b_k\}$

$$\min_{\mathbf{a}} \underbrace{\sum_{i=1}^k \|\mathbf{x}_{b_i}(\mathbf{a}) - \hat{\mathbf{x}}_{b_i}\|^2}_{E(\mathbf{x}_b(\mathbf{a}))}$$



# Inverse Kinematics: Energy

Over all choices of Euler angles  $\mathbf{a}$ , we want the angles that ensure all selected end effectors go to their prescribed locations, subject to angle constraints:

$$\min_{\mathbf{a}^{\min} \leq \mathbf{a} \leq \mathbf{a}^{\max}} E(\mathbf{x}_b(\mathbf{a}))$$

The diagram illustrates the energy minimization equation for inverse kinematics. It features the equation 
$$\min_{\mathbf{a}} \sum_{i=1}^k \|\mathbf{x}_{b_i}(\mathbf{a}) - \hat{\mathbf{x}}_{b_i}\|^2$$
 with several annotations and arrows. A red arrow points from the text "Sum over all constrained end effectors" to the summation symbol  $\sum_{i=1}^k$ . Another red arrow points from the text "Position + rotation of bone  $\mathbf{x}_{b_i}$  found using forward kinematics" to the term  $\mathbf{x}_{b_i}(\mathbf{a})$ . A third red arrow points from the text "Constrained end effector position" to the term  $\hat{\mathbf{x}}_{b_i}$ . A large curly brace is placed under the entire sum, with the label  $E(\mathbf{x}_b(\mathbf{a}))$  centered below it.

Position + rotation of bone  $\mathbf{x}_{b_i}$  found using forward kinematics

Constrained end effector position

$\min_{\mathbf{a}} \sum_{i=1}^k \|\mathbf{x}_{b_i}(\mathbf{a}) - \hat{\mathbf{x}}_{b_i}\|^2$

