

CS 523: Computer Graphics, Spring 2011

# Interactive Shape Modeling

Space deformations

# Space Deformation

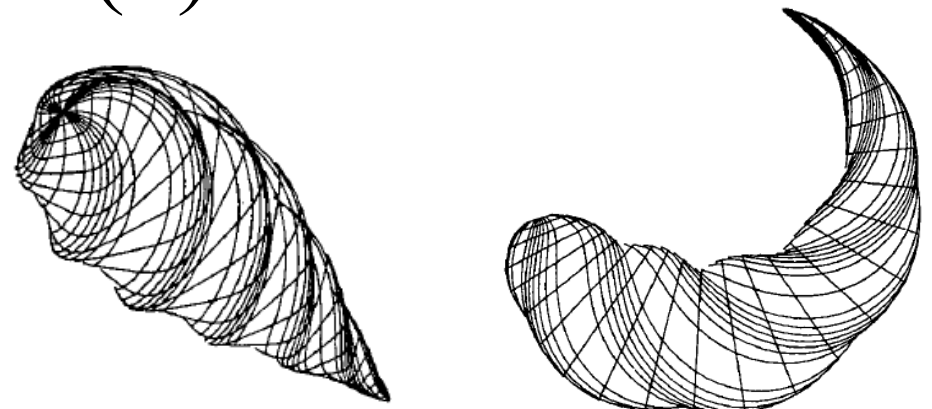
- Displacement function defined on the ambient space

$$\mathbf{d} : \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

- Evaluate the function on the points of the shape embedded in the space

$$\mathbf{x}' = \mathbf{x} + \mathbf{d}(\mathbf{x})$$

Twist warp  
Global and local deformation of solids  
[A. Barr, SIGGRAPH 84]

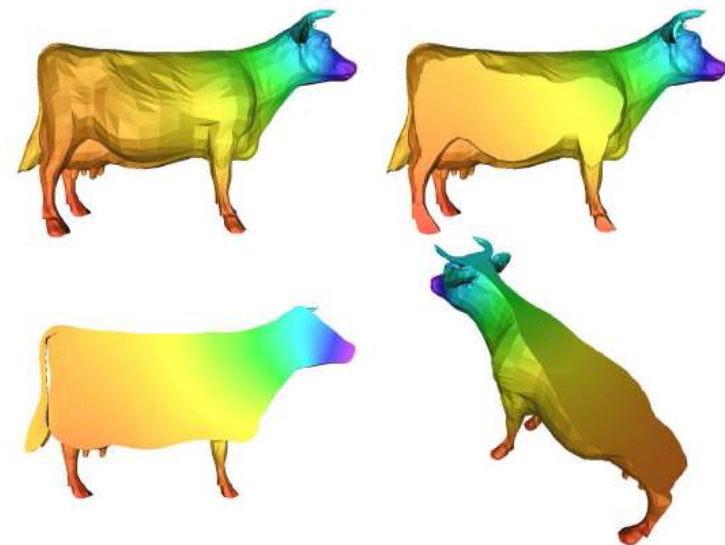


# Freeform Deformations

- Control object
- User defines displacements  $\mathbf{d}_i$  for each element of the control object
- Displacements are interpolated to the entire space using basis functions  $B_i(\mathbf{x}) : \mathbb{R}^3 \rightarrow \mathbb{R}$

$$\mathbf{d}(\mathbf{x}) = \sum_{i=1}^k \mathbf{d}_i B_i(\mathbf{x})$$

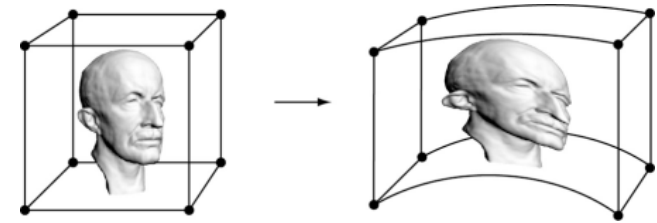
- Basis functions should be smooth for aesthetic results



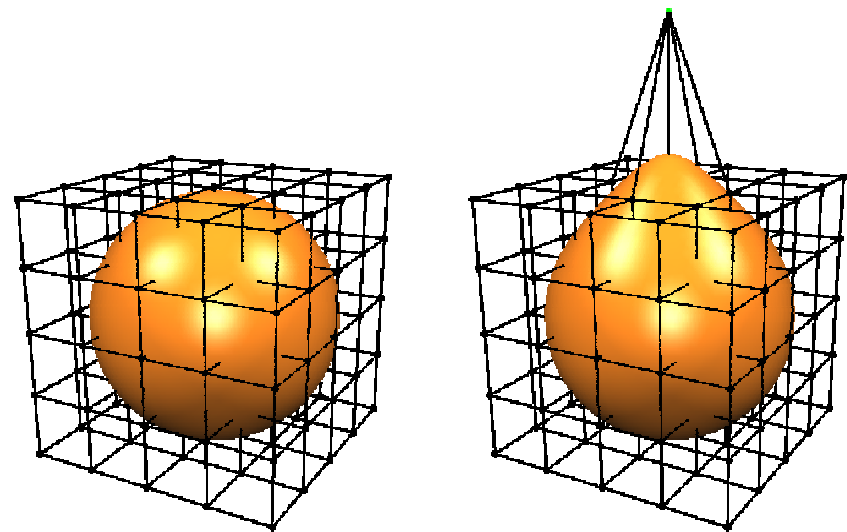
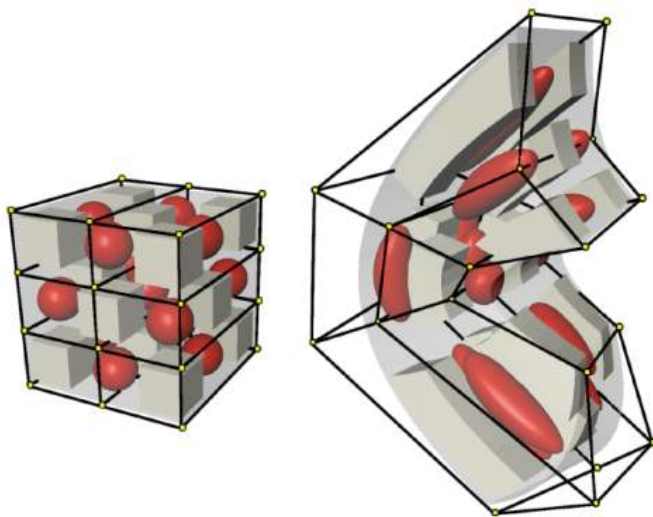
# Trivariate Tensor Product Bases

[Sederberg and Parry 86]

- Control object = lattice
- Basis functions  $B_i(\mathbf{x})$  are trivariate tensor-product splines:



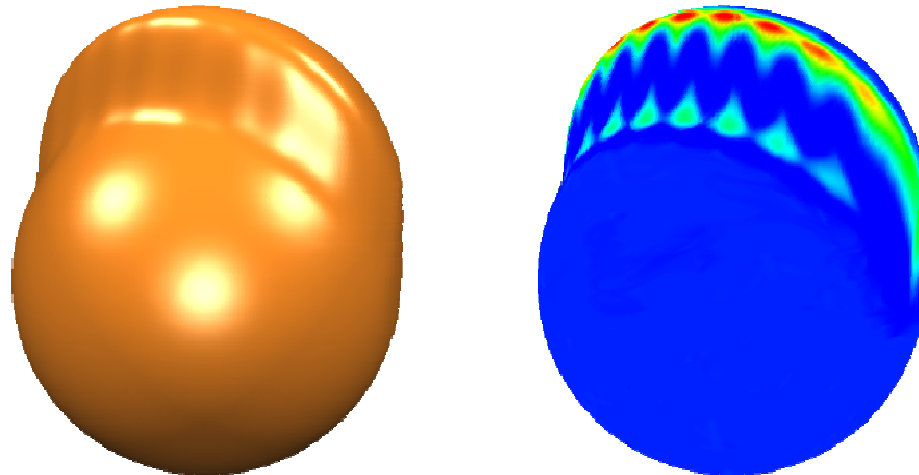
$$\mathbf{d}(x, y, z) = \sum_{i=0}^l \sum_{j=0}^m \sum_{k=0}^n \mathbf{d}_{ijk} N_i(x) N_j(y) N_k(z)$$





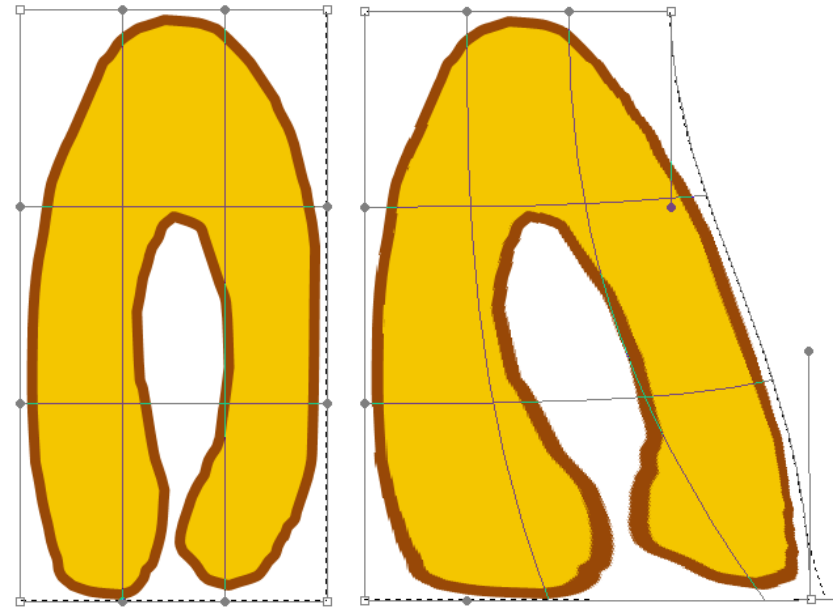
# Trivariate Tensor Product Bases

- Similar to the surface case
  - Aliasing artifacts
- Interpolate deformation constraints?
  - Only in least squares sense



# Lattice as Control Object

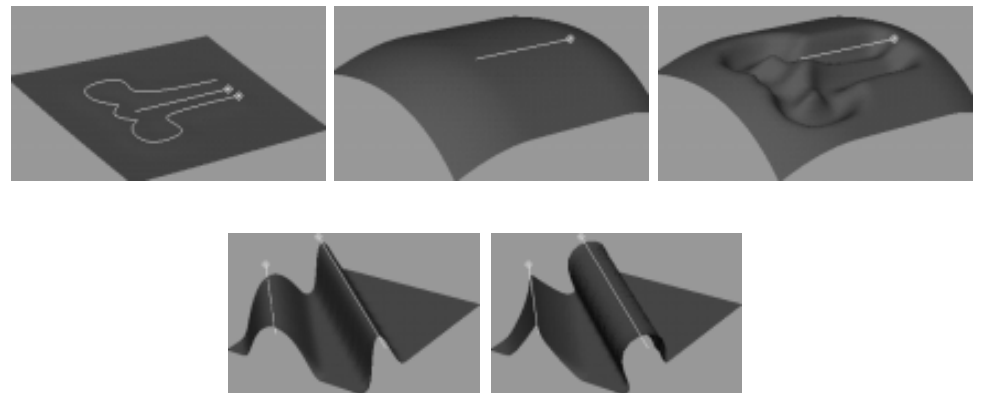
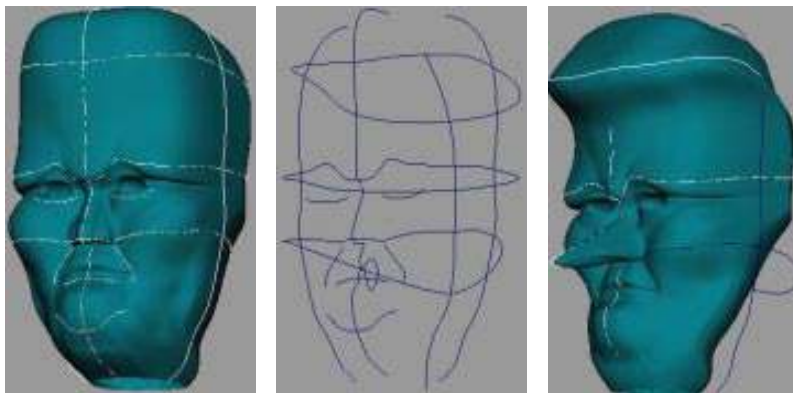
- Difficult to manipulate
- The control object is not related to the shape of the edited object
- Part of the shape in close Euclidean distance always deform similarly, even if geodesically far



# Wires

[Singh and Fiume 98]

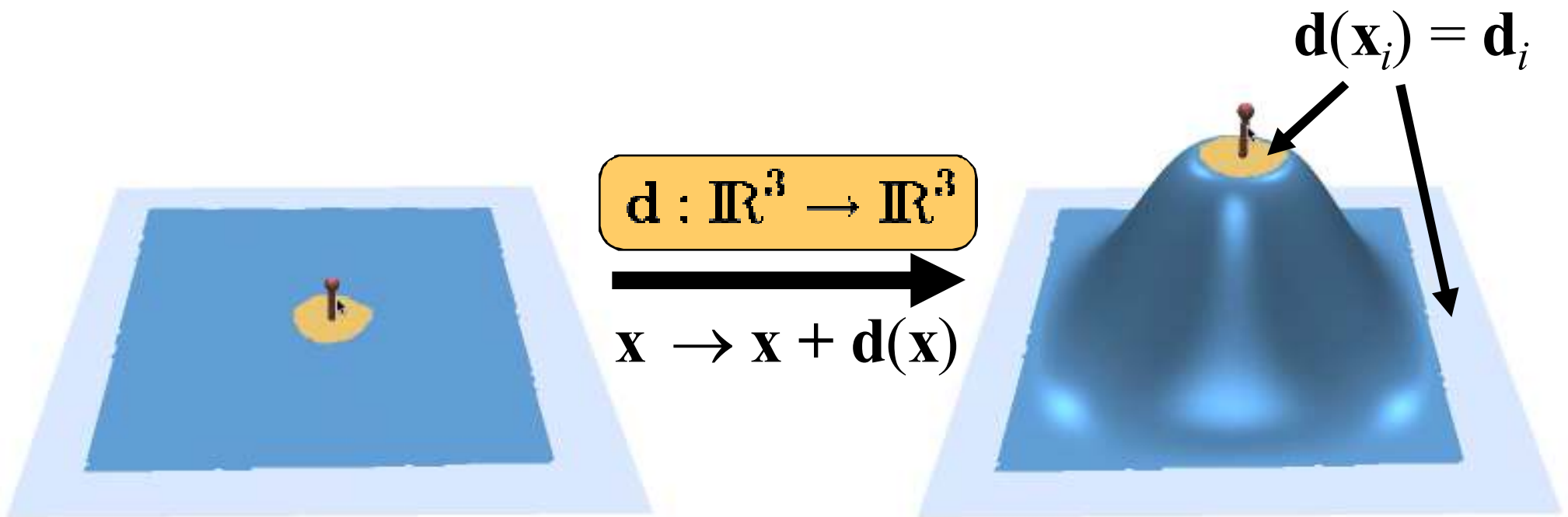
- Control objects are arbitrary space curves
- Can place curves along meaningful features of the edited object
- Smooth deformations around the curve with decreasing influence



# Handle Metaphor

[RBF, Botsch and Kobbelt 05]

- Wish list for the displacement function  $\mathbf{d}(\mathbf{x})$  :
  - Interpolate prescribed constraints
  - Smooth, intuitive deformation



# Volumetric Energy Minimization

[RBF, Botsch and Kobbelt 05]

- Minimize similar energies to surface case

$$\int_{\mathbb{R}^3} \left\| \mathbf{d}_{xx} \right\|^2 + \left\| \mathbf{d}_{xy} \right\|^2 + \dots + \left\| \mathbf{d}_{zz} \right\|^2 dx dy dz \rightarrow \min$$

- But displacements function lives in 3D...
  - Need a volumetric space tessellation?
  - No, same functionality provided by RBFs!

# Radial Basis Functions

[RBF, Botsch and Kobbelt 05]

- Represent deformation by RBFs

$$\mathbf{d}(\mathbf{x}) = \sum_j \mathbf{w}_j \cdot \varphi(\|\mathbf{c}_j - \mathbf{x}\|) + \mathbf{p}(\mathbf{x})$$

- Triharmonic basis function  $\varphi(r) = r^3$ 
  - $C^2$  boundary constraints
  - Highly smooth / fair interpolation

$$\int_{\mathcal{R}^3} \|\mathbf{d}_{xxx}\|^2 + \|\mathbf{d}_{xyy}\|^2 + \dots + \|\mathbf{d}_{zzz}\|^2 dx dy dz \rightarrow \min$$

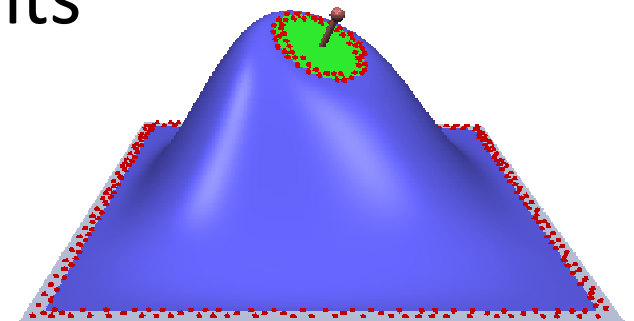
# RBF Fitting

[RBF, Botsch and Kobbelt 05]

- Represent deformation by RBFs

$$\mathbf{d}(\mathbf{x}) = \sum_j \mathbf{w}_j \cdot \varphi(\|\mathbf{c}_j - \mathbf{x}\|) + \mathbf{p}(\mathbf{x})$$

- RBF fitting
  - Interpolate displacement constraints
  - Solve linear system for  $\mathbf{w}_j$  and  $\mathbf{p}$



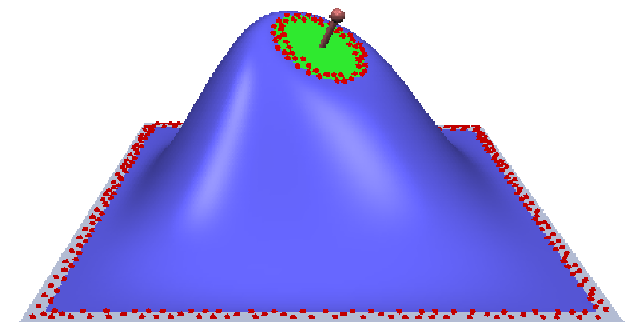
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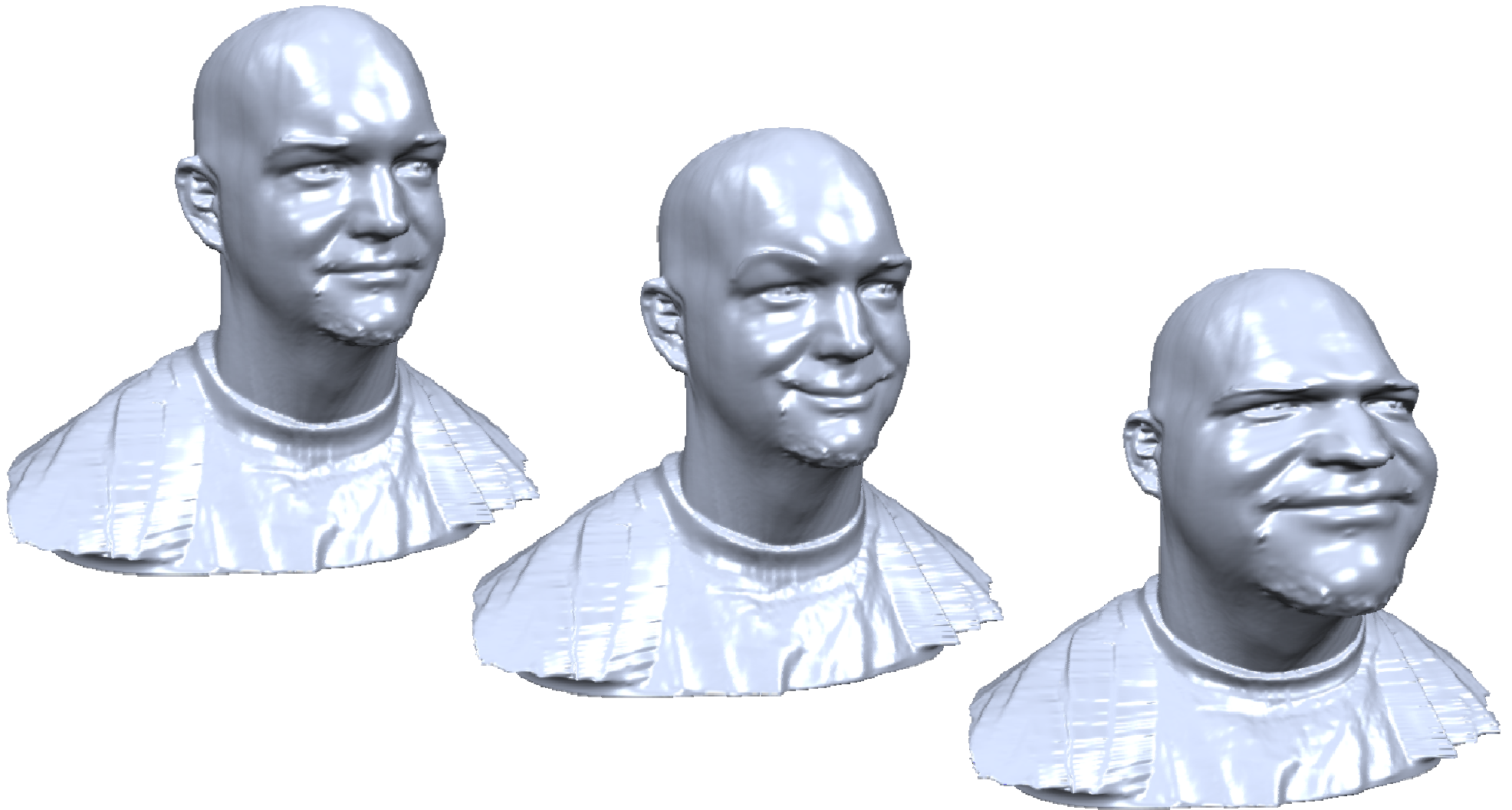
- RBF evaluation
  - Function  $\mathbf{d}$  transforms points
  - Jacobian  $\nabla \mathbf{d}$  transforms normals
  - Precompute basis functions
  - Evaluate on the GPU!