Sum over all constrained end effectors

$E(\mathbf{x}_b(\mathbf{a}))$

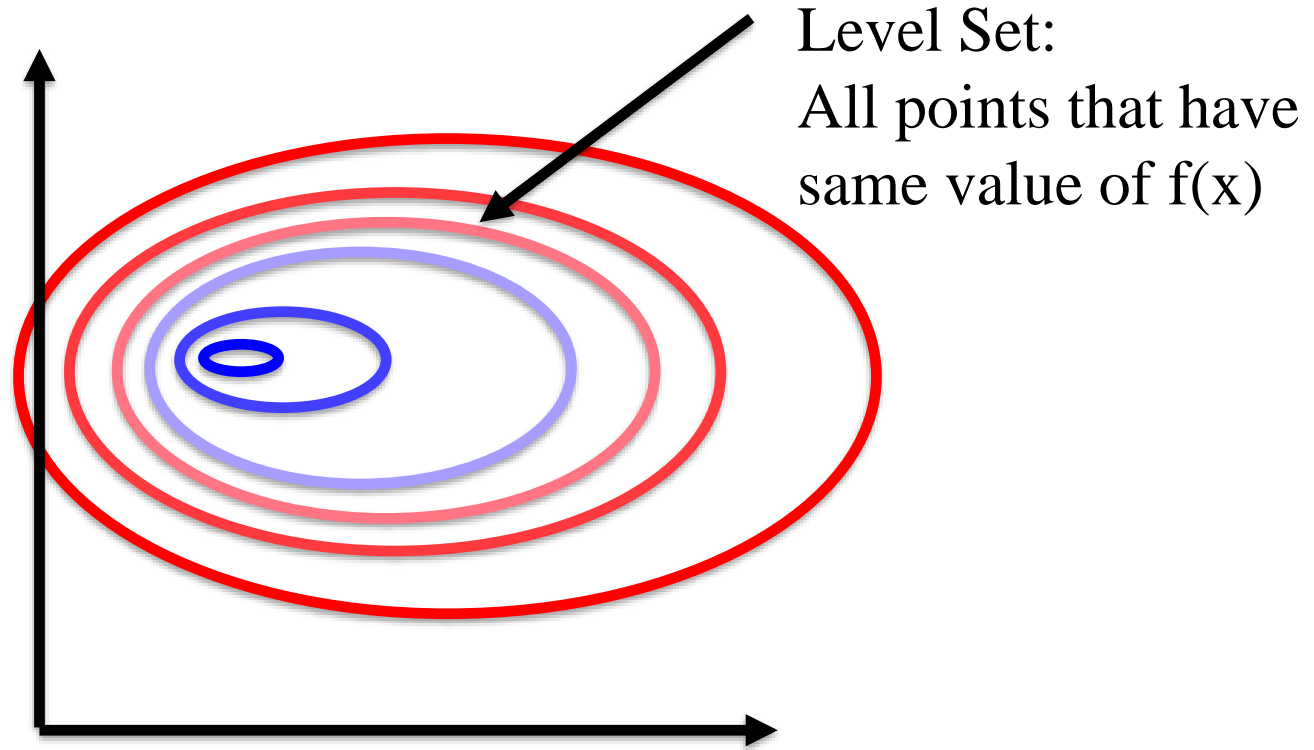
# Gradient Descent

- Make a guess.
- Iteratively move in a direction lower energy.

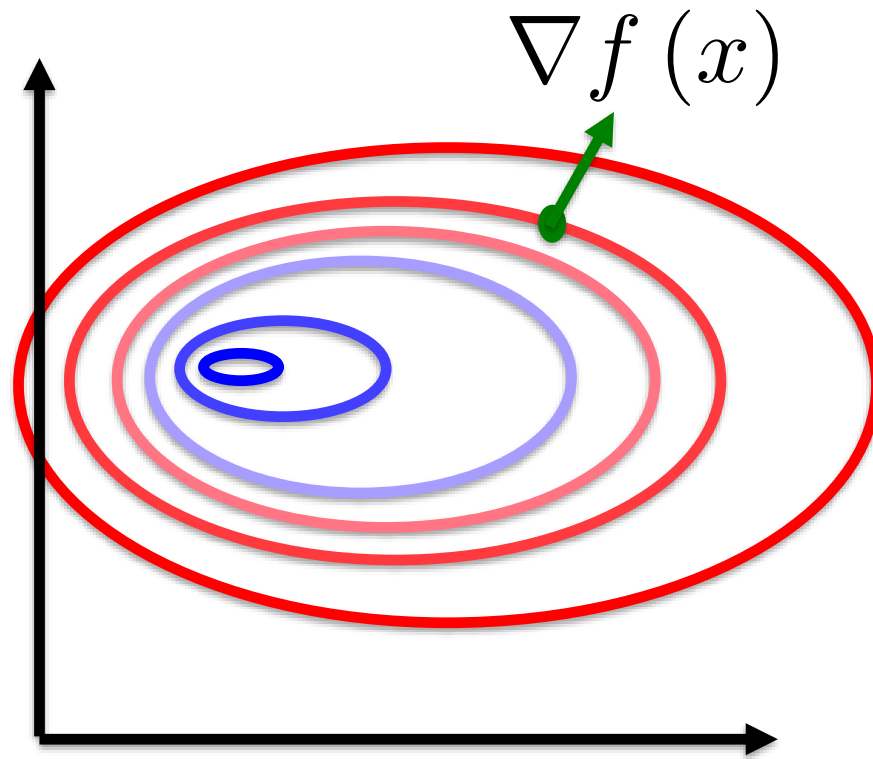
Recall that the gradient of a function points in direction of maximum ascent

$$\nabla f(\mathbf{x}) = \left( \frac{\partial f}{\partial \mathbf{x}_1} \quad \frac{\partial f}{\partial \mathbf{x}_2} \quad \cdots \quad \frac{\partial f}{\partial \mathbf{x}_n} \right)$$