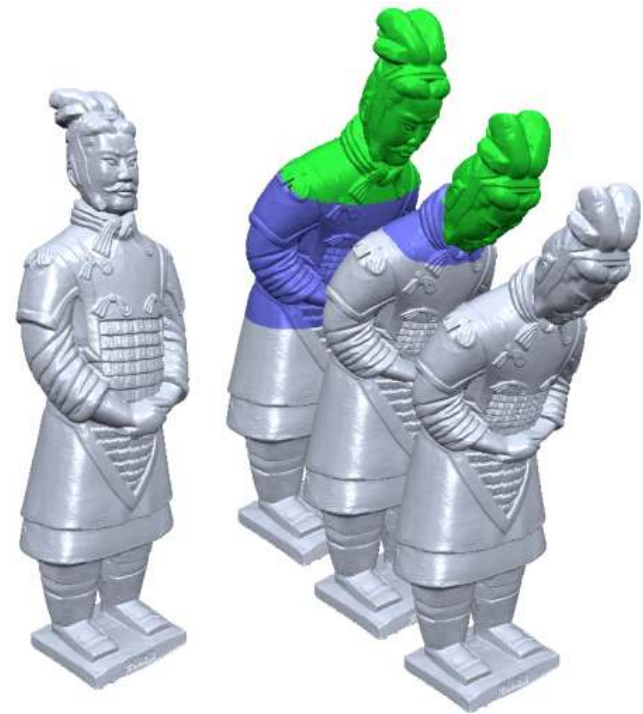
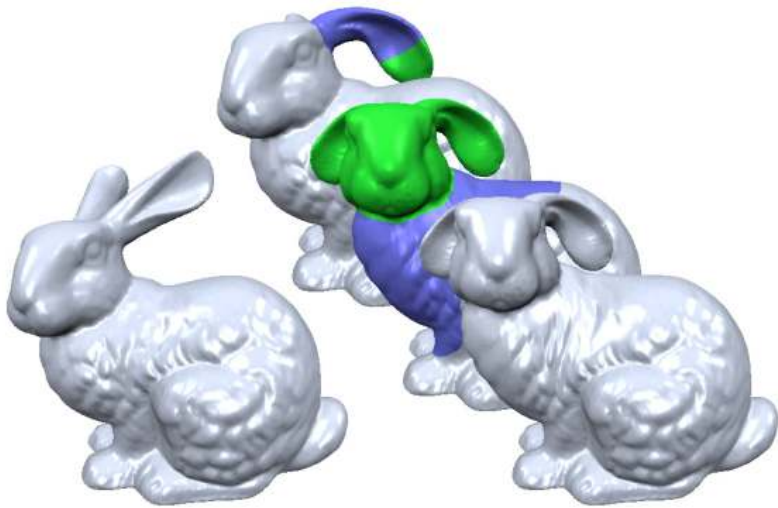
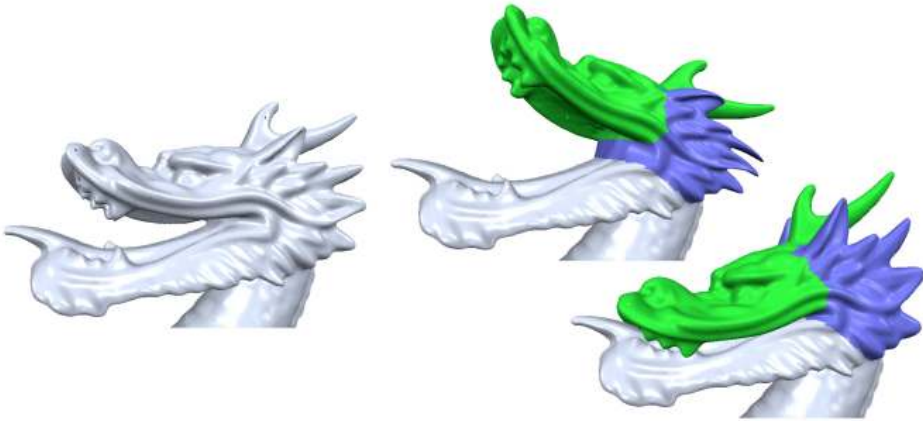
# Local & Global Deformations

[RBF, Botsch and Kobbelt 05]



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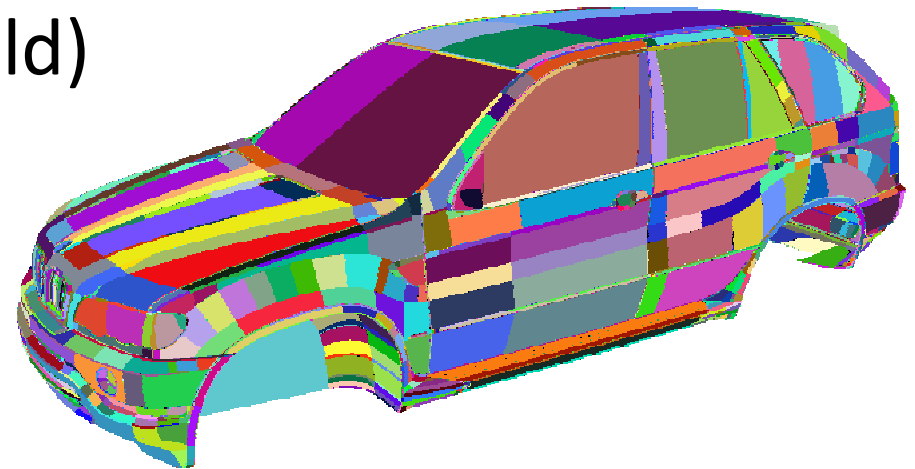


1M vertices  
movie

# Space Deformations

Summary so far

- Handle arbitrary input
  - Meshes (also non-manifold)
  - Point sets
  - Polygonal soups
  - ...
- Complexity mainly depends on the control object, not the surface



- 3M triangles
- 10k components
- Not oriented
- Not manifold

# Space Deformations

Summary so far

- Handle arbitrary input
  - Meshes (also non-manifold)
  - Point sets
  - Polygonal soups
  - ...



- Easier to analyze: functions on Euclidean domain

- Volume preservation:  $|\text{Jacobian}| = 1$

$$\mathbf{F}(x,y,z) = (F(x,y,z), G(x,y,z), H(x,y,z))$$

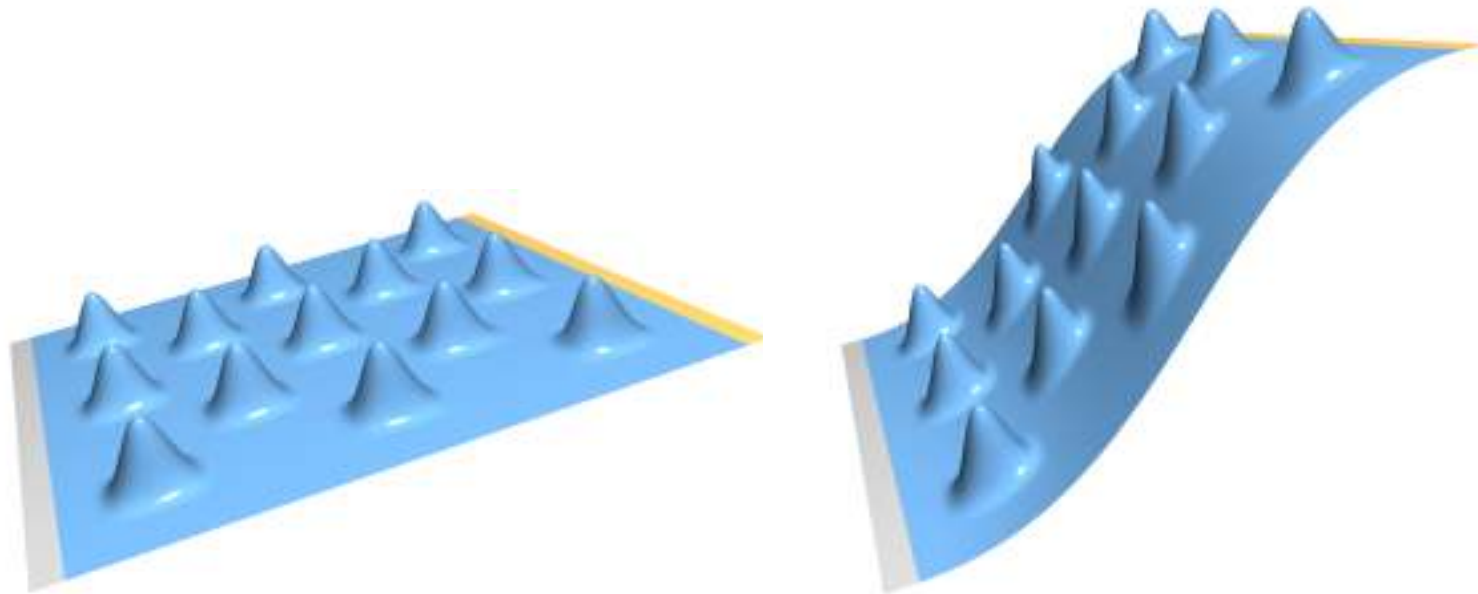
then the Jacobian is the determinant

$$Jac(\mathbf{F}) = \begin{vmatrix} \frac{\partial F}{\partial x} & \frac{\partial F}{\partial y} & \frac{\partial F}{\partial z} \\ \frac{\partial G}{\partial x} & \frac{\partial G}{\partial y} & \frac{\partial G}{\partial z} \\ \frac{\partial H}{\partial x} & \frac{\partial H}{\partial y} & \frac{\partial H}{\partial z} \end{vmatrix}$$

# Space Deformations

Summary so far

- The deformation is only loosely aware of the shape that is being edited
- Small Euclidean distance  $\rightarrow$  similar deformation
- Local surface detail may be distorted

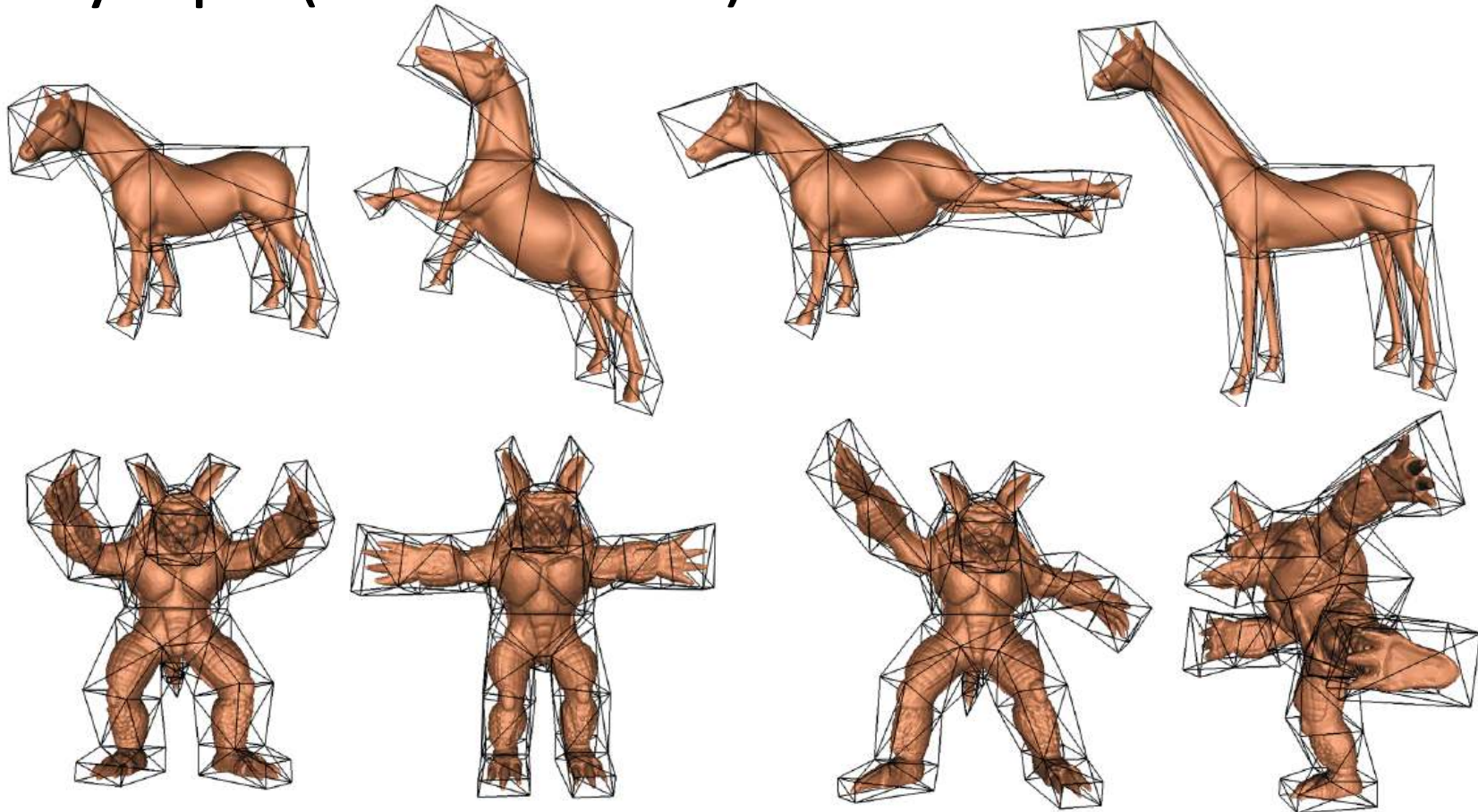




# Cage-based Deformations

[Ju et al. 2005]

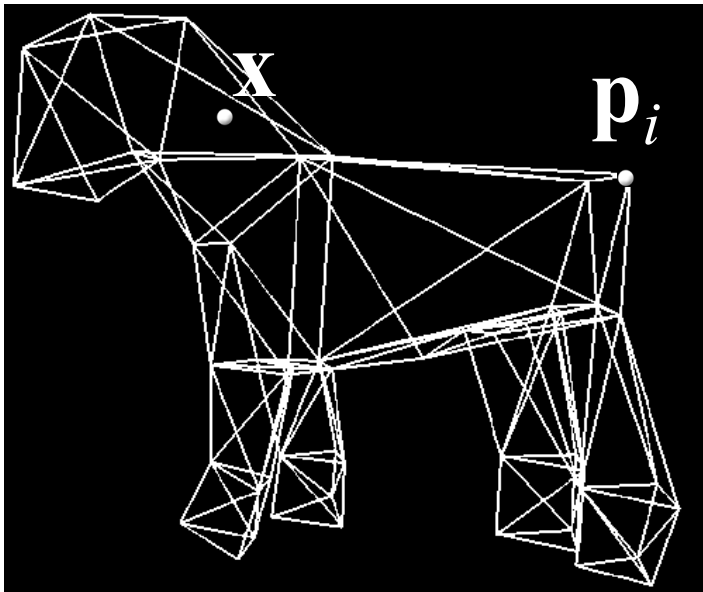
- Cage = crude version of the input shape
- Polytope (not a lattice)



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- Cage = crude version of the input shape
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- Each point  $\mathbf{x}$  in space is represented w.r.t. to the cage elements using coordinate functions