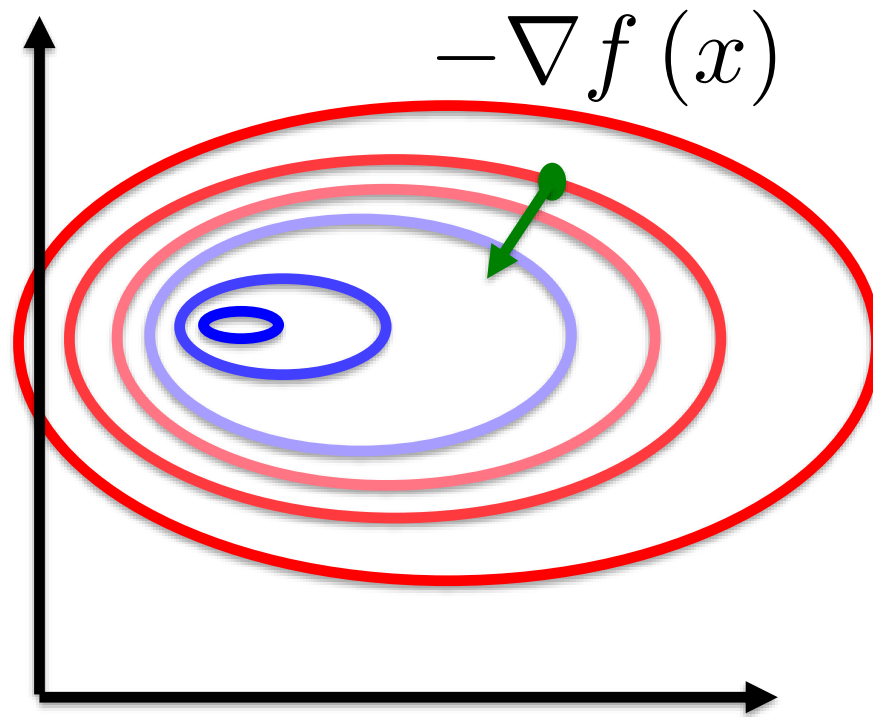
## An Aside: Level Sets



# Gradient Descent



# Gradient Descent

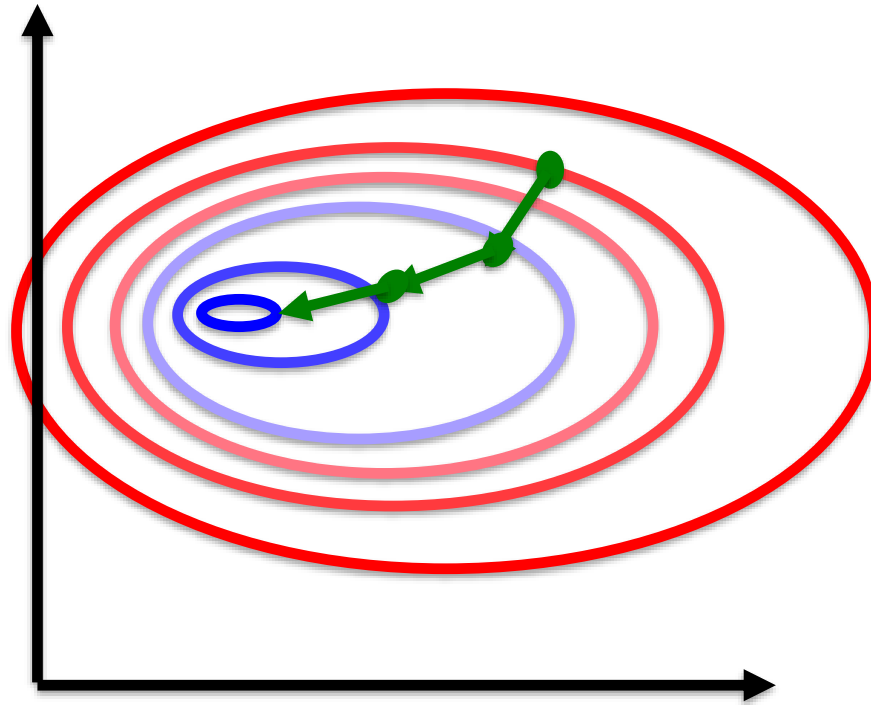


# Projected Gradient Descent Algorithm

While not at an optimal point

- Compute the gradient at current point ( $x$ )
- Move to new point  $x = x - h \nabla f(x)$
- Project  $x$  to satisfy any constraints like joint angle limits.
- Iterate to a minima.