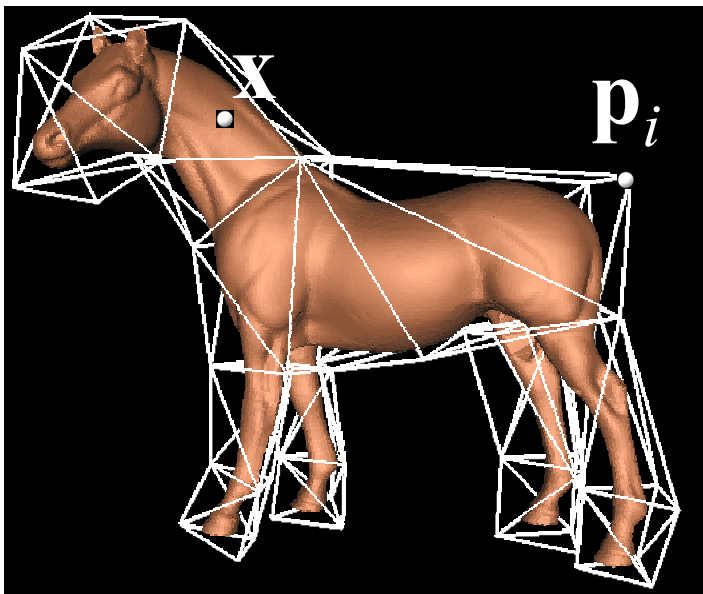


$$\mathbf{x} = \sum_{i=1}^k w_i(\mathbf{x}) \mathbf{p}_i$$

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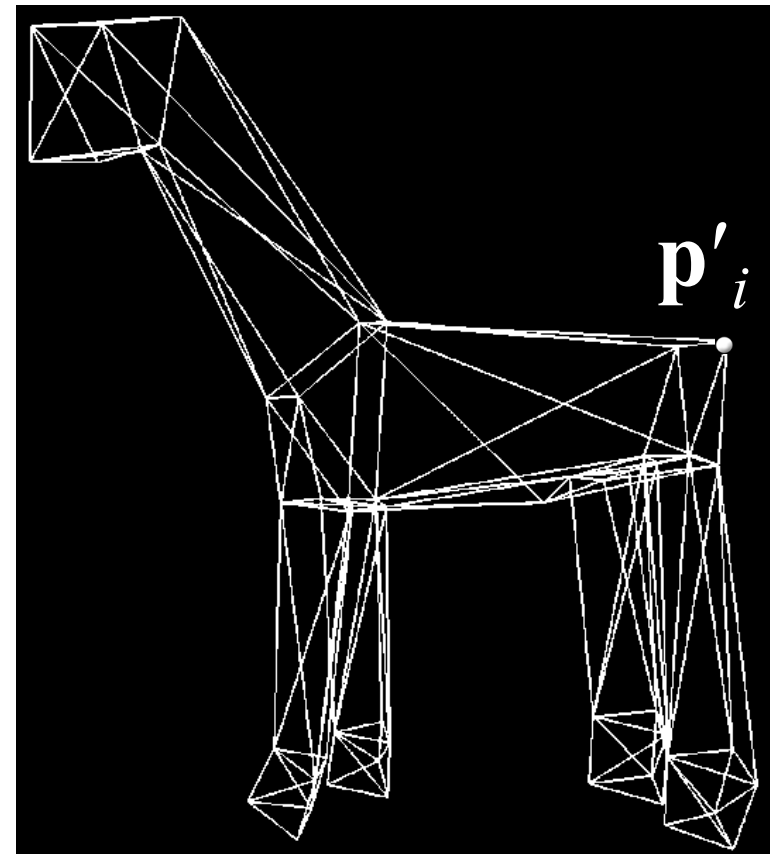
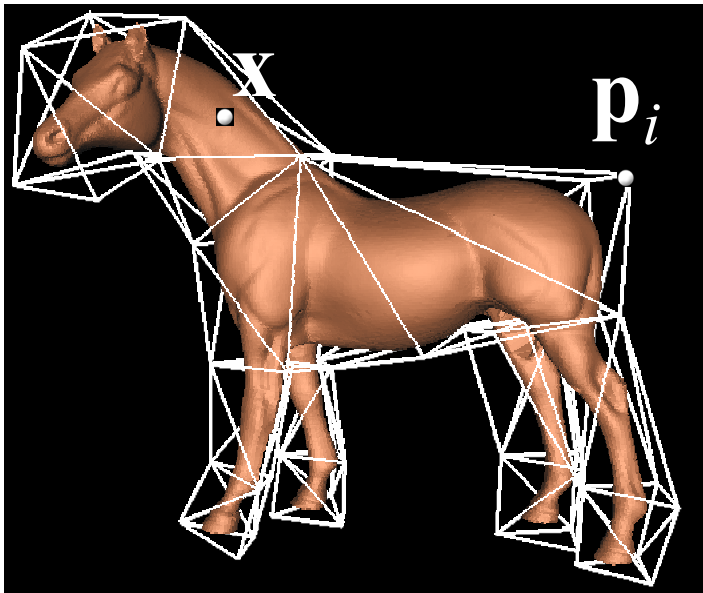
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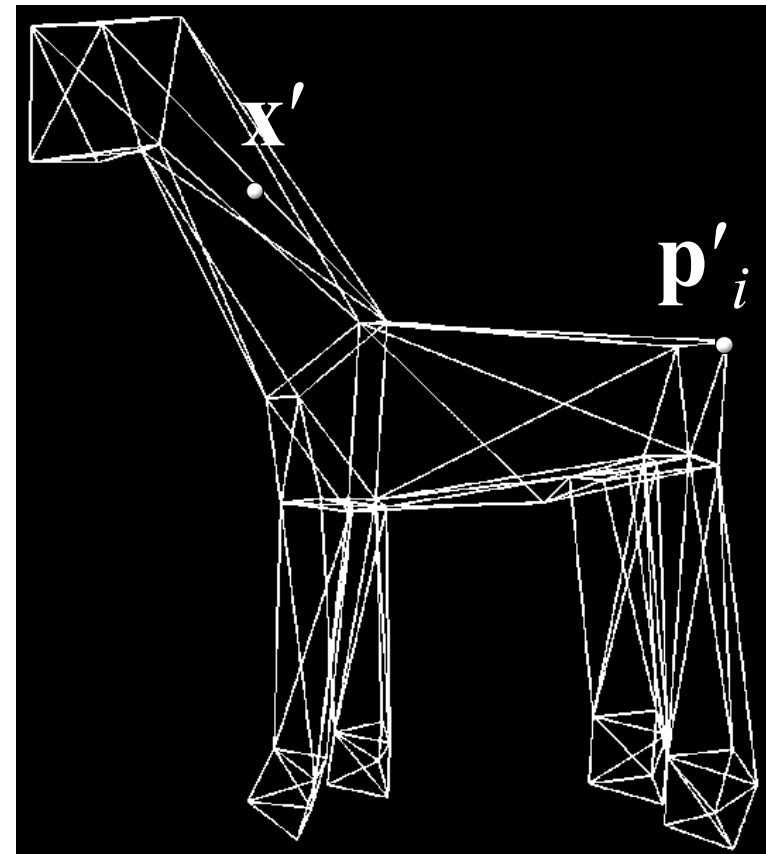
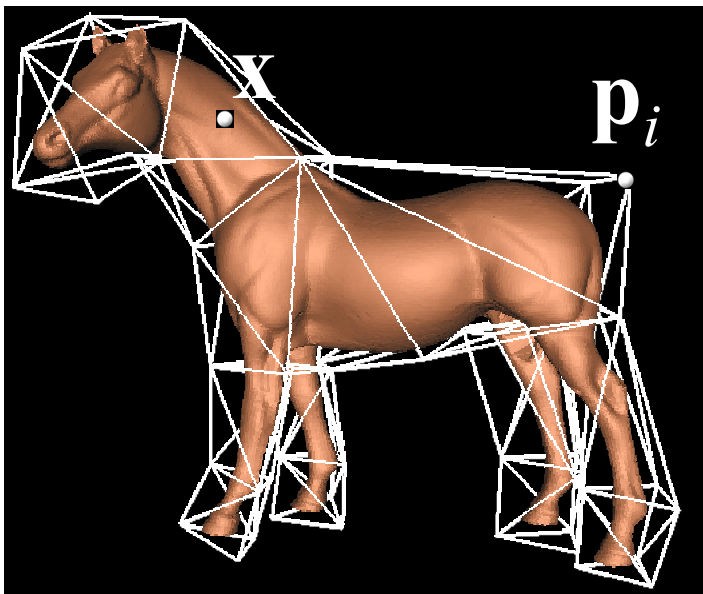


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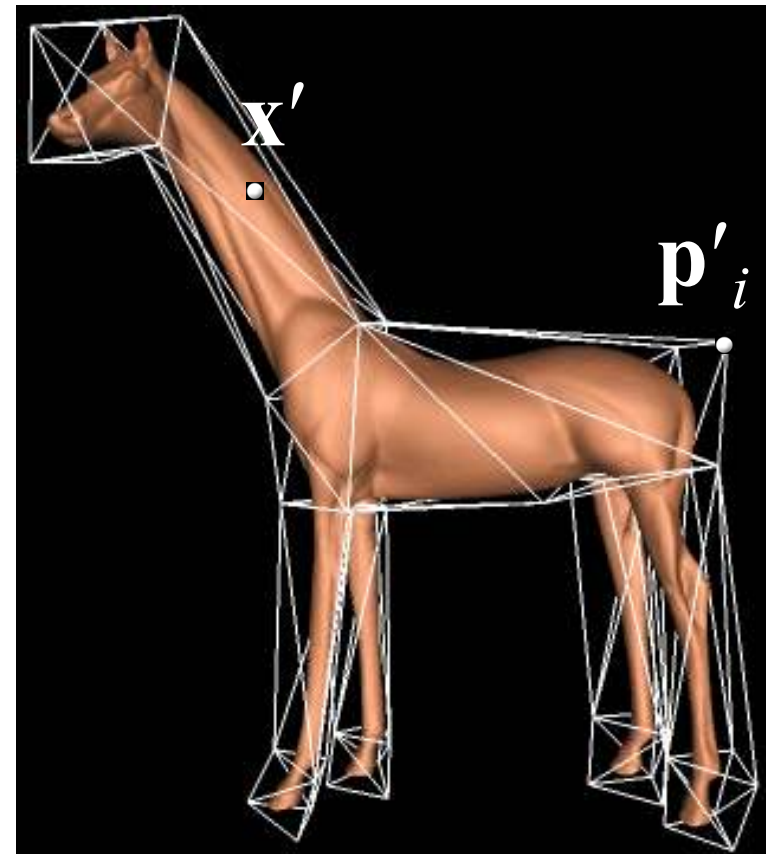
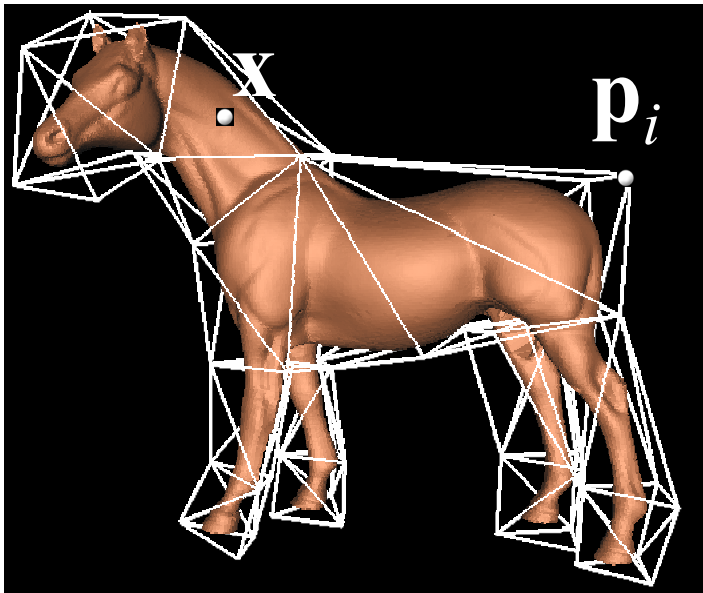


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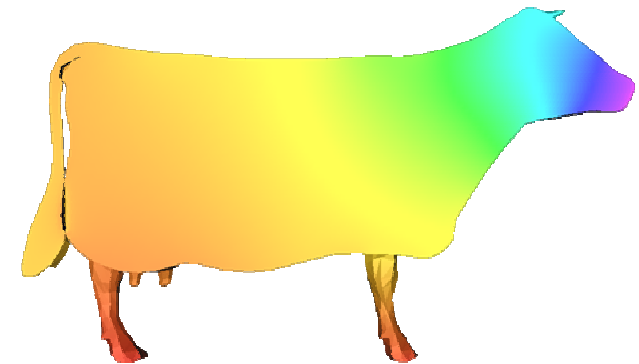
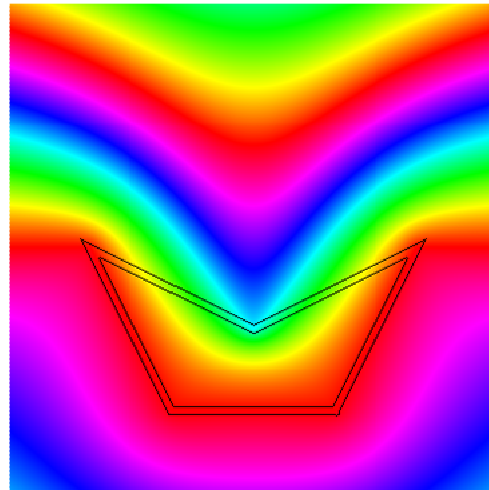
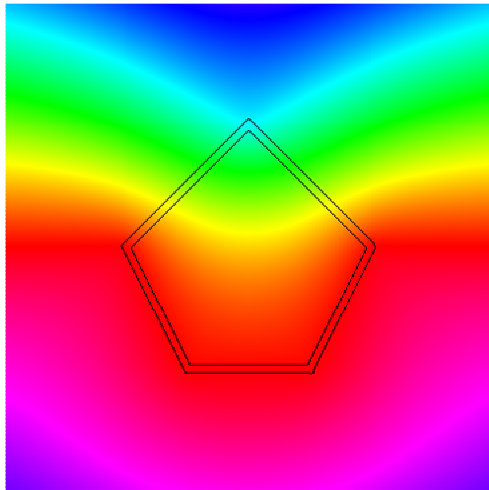
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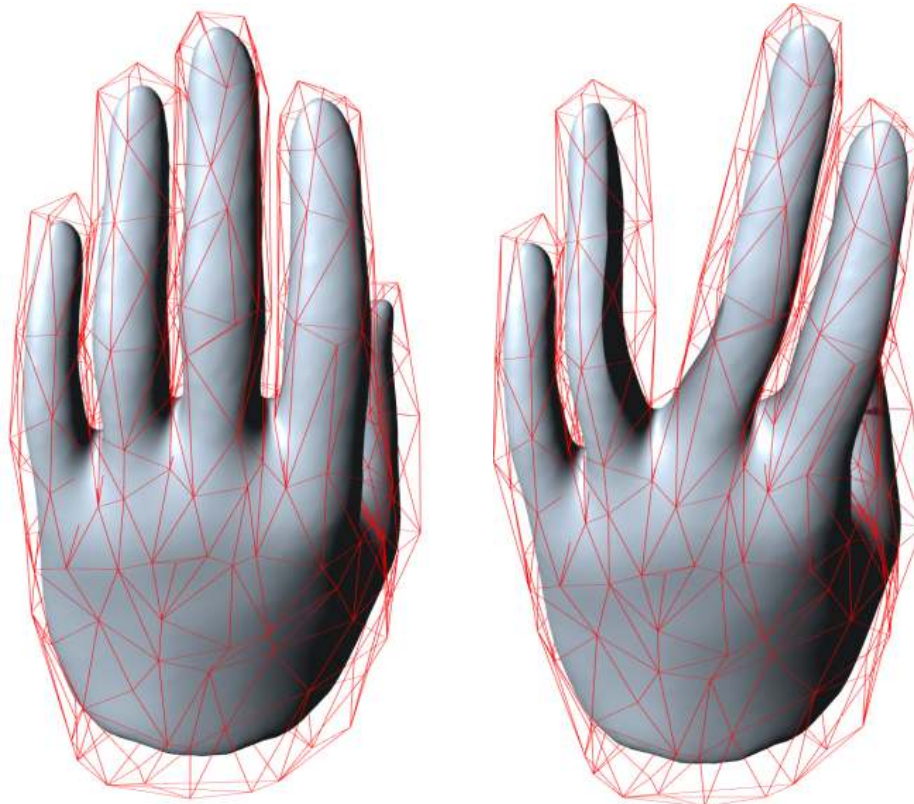
# Coordinate Functions

- Mean-value coordinates (Floater, Ju et al. 2005)
  - Generalization of barycentric coordinates
  - Closed-form solution for  $w_i(\mathbf{x})$



# Coordinate Functions

- Mean-value coordinates (Floater, Ju et al. 2005)
  - Not necessarily positive on non-convex domains



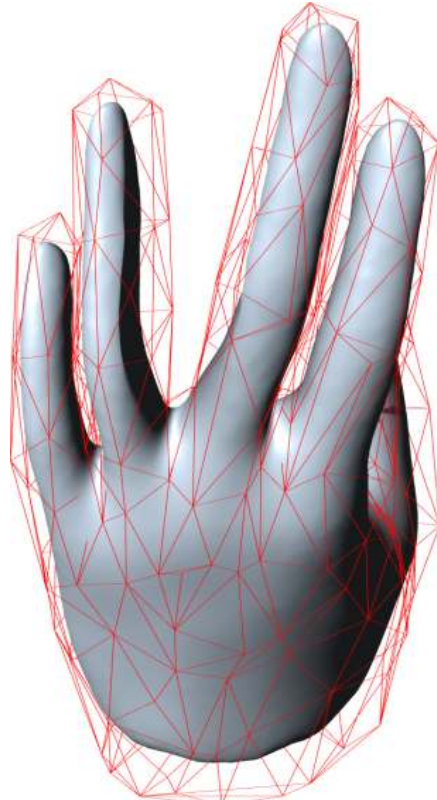
MVC

# Coordinate Functions

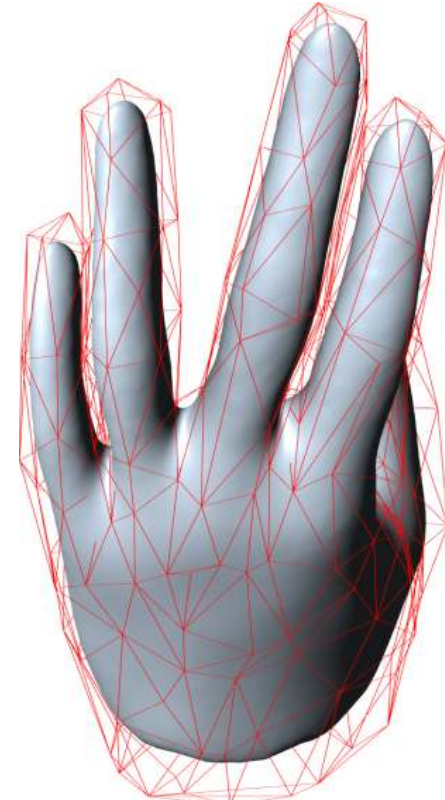
- PMVC (Lipman et al. 2007) – ensures positivity, but no longer closed-form and only  $C^0$



MVC



PMVC



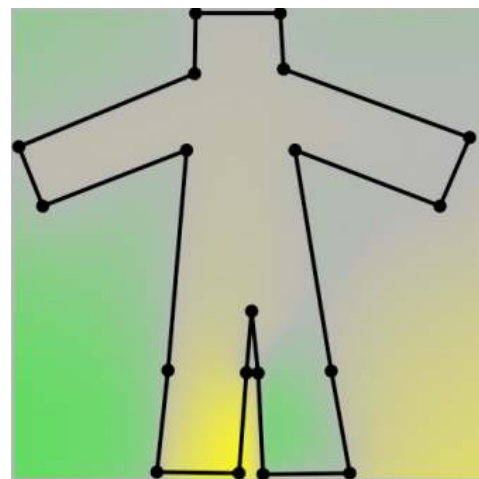


# Coordinate Functions

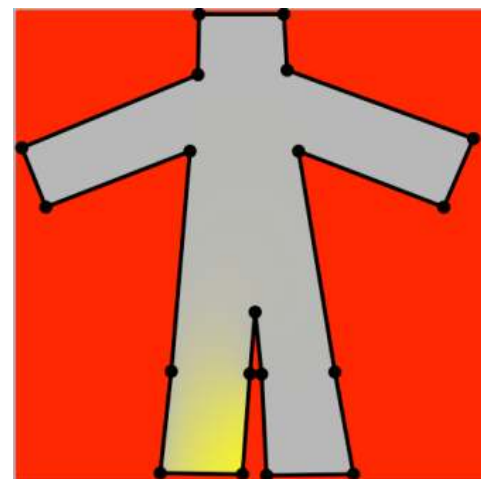
- Harmonic coordinates (Joshi et al. 2007)
  - Harmonic functions  $h_i(\mathbf{x})$  for each cage vertex  $\mathbf{p}_i$
  - Solve

$$\Delta h = 0$$

subject to:  $h_i$  linear on the boundary s.t.  $h_i(\mathbf{p}_i) = \delta_{ij}$



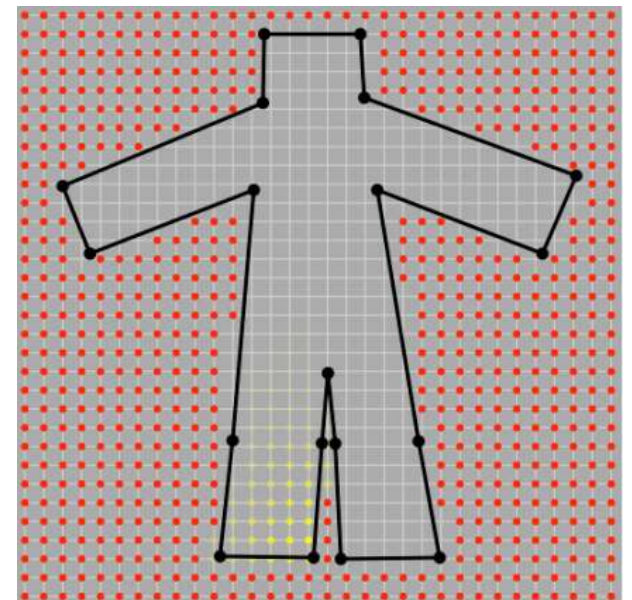
MVC



HC

# Coordinate Functions

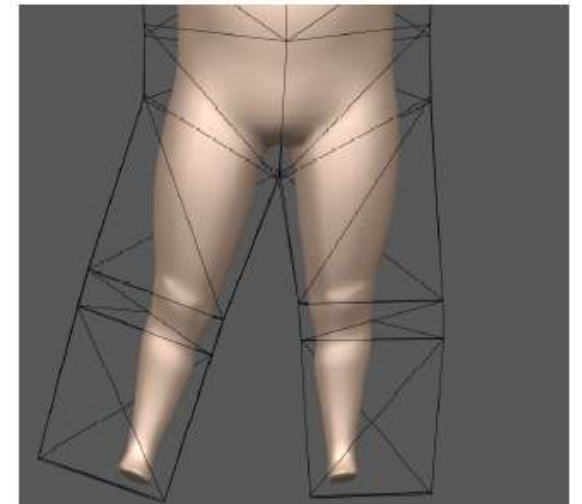
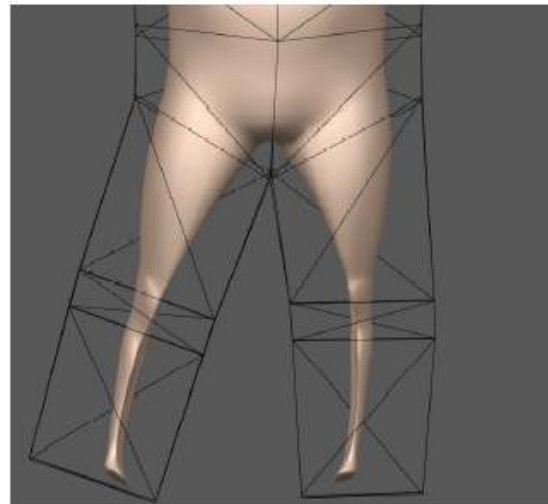
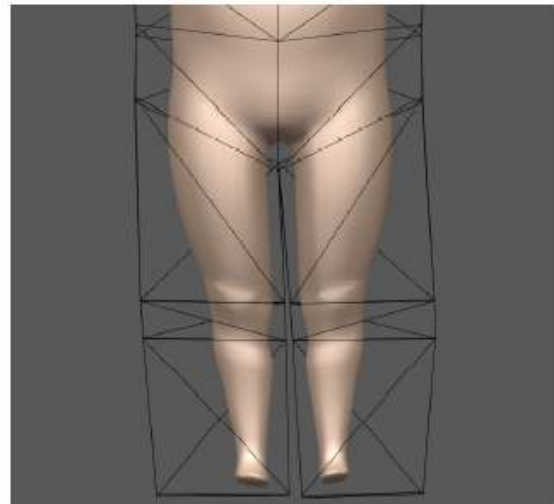
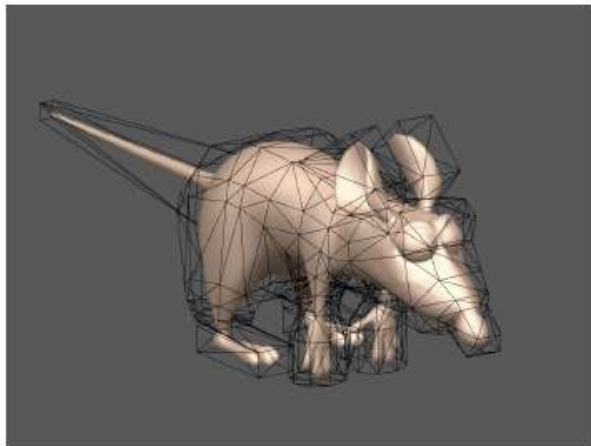
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  - Solve
$$\Delta h = 0$$
subject to:  $h_i$  linear on the boundary s.t.  $h_i(\mathbf{p}_i) = \delta_{ij}$
- Volumetric Laplace equation
- Discretization, no closed-form





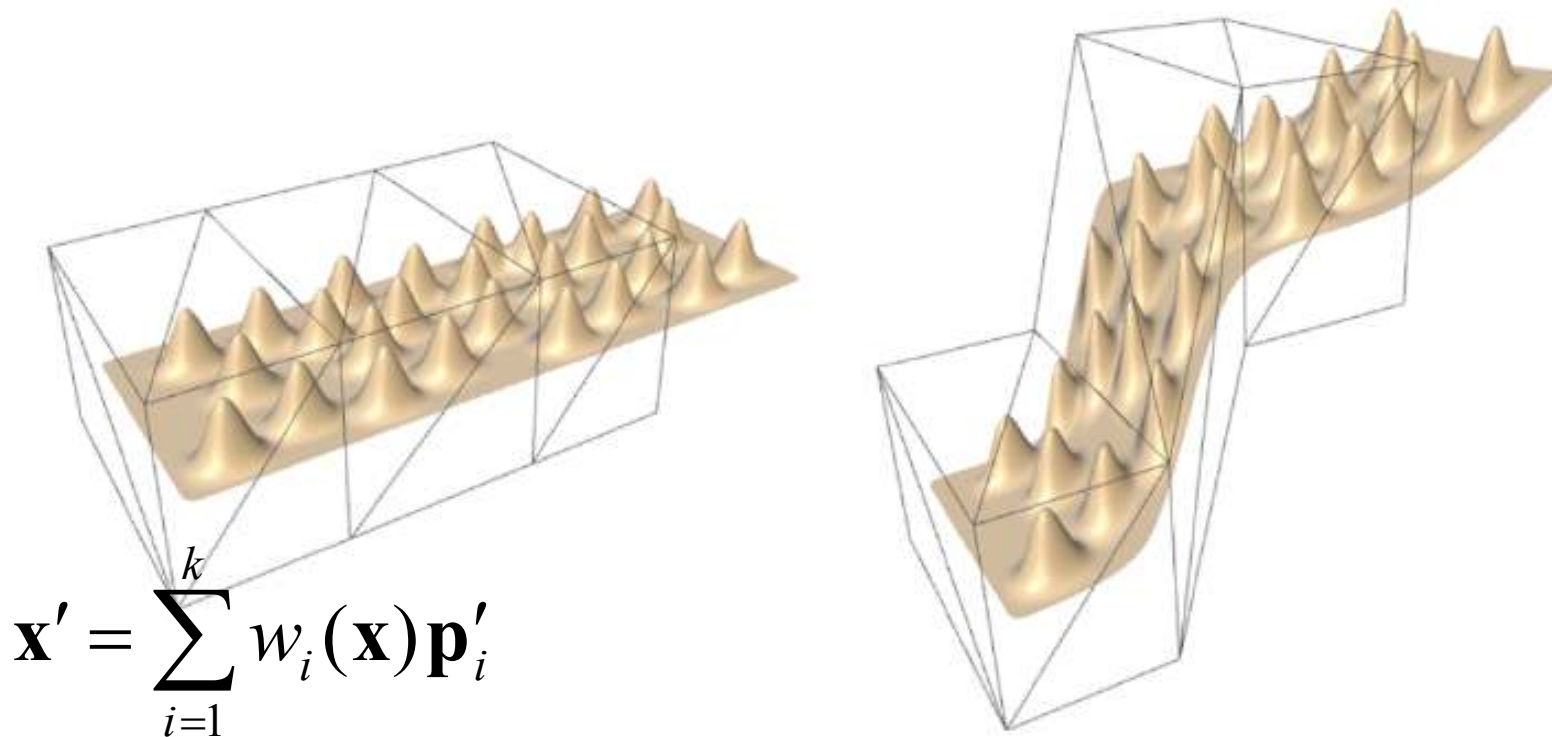
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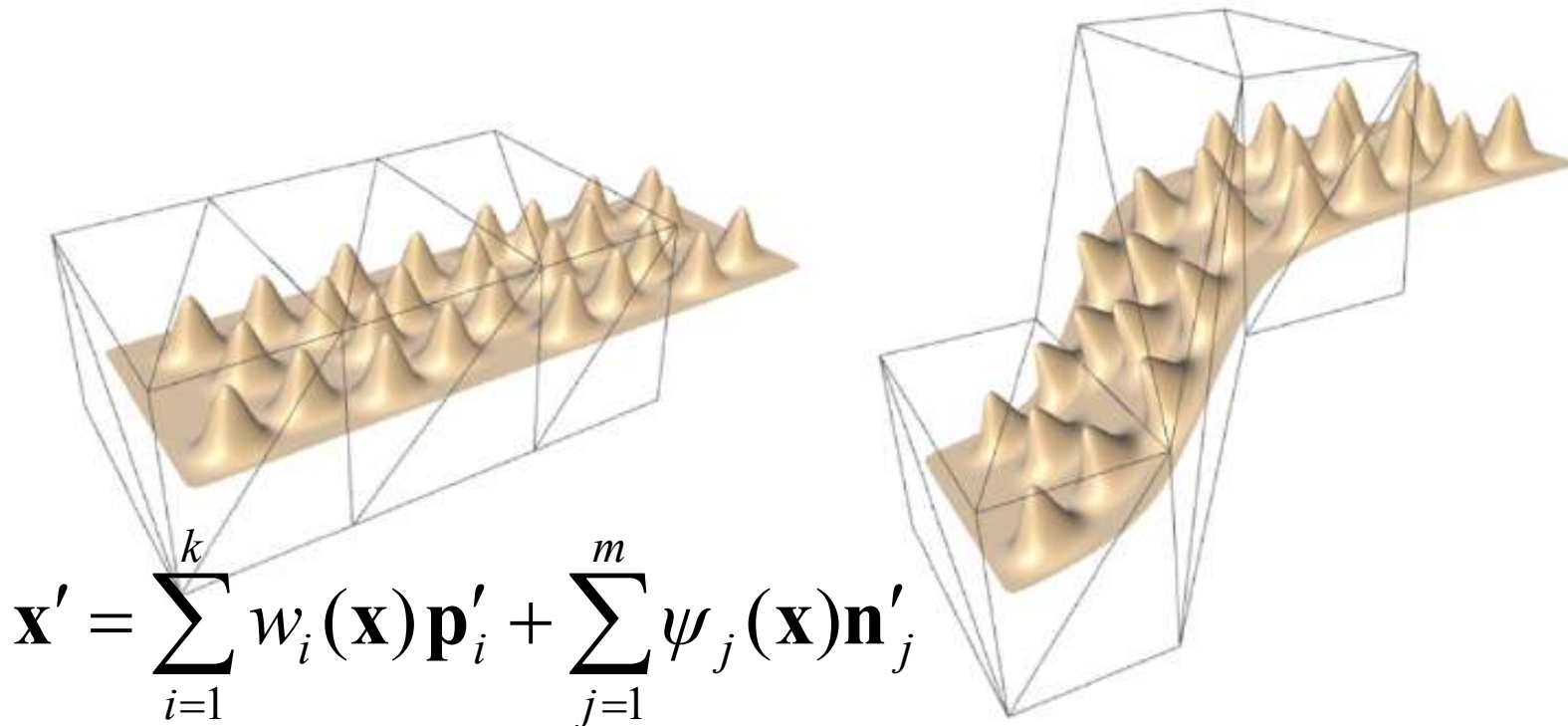
# Coordinate Functions

- Green coordinates (Lipman et al. 2008)
- Observation: previous vertex-based basis functions always lead to affine-invariance!



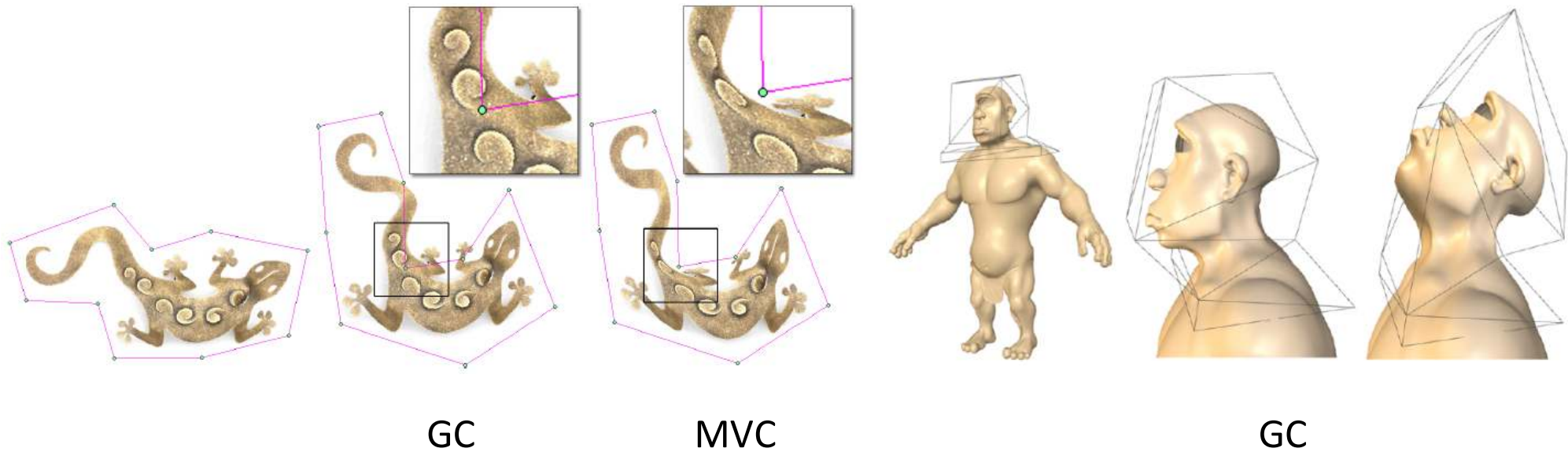
# Coordinate Functions

- Green coordinates (Lipman et al. 2008)
- Correction: Make the coordinates depend on the cage faces as well



# Coordinate Functions

- Green coordinates (Lipman et al. 2008)
- Closed-form solution
- Conformal in 2D, quasi-conformal in 3D



# Coordinate Functions

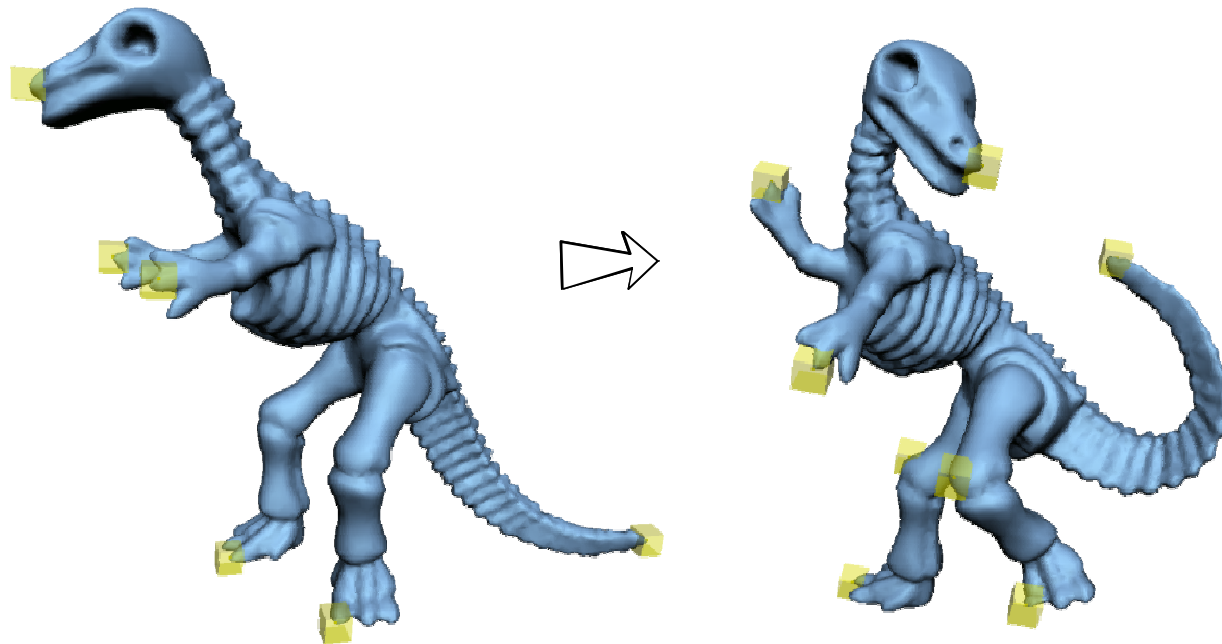
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- Conformal in 2D, quasi-conformal in 3D

Alternative interpretation in 2D via holomorphic functions and extension to point handles : **Weber et al. Eurographics 2009**



# Nonlinear Space Deformations

- Involve nonlinear optimization
- Enjoy the advantages of space warps
- Additionally, have shape-preserving properties

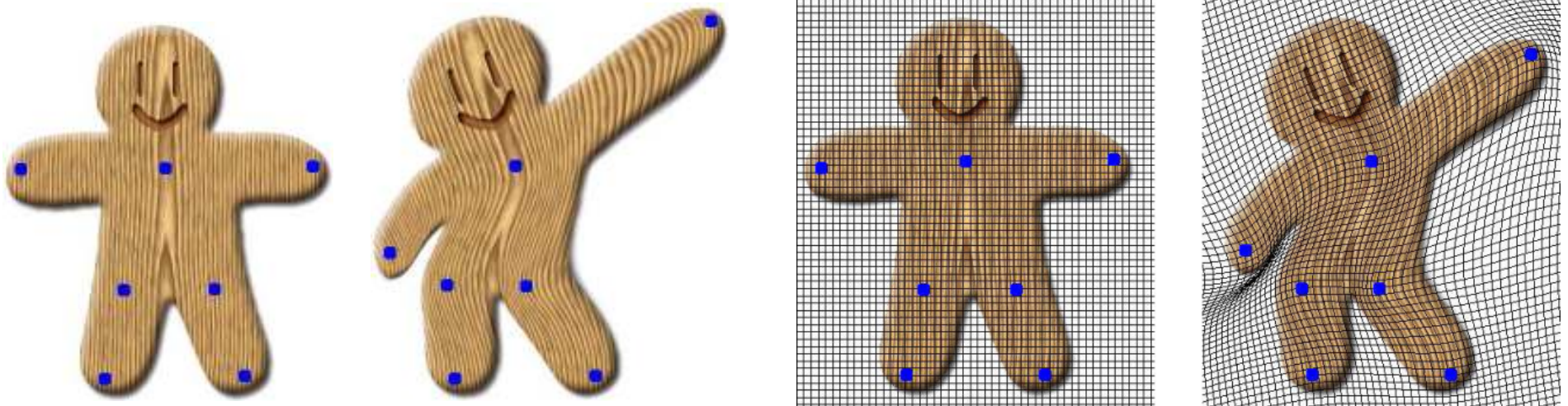




# As-Rigid-As-Possible Deformation

Moving-Least-Squares (MLS) approach [Schaefer et al. 2006]

- Points or segments as control objects
- First developed in 2D and later extended to 3D by Zhu and Gortler (2007)



# As-Rigid-As-Possible Deformation

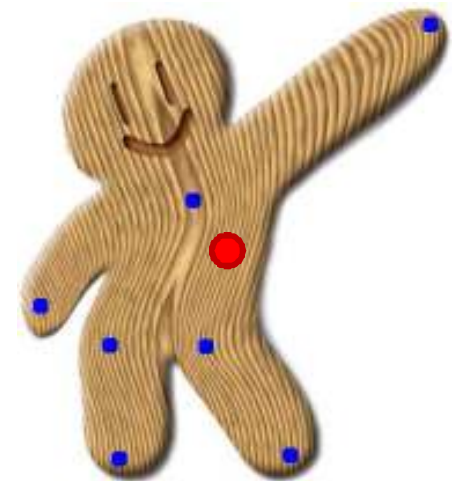
Moving-Least-Squares (MLS) approach [Schaefer et al. 2006]

- Attach an affine transformation to each point  $\mathbf{x} \in \mathbb{R}^3$ :

$$A_{\mathbf{x}}(\mathbf{p}) = M_{\mathbf{x}}\mathbf{p} + \mathbf{t}_{\mathbf{x}}$$

- The space warp:

$$\mathbf{x} \rightarrow A_{\mathbf{x}}(\mathbf{x})$$





# As-Rigid-As-Possible Deformation

Moving-Least-Squares (MLS) approach [Schaefer et al. 2006]

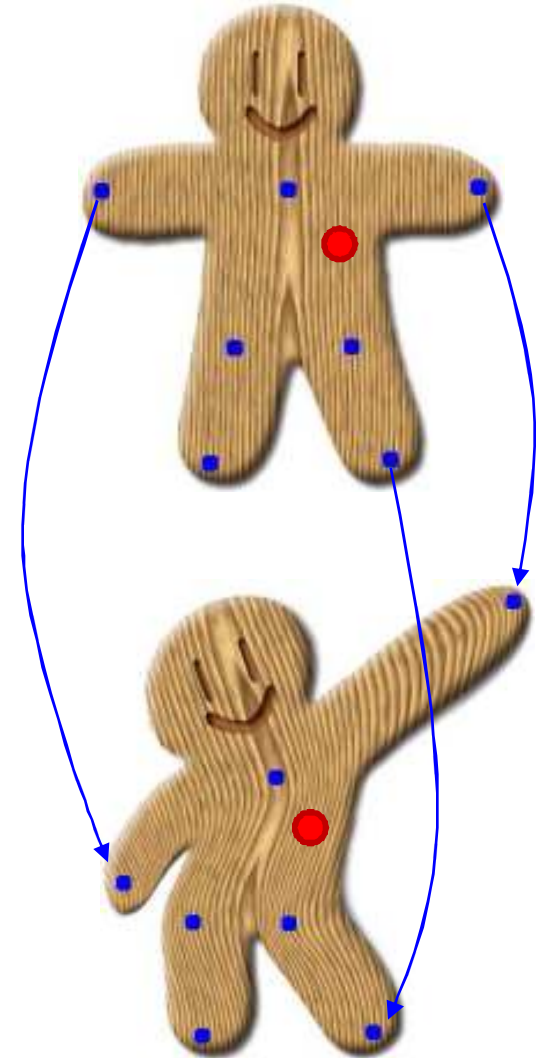
- Handles  $\mathbf{p}_i$  are displaced to  $\mathbf{q}_i$
- The local transformation at  $\mathbf{x}$ :

$$A_{\mathbf{x}}(\mathbf{p}) = M_{\mathbf{x}}\mathbf{p} + \mathbf{t}_{\mathbf{x}} \quad \text{s.t.}$$

$$\sum_{i=1}^k w_i(\mathbf{x}) \|A_{\mathbf{x}}(\mathbf{p}_i) - \mathbf{q}_i\|^2 \rightarrow \min$$

- The weights depend on  $\mathbf{x}$ :

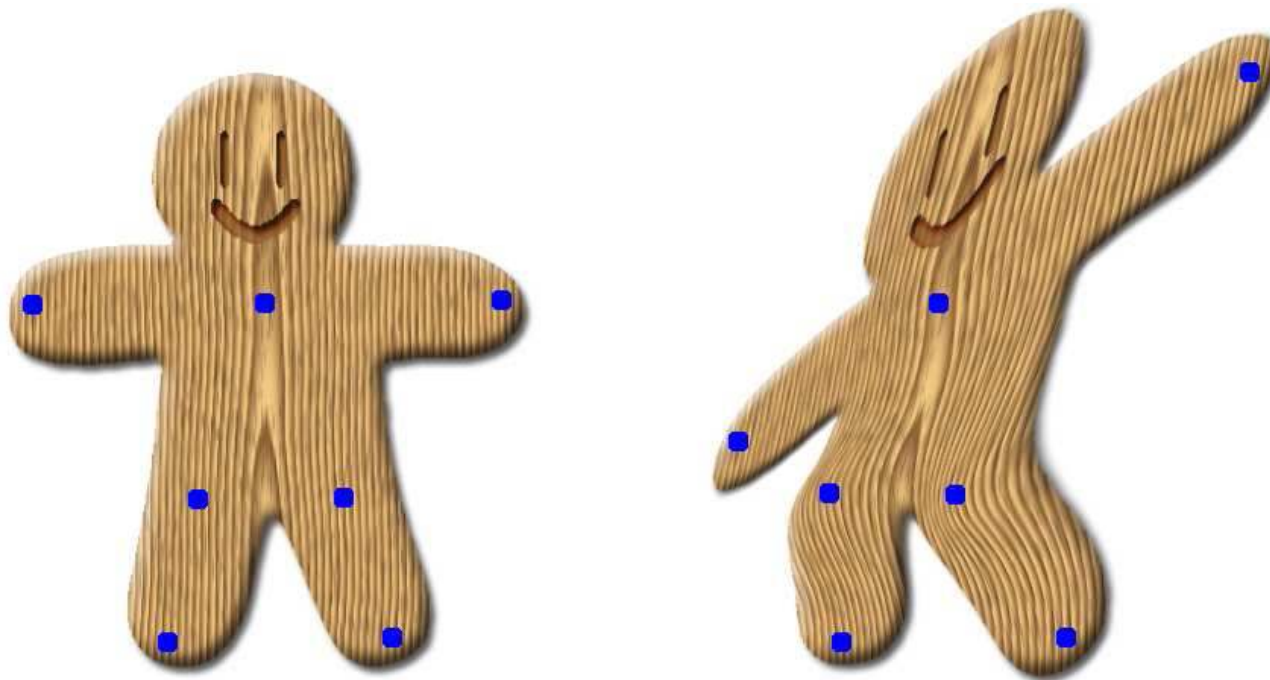
$$w_i(\mathbf{x}) = \|\mathbf{p}_i - \mathbf{x}\|^{-2\alpha}$$