# Gradient Descent



# Gradient Descent

So let's take a step in the *negative* gradient direction of the objective

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{dE(\mathbf{x}(\mathbf{a}))}{d\mathbf{a}} \right)^T$$



# Gradient Descent

So let's take a step in the negative gradient direction of the objective

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{dE(\mathbf{x}(\mathbf{a}))}{d\mathbf{a}} \right)^T$$

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{d\mathbf{x}(\mathbf{a})}{d\mathbf{a}} \right)^T \left( \frac{dE(\mathbf{x})}{d\mathbf{x}} \right)$$

# Gradient Descent

So let's take a step in the negative gradient direction of the objective

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{dE(\mathbf{x}(\mathbf{a}))}{d\mathbf{a}} \right)^T$$

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{d\mathbf{x}(\mathbf{a})}{d\mathbf{a}} \right)^T \left( \frac{dE(\mathbf{x})}{d\mathbf{x}} \right)$$

# Gradient Descent: Kinematic Jacobian

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{d\mathbf{x}(\mathbf{a})}{d\mathbf{a}} \right)^T \left( \frac{dE(\mathbf{x})}{d\mathbf{x}} \right)$$

The change in tip positions  $\mathbf{x}$  with respect to joint angles  $\mathbf{a}$

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \mathbf{J}^T \left( \frac{dE(\mathbf{x})}{d\mathbf{x}} \right)$$

Computed using finite differences

# Line Search

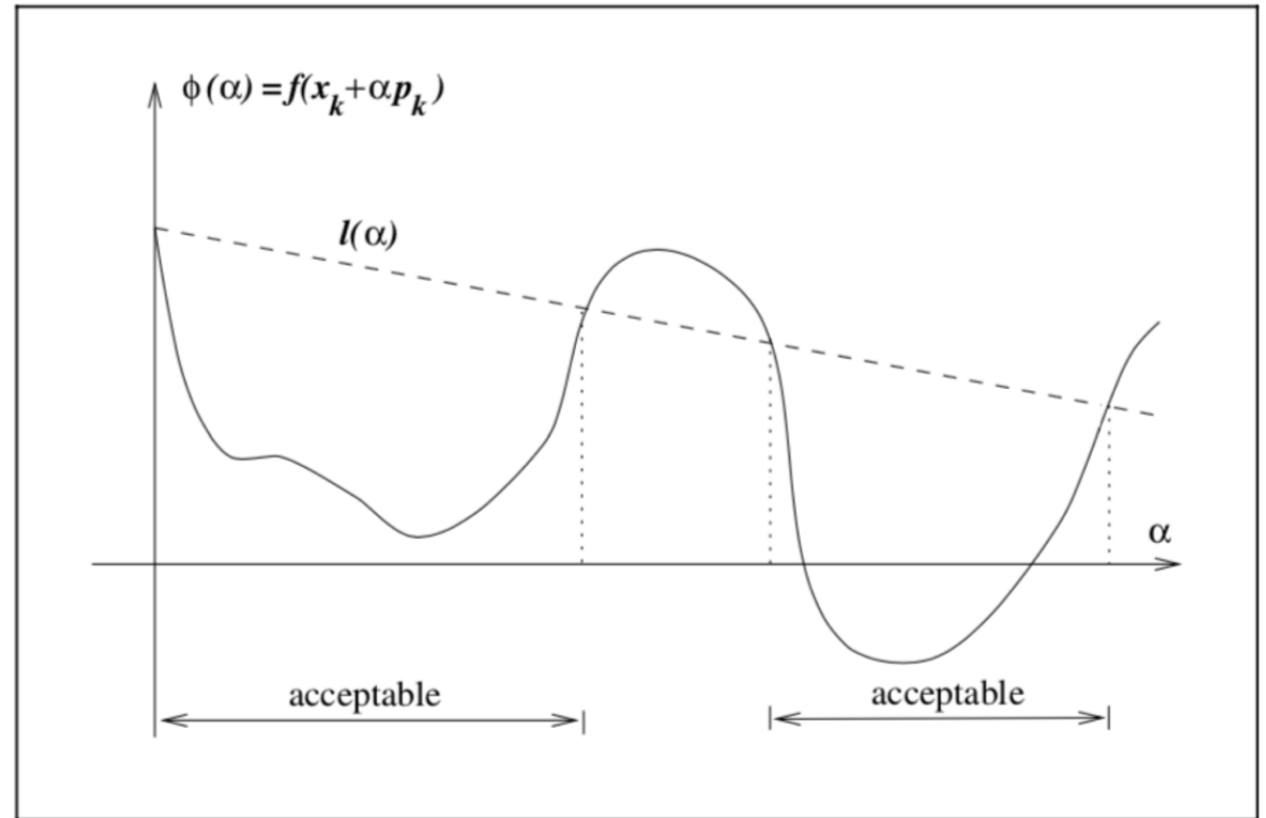
$$\mathbf{a} \leftarrow \mathbf{a} - \boxed{\sigma} \mathbf{J}^T \left( \frac{dE(\mathbf{x})}{d\mathbf{x}} \right)$$