# As-Rigid-As-Possible Deformation

Moving-Least-Squares (MLS) approach [Schaefer et al. 2006]

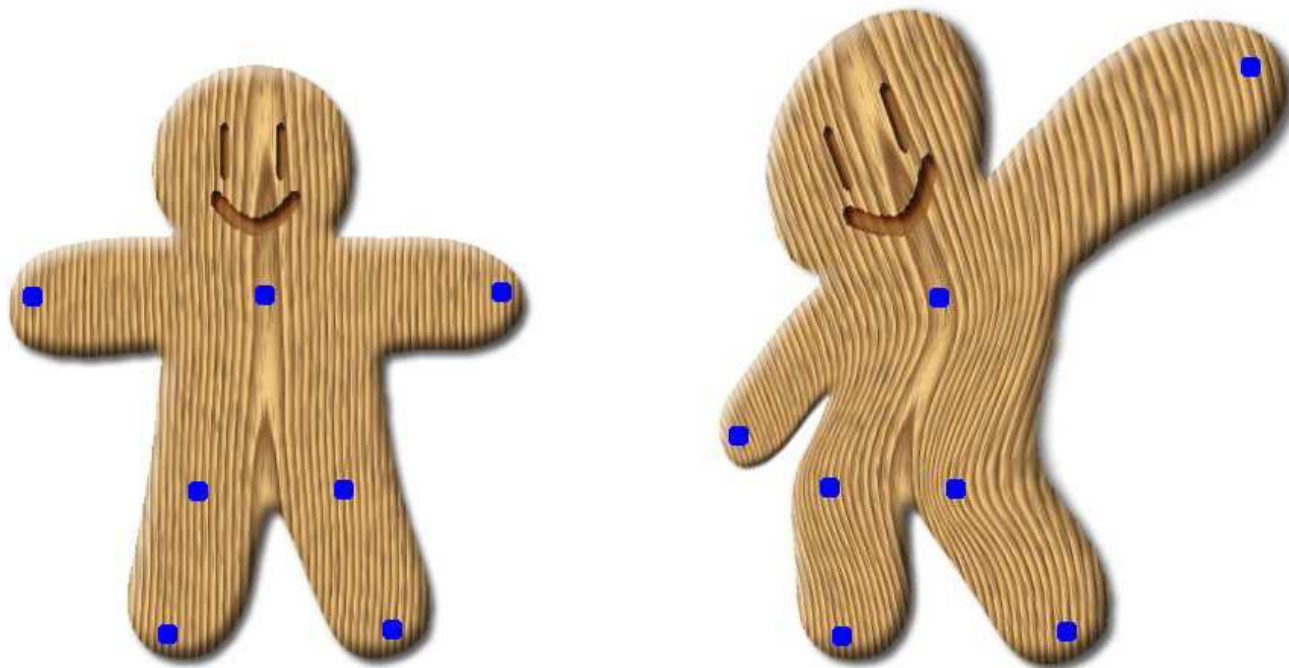
- No additional restriction on  $A_x(\cdot)$  – affine local transformations



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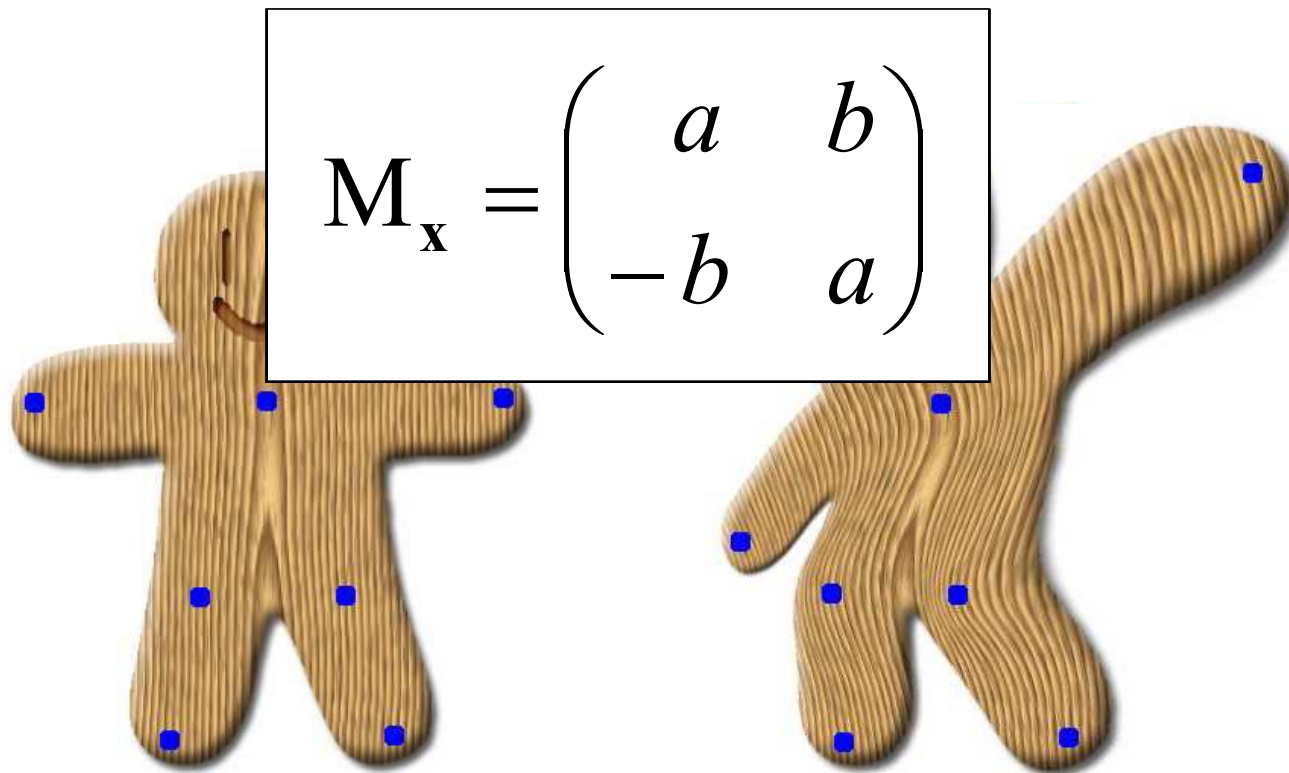
- Restrict  $A_x(\cdot)$  to similarity



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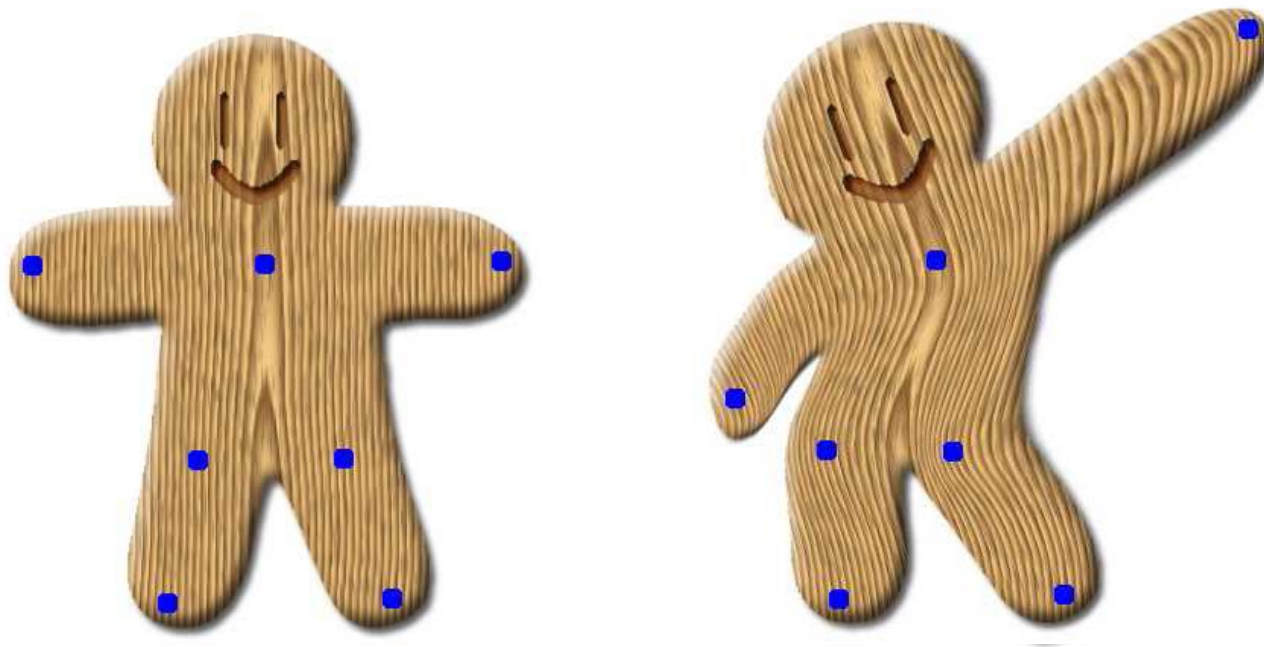
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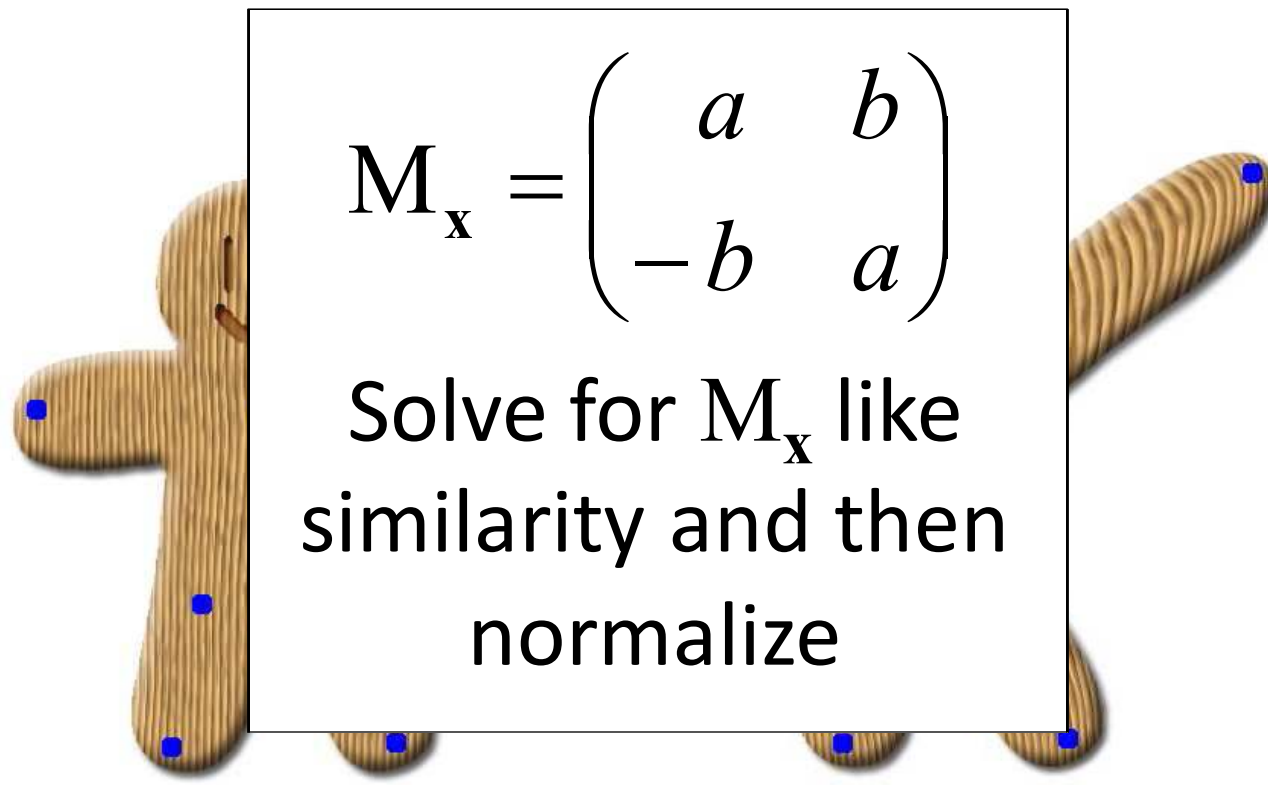
- Restrict  $A_x(\cdot)$  to rigid



# As-Rigid-As-Possible Deformation

Moving-Least-Squares (MLS) approach [Schaefer et al. 2006]

- Restrict  $A_x(\cdot)$  to rigid


$$M_x = \begin{pmatrix} a & b \\ -b & a \end{pmatrix}$$

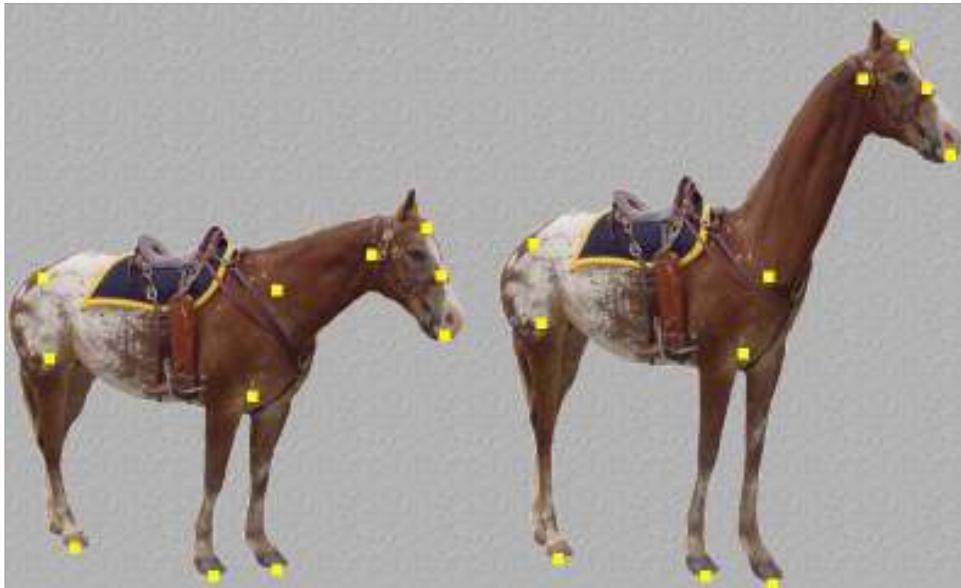
Solve for  $M_x$  like  
similarity and then  
normalize



# As-Rigid-As-Possible Deformation

Moving-Least-Squares (MLS) approach [Schaefer et al. 2006]

- Examples



# As-Rigid-As-Possible Deformation

MLS approach – extension to 3D [Zhu & Gortler 2007]

- No linear expression for similarity in 3D
- Instead, can solve for the minimizing rotation

$$\arg \min_{\mathbf{R} \in \text{SO}(3)} \sum_{i=1}^k w_i(\mathbf{x}) \|\mathbf{R}\mathbf{p}_i - \mathbf{q}_i\|^2$$

by polar decomposition of the  $3 \times 3$  covariance matrix

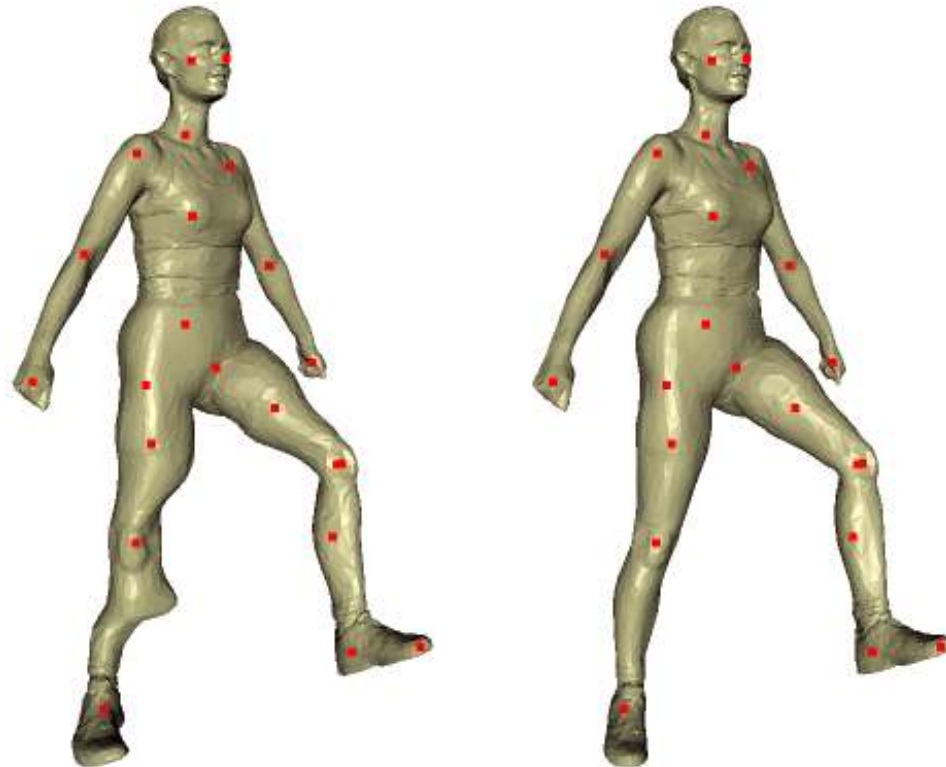


# As-Rigid-As-Possible Deformation

MLS approach – extension to 3D [Zhu & Gortler 2007]

- Zhu and Gortler also replace the Euclidean distance in the weights by “distance within the shape”

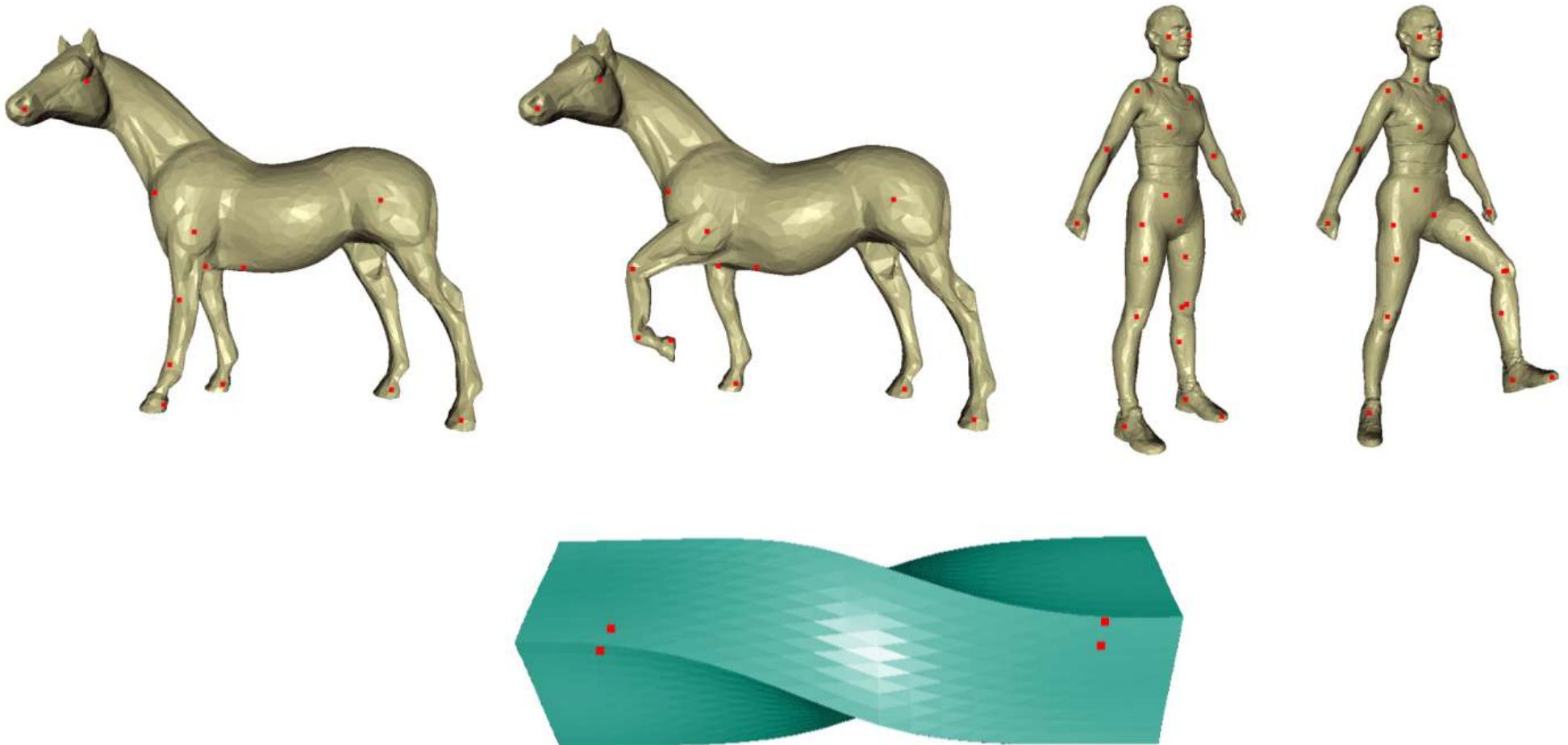
$$w_i(\mathbf{x}) = d(\mathbf{p}_i, \mathbf{x})^{-2\alpha}$$



# As-Rigid-As-Possible Deformation

MLS approach – extension to 3D [Zhu & Gortler 2007]

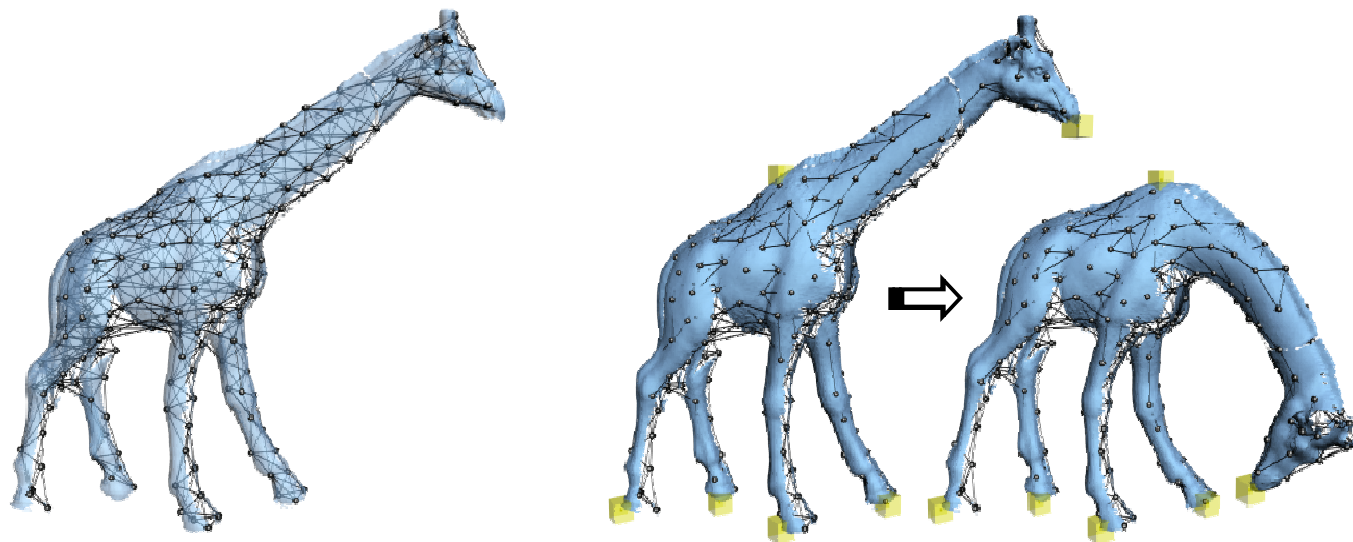
- More results



# As-Rigid-As-Possible Deformation

Deformation Graph approach [Sumner et al. 2007]

- Surface handles as interface
- Underlying graph to represent the deformation; nodes store rigid transformations
- Decoupling of handles from def. representation

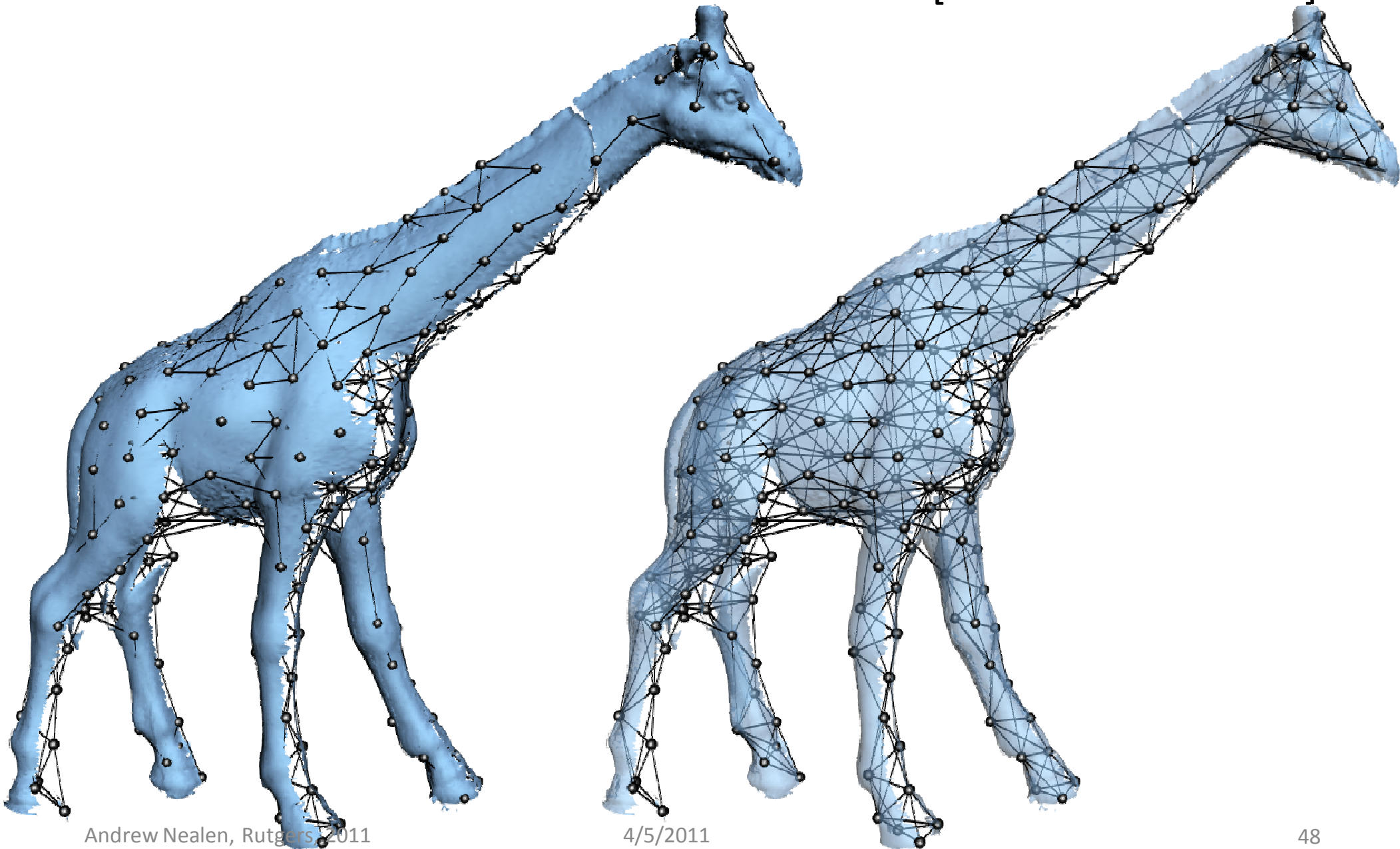


Deformation Graph

Optimization Procedure

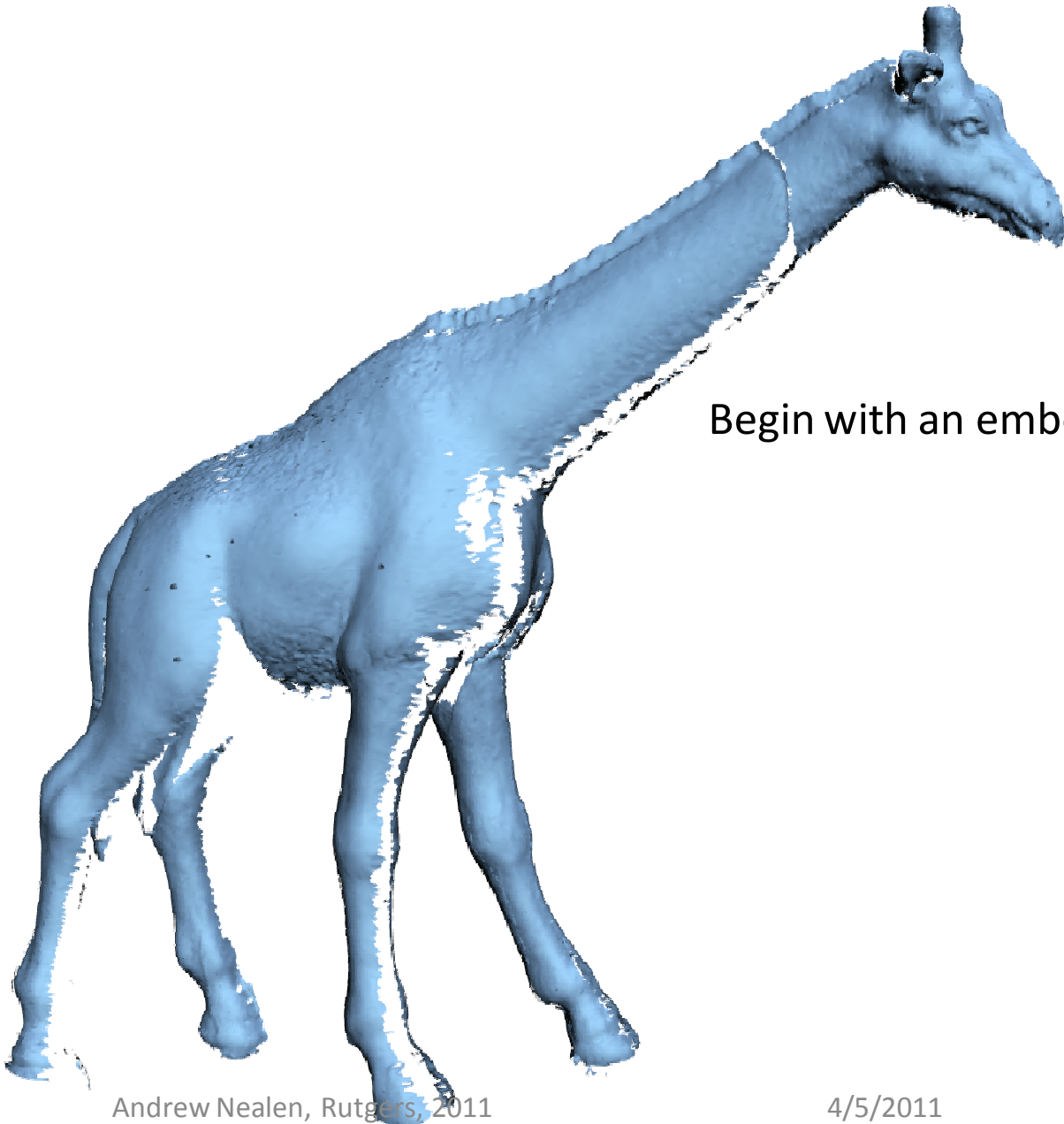
# Deformation Graph

[Sumner et al. 2007]



# Deformation Graph

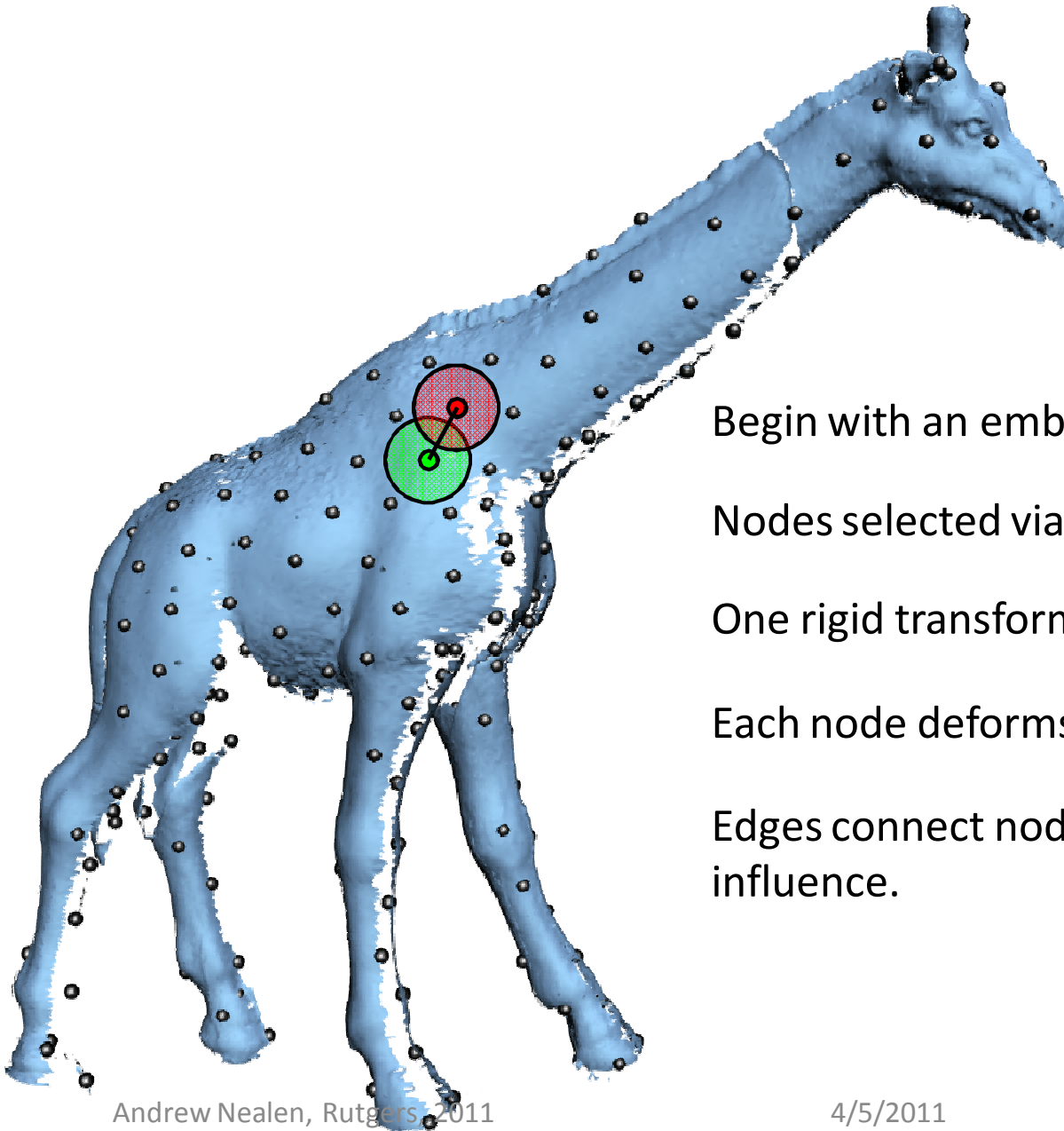
[Sumner et al. 2007]



Begin with an embedded object.

# Deformation Graph

[Sumner et al. 2007]



Begin with an embedded object.

Nodes selected via uniform sampling; located at  $\mathbf{g}_j$

One rigid transformation for each node:  $\mathbf{R}_j, \mathbf{t}_j$

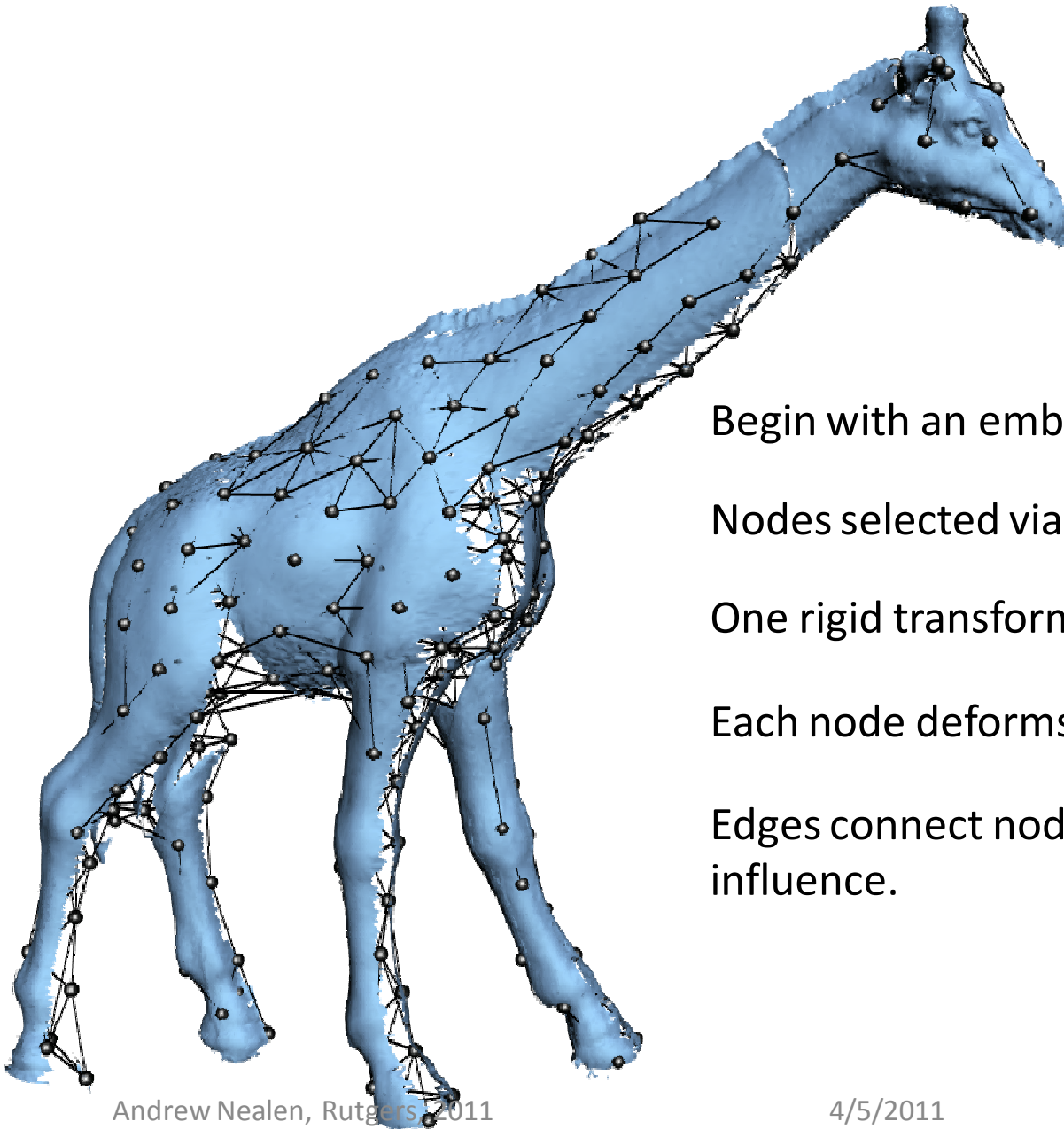
Each node deforms nearby space.

Edges connect nodes of overlapping influence.



# Deformation Graph

[Sumner et al. 2007]



Begin with an embedded object.