AKA we are moving in descent direction and then projecting:

$$\mathbf{a} \leftarrow \text{proj}(\mathbf{a} + \Delta \mathbf{a})$$

Start with large  $\sigma$   
and decrease by  $\frac{1}{2}$  until

$$E(\text{proj}(\mathbf{a} + \sigma \Delta \mathbf{a})) < E(\mathbf{a})$$



# Skeletons: Inverse Kinematics Minimization

## Projected Gradient Descent:

Start with an initial pose  $\mathbf{a}$ , and move in direction of decrease in  $E$ , project the pose to stay within limits and iterate towards solution.

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{dE(\mathbf{x}(\mathbf{a}))}{d\mathbf{a}} \right)^T$$

$$\mathbf{a} \leftarrow \mathbf{a} - \sigma \left( \frac{d\mathbf{x}(\mathbf{a})}{d\mathbf{a}} \right)^T \left( \frac{dE(\mathbf{x})}{d\mathbf{x}} \right) \quad \text{chain rule}$$

$$\frac{dE}{d\mathbf{a}} \in \mathbb{R}^{|\mathbf{a}|}, \quad \frac{dE}{d\mathbf{x}} \in \mathbb{R}^{|\mathbf{x}|}, \quad \text{and} \quad \frac{d\mathbf{x}}{d\mathbf{a}} \in \mathbb{R}^{|\mathbf{x}| \times |\mathbf{a}|}$$

$$\mathbf{J} = \frac{d\mathbf{x}}{d\mathbf{a}}. \quad \text{also known as Jacobian measures the change in } \mathbf{x} \text{ for changes in joint angles } \mathbf{a},$$

$$\mathbf{J} \text{ can be computed using Finite Differences:} \quad \mathbf{J}_{i,j} \approx \frac{\mathbf{x}_i(\mathbf{a} + h\delta_j)}{h}. \quad h = 10^{-7}$$

$$\left( \frac{dE(\mathbf{x})}{d\mathbf{x}} \right) \text{ is gradient of } \sum_{i=1}^k \|\mathbf{x}_{b_i}(\mathbf{a}) - \hat{\mathbf{x}}_{b_i}\|^2$$

$$\text{Project to within limits: } \mathbf{a}_i \leftarrow \max[\mathbf{a}_i^{\min}, \min[\mathbf{a}_i^{\max}, \mathbf{a}_i]].$$

$$\text{Find a good step that lowers energy: } E(\text{proj}(\mathbf{a} + \sigma \Delta \mathbf{a})) < E(\mathbf{a}).$$

## Next: Simulation, mass-spring systems