Nodes selected via uniform sampling; located at  $\mathbf{g}_j$

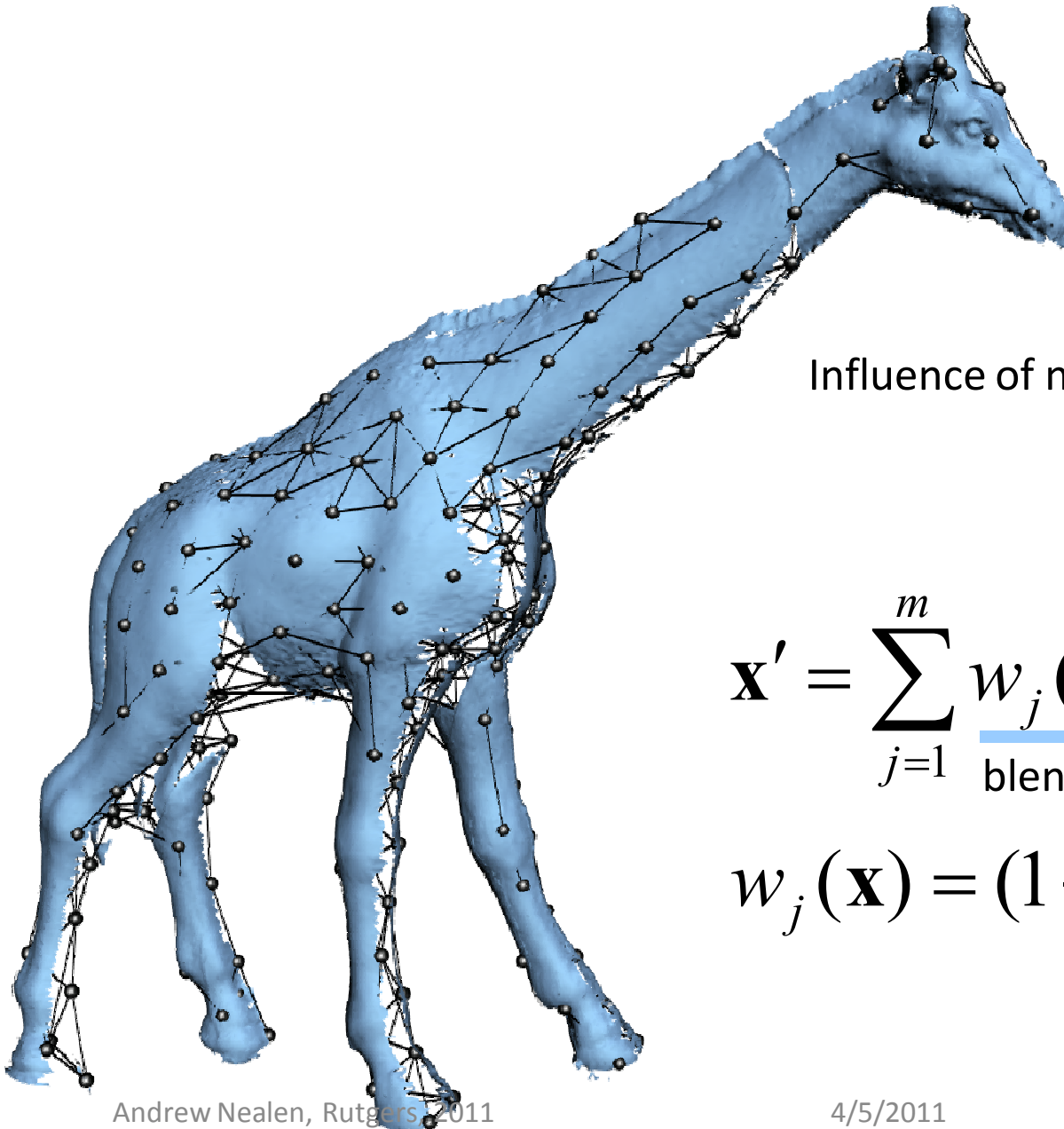
One rigid transformation for each node:  $\mathbf{R}_j, \mathbf{t}_j$

Each node deforms nearby space.

Edges connect nodes of overlapping influence.

# Deformation Graph

[Sumner et al. 2007]



Influence of nearby transformations is blended.

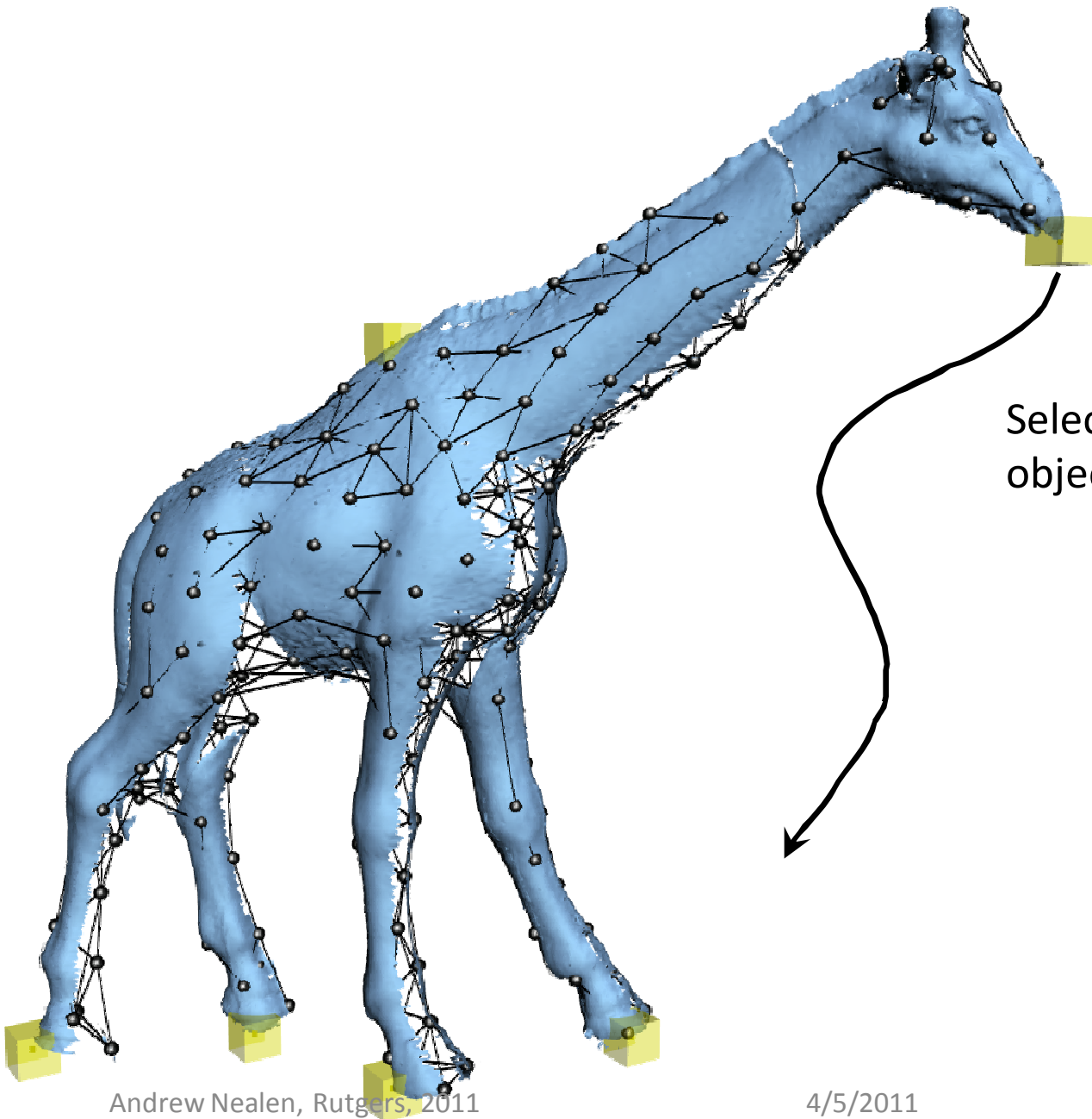
$$\mathbf{x}' = \sum_{j=1}^m \underbrace{w_j(\mathbf{x})}_{\text{blending weights}} \left[ \underbrace{R_j(\mathbf{x} - \mathbf{g}_j) + \mathbf{g}_j + \mathbf{t}_j}_{\text{point } \mathbf{x} \text{ transformed by node } j} \right]$$

$$w_j(\mathbf{x}) = \left( 1 - \frac{\|\mathbf{x} - \mathbf{g}_j\|}{d_{\max}} \right)^2$$



# Optimization

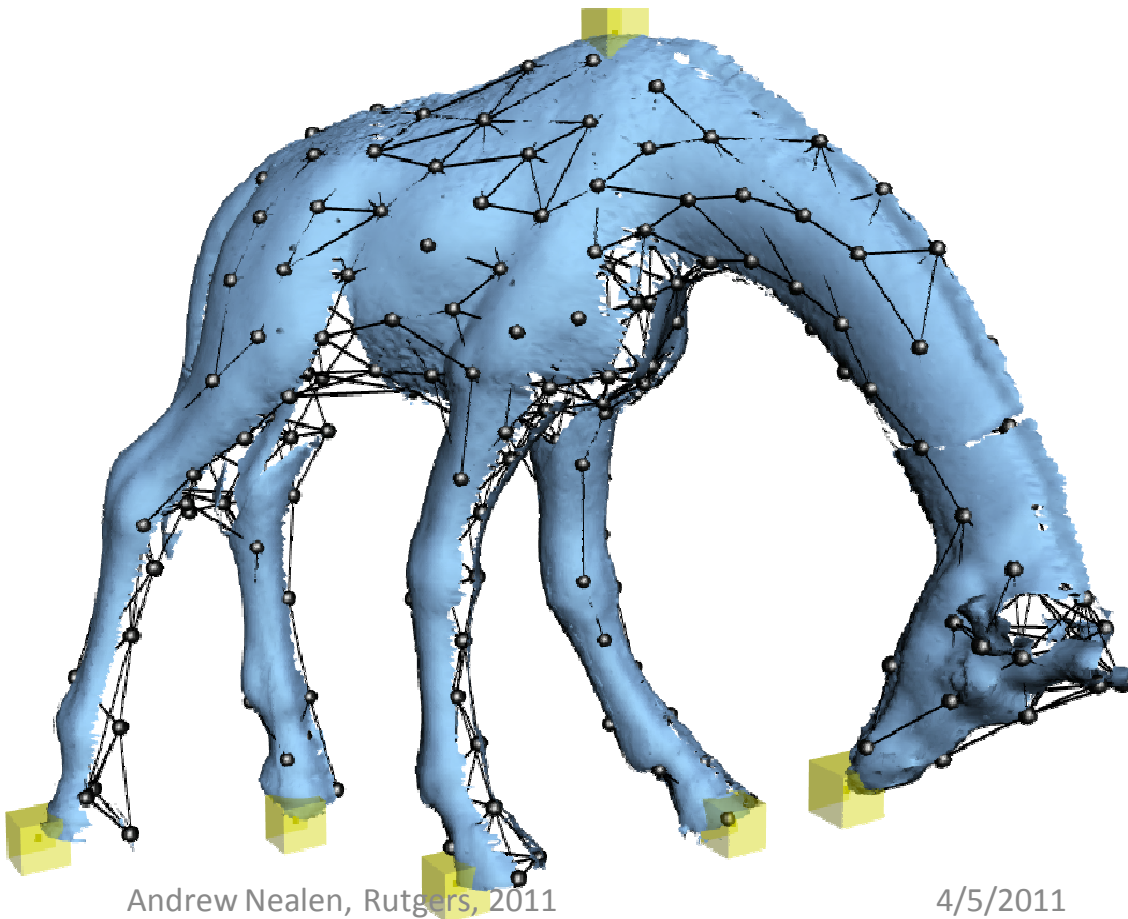
[Sumner et al. 2007]



Select & drag vertices of embedded object.

# Optimization

[Sumner et al. 2007]



Select & drag vertices of embedded object.

Optimization finds

deformation parameters  $\mathbf{R}_j$ ,  $\mathbf{t}_j$ .

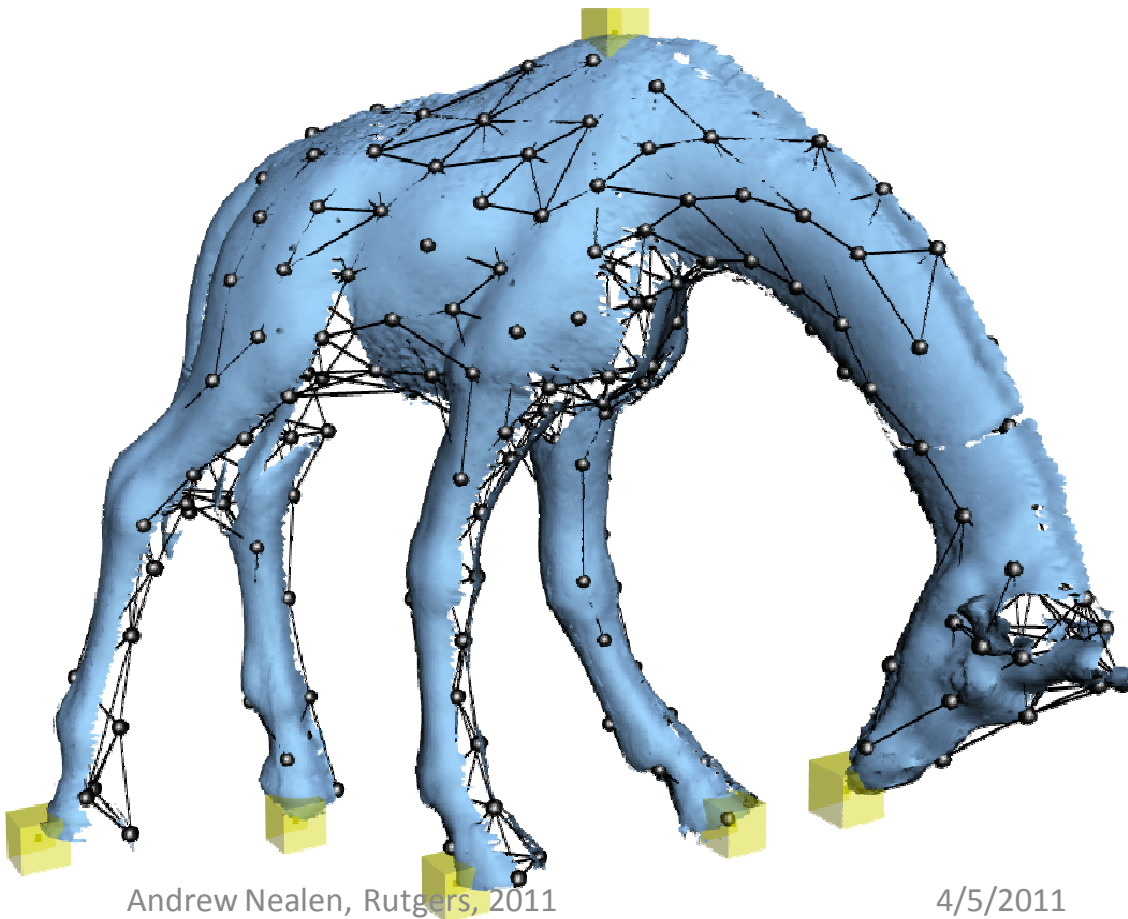
$$\min_{\mathbf{R}_1, \mathbf{t}_1, \dots, \mathbf{R}_m, \mathbf{t}_m} \quad \underbrace{w_{\text{rot}} \mathbf{E}_{\text{rot}}}_{\text{Rotation term}} + \underbrace{w_{\text{reg}} \mathbf{E}_{\text{reg}}}_{\text{Regularization term}} + \underbrace{w_{\text{con}} \mathbf{E}_{\text{con}}}_{\text{Constraint term}}$$

Graph  
parameters

Rotation  
term

Regularization  
term

Constraint  
term

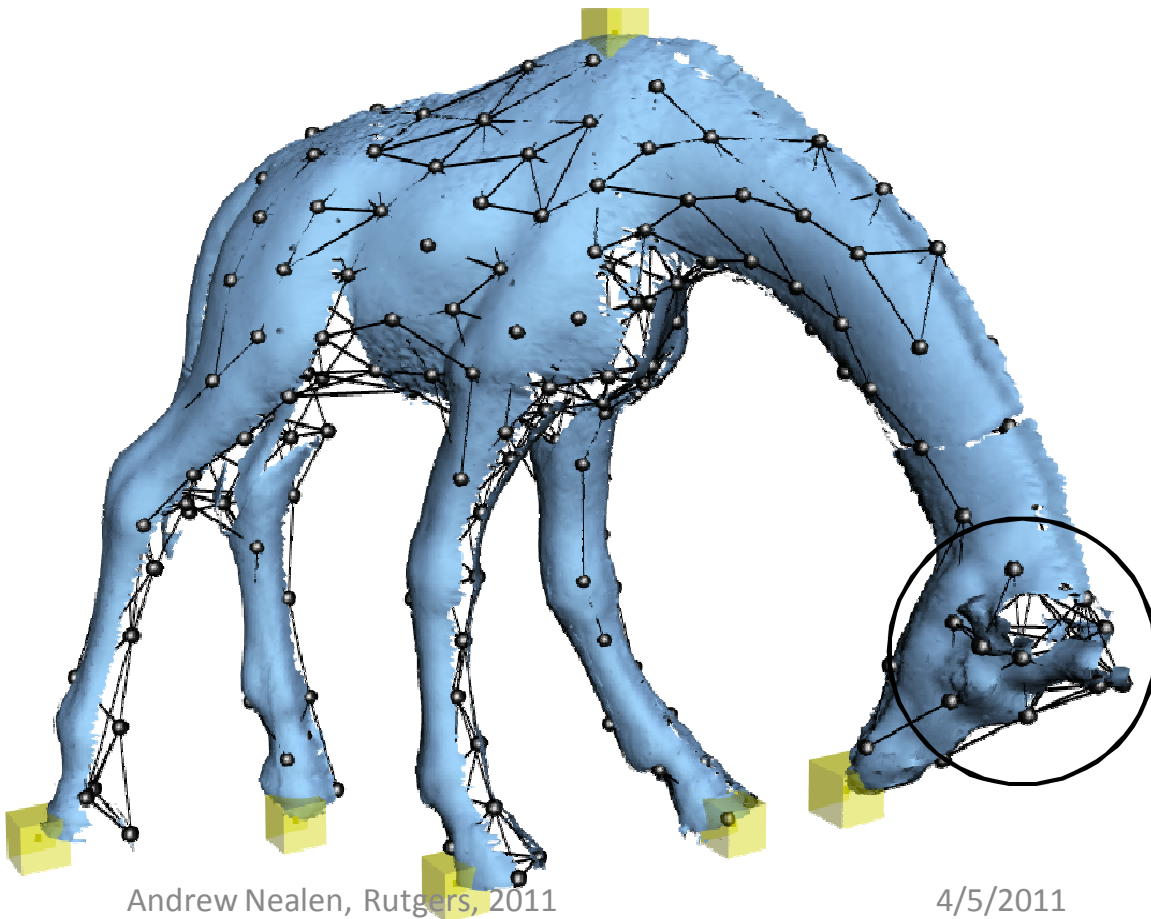


Select & drag vertices of embedded object.

Optimization finds deformation parameters  $\mathbf{R}_j, \mathbf{t}_j$ .

$$\min_{\mathbf{R}_1, \mathbf{t}_1, \dots, \mathbf{R}_m, \mathbf{t}_m} \quad \underline{w_{\text{rot}} E_{\text{rot}} + w_{\text{reg}} E_{\text{reg}} + w_{\text{con}} E_{\text{con}}}$$

$$\text{Rot}(\mathbf{R}) = (\mathbf{c}_1 \cdot \mathbf{c}_2)^2 + (\mathbf{c}_1 \cdot \mathbf{c}_3)^2 + (\mathbf{c}_2 \cdot \mathbf{c}_3)^2 + (\mathbf{c}_1 \cdot \mathbf{c}_1 - 1)^2 + (\mathbf{c}_2 \cdot \mathbf{c}_2 - 1)^2 + (\mathbf{c}_3 \cdot \mathbf{c}_3 - 1)^2$$



$$E_{\text{rot}} = \sum_{j=1}^m \text{Rot}(\mathbf{R}_j)$$

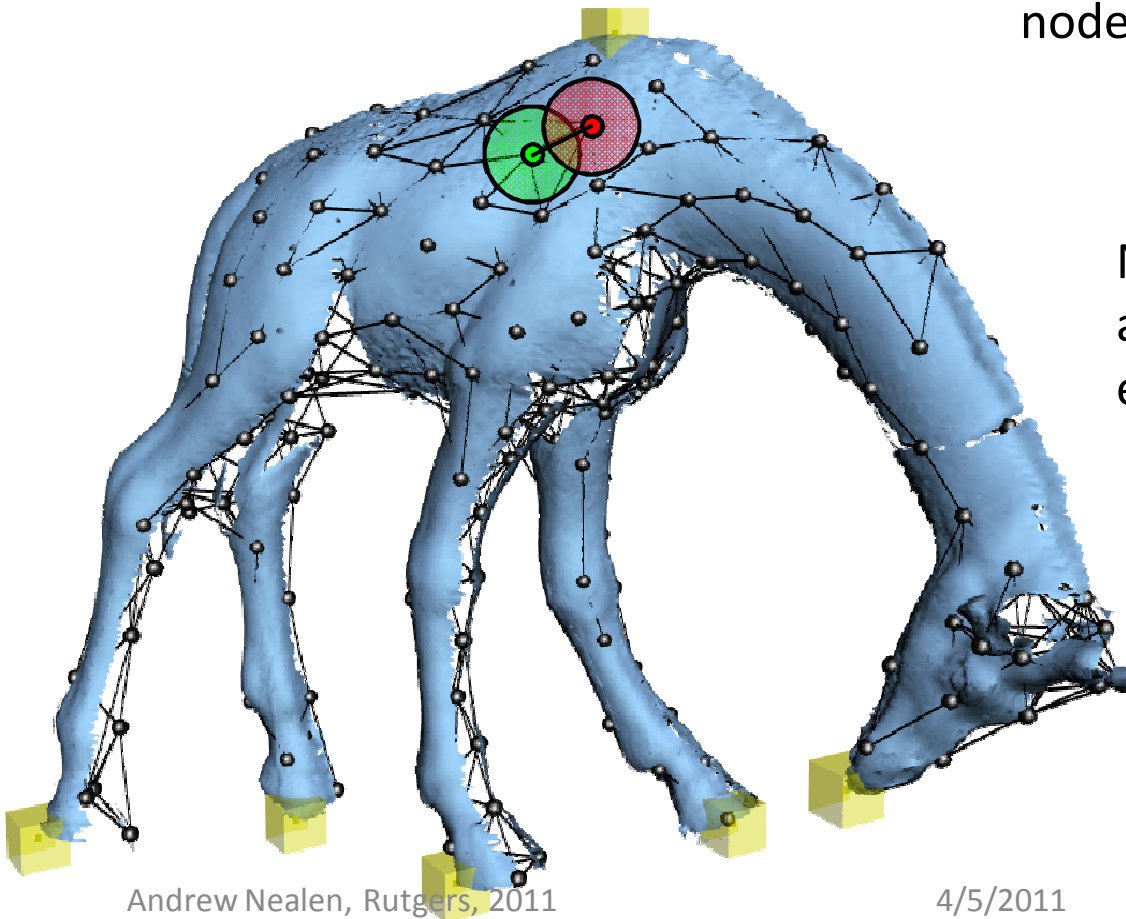
For detail preservation, features should rotate and not scale or skew.

$$\min_{\mathbf{R}_1, \mathbf{t}_1, \dots, \mathbf{R}_m, \mathbf{t}_m} w_{\text{rot}} E_{\text{rot}} + w_{\text{reg}} E_{\text{reg}} + w_{\text{con}} E_{\text{con}}$$

$$E_{\text{reg}} = \sum_{j=1}^m \sum_{k \in N(j)} \alpha_{jk} \left\| \mathbf{R}_j (\mathbf{g}_k - \mathbf{g}_j) + \mathbf{g}_j + \mathbf{t}_j - (\mathbf{g}_k + \mathbf{t}_k) \right\|_2^2$$

where node  $j$  thinks  
node  $k$  should go

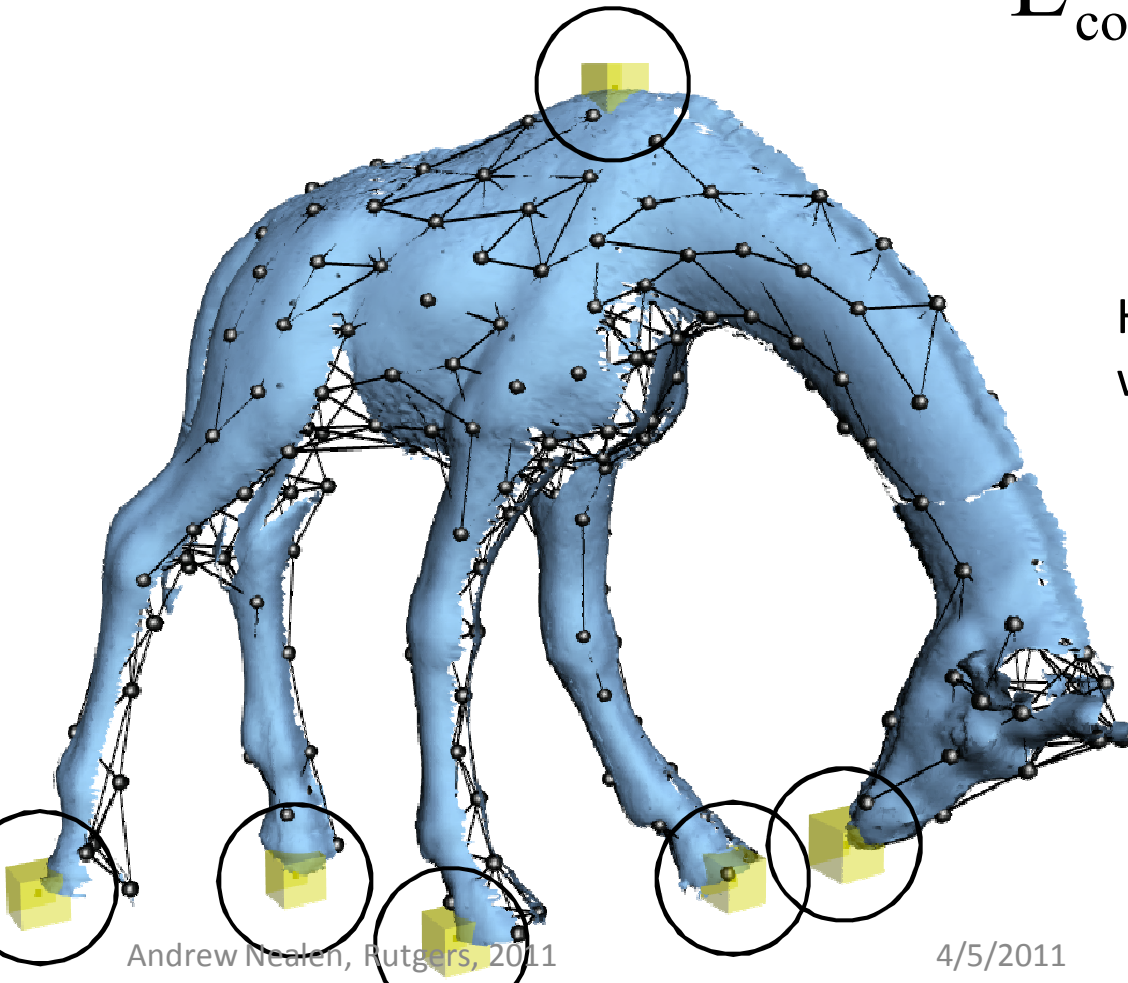
where node  $k$   
actually goes



Neighboring nodes should  
agree on where they transform  
each other.

$$\min_{\mathbf{R}_1, \mathbf{t}_1, \dots, \mathbf{R}_m, \mathbf{t}_m} w_{\text{rot}} \mathbf{E}_{\text{rot}} + w_{\text{reg}} \mathbf{E}_{\text{reg}} + \underline{w_{\text{con}} \mathbf{E}_{\text{con}}}$$

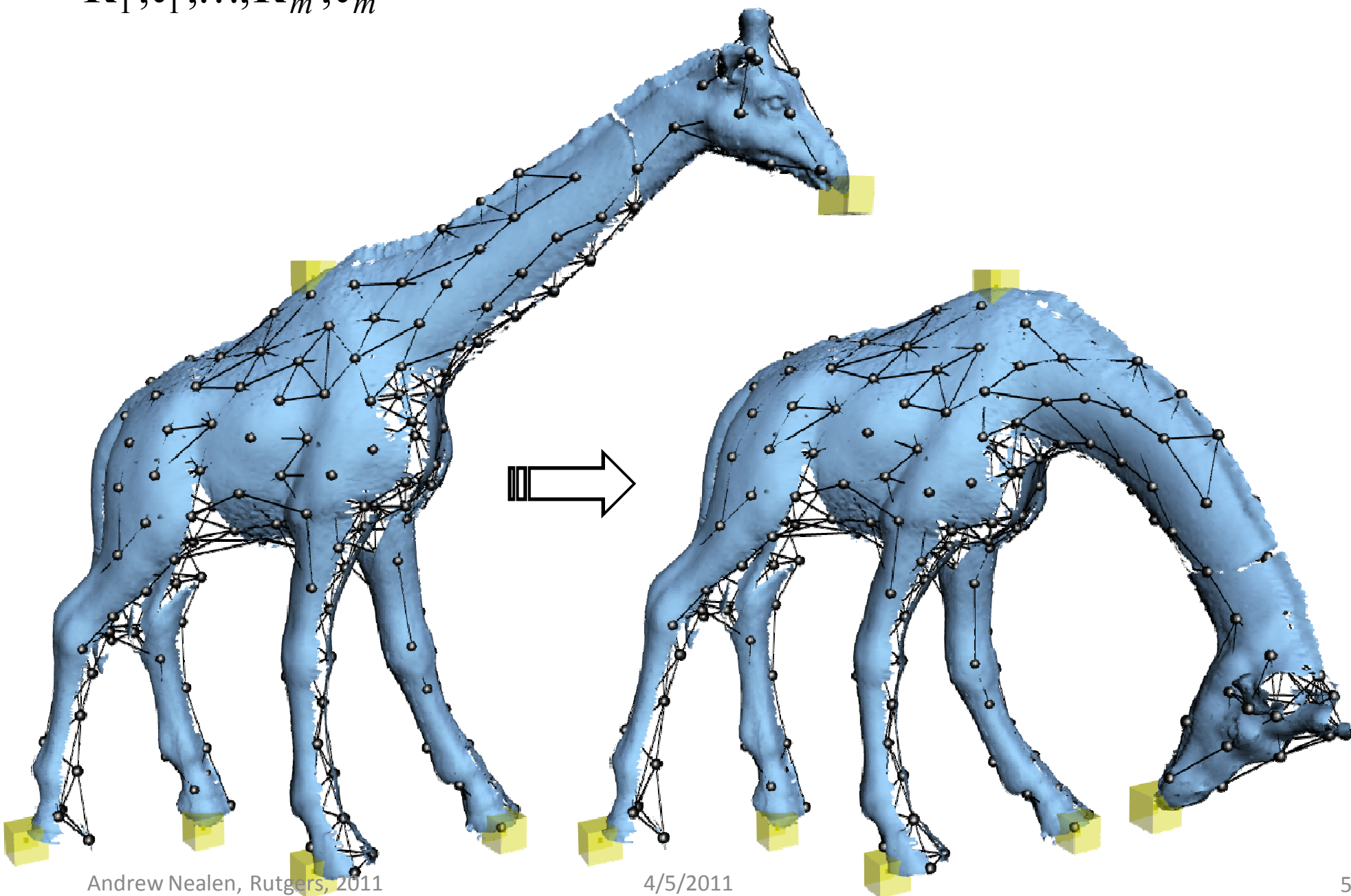
$$\mathbf{E}_{\text{con}} = \sum_{l=1}^p \left\| \tilde{\mathbf{v}}_{\text{index}(l)} - \mathbf{q}_l \right\|_2^2$$



Handle vertices should go where the user puts them.



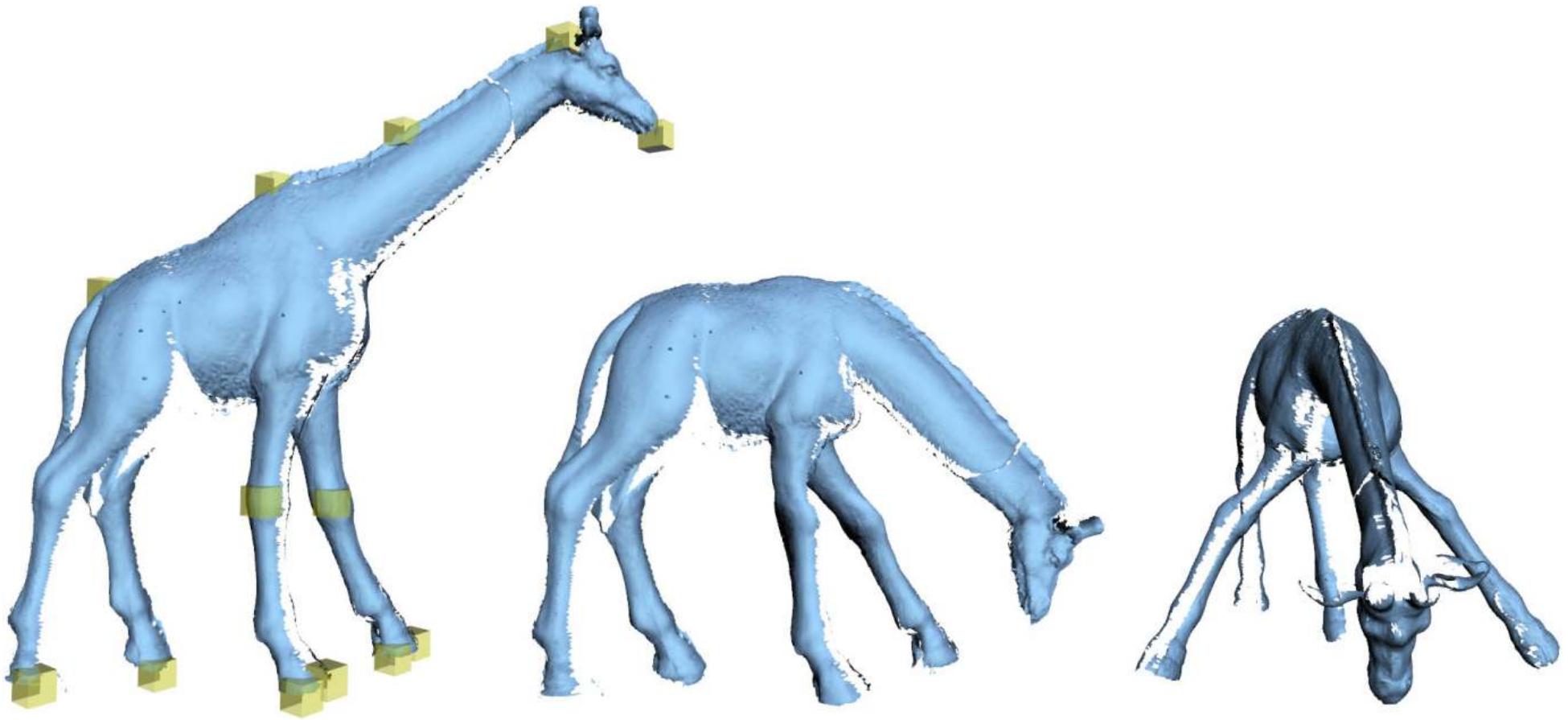
$$\min_{\mathbf{R}_1, \mathbf{t}_1, \dots, \mathbf{R}_m, \mathbf{t}_m} w_{\text{rot}} E_{\text{rot}} + w_{\text{reg}} E_{\text{reg}} + w_{\text{con}} E_{\text{con}}$$





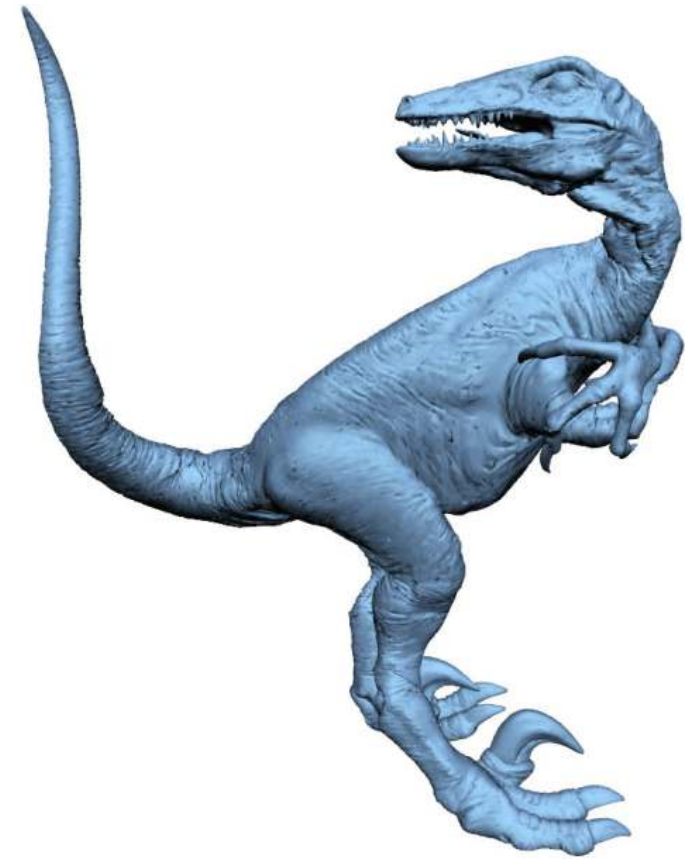
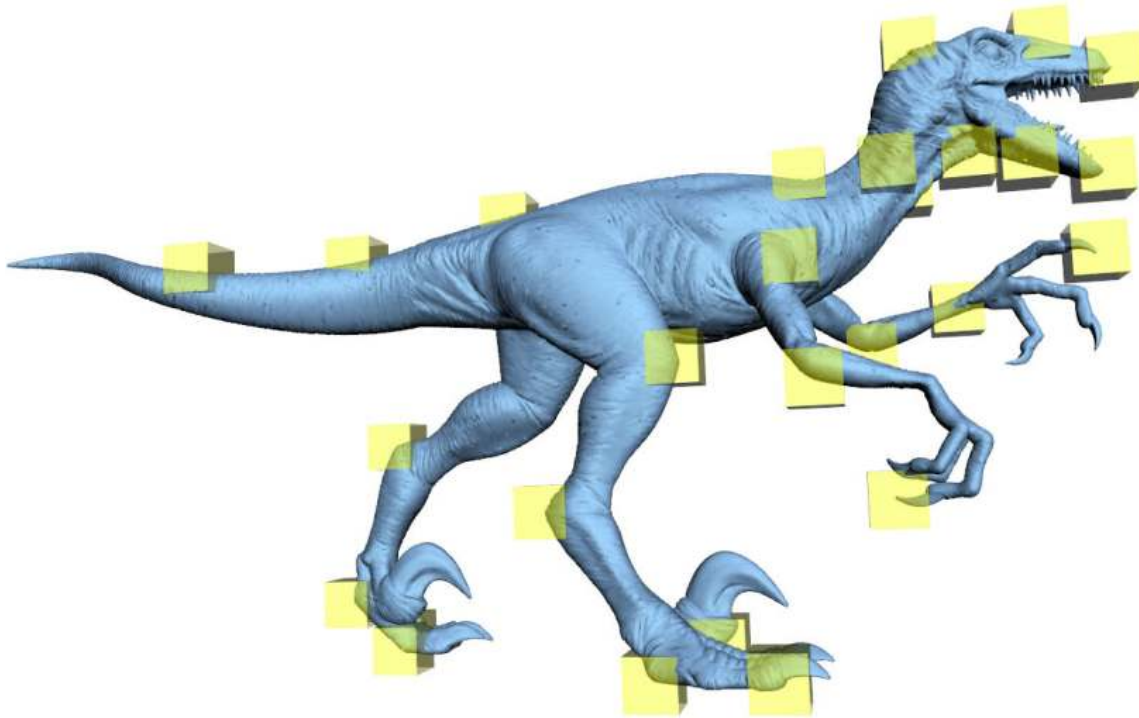
# Results: Polygon Soup

[Sumner et al. 2007]



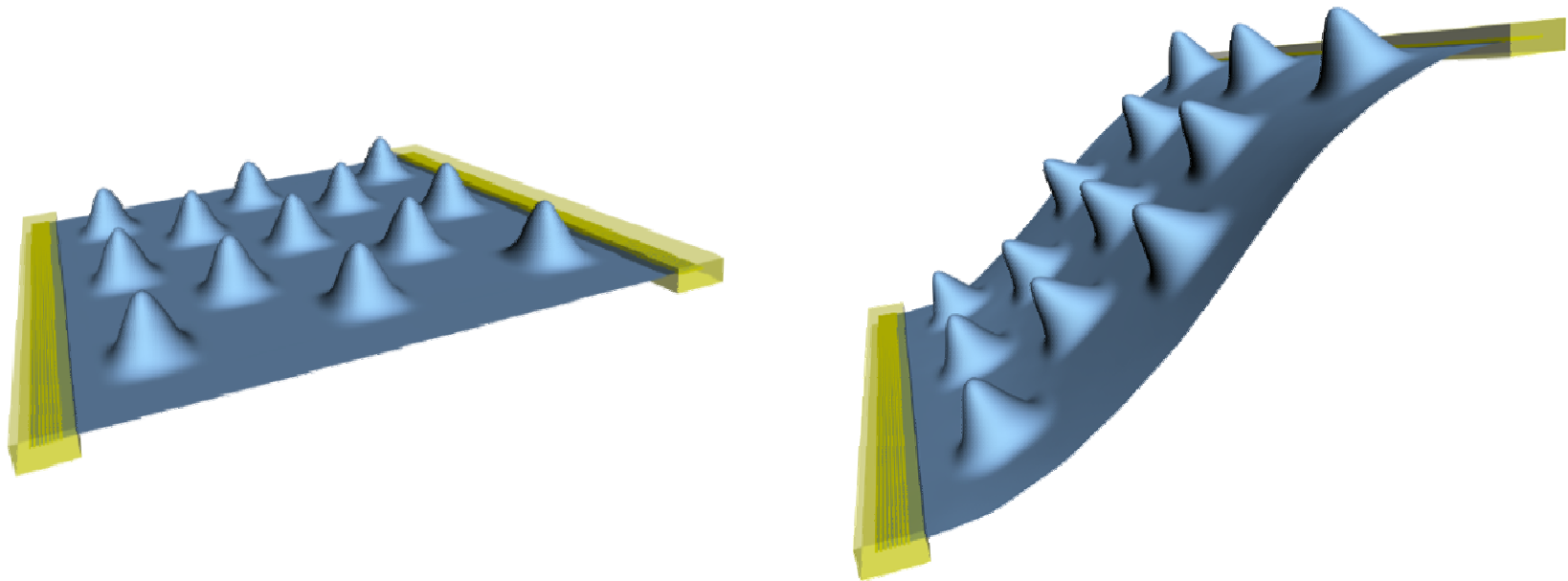
# Results: Giant Mesh

[Sumner et al. 2007]



# Results: Detail Preservation

[Sumner et al. 2007]



Demo

# Discussion

- Decoupling of deformation complexity and model complexity
- Nonlinear energy optimization – results comparable to surface-based approaches

