

CS 523: Computer Graphics, Spring 2009

Interactive Shape Modeling

Space deformations

Space Deformation

- Displacement function defined on the ambient space

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

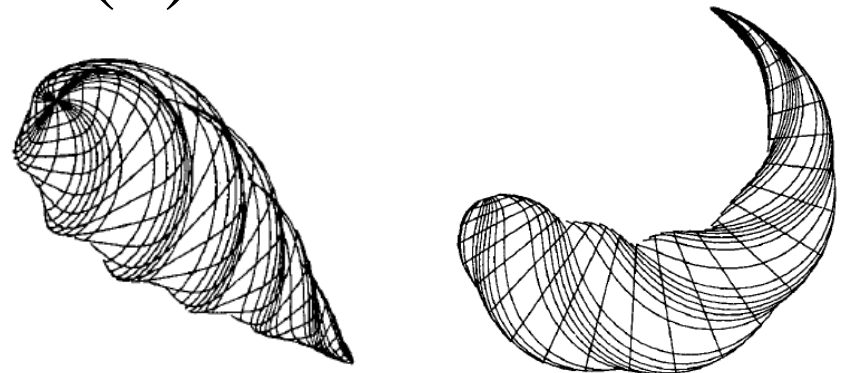
- Evaluate the function on the points of the shape embedded in the space

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

Twist warp

Global and local deformation of solids

[A. Barr, SIGGRAPH 84]

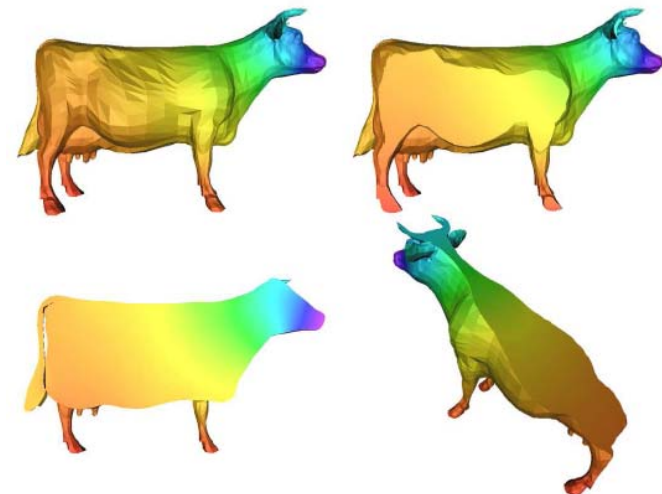


Freeform Deformations

- 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 $B_i(\mathbf{x}) : \mathbb{R}^3 \rightarrow \mathbb{R}$

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

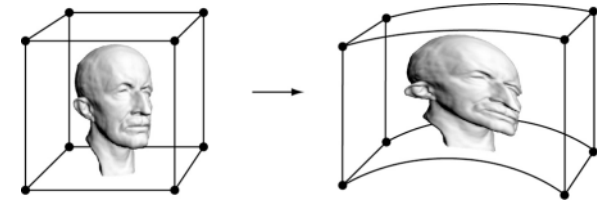
- Basis functions should be smooth for aesthetic results



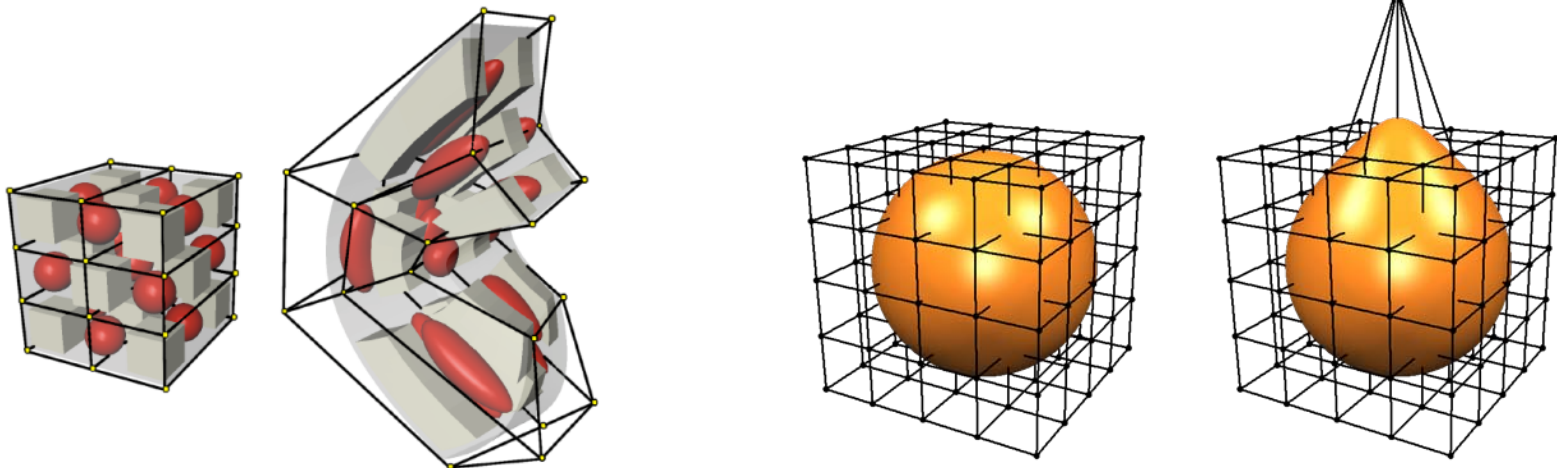
Trivariate Tensor Product Bases

[Sederberg and Parry 86]

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

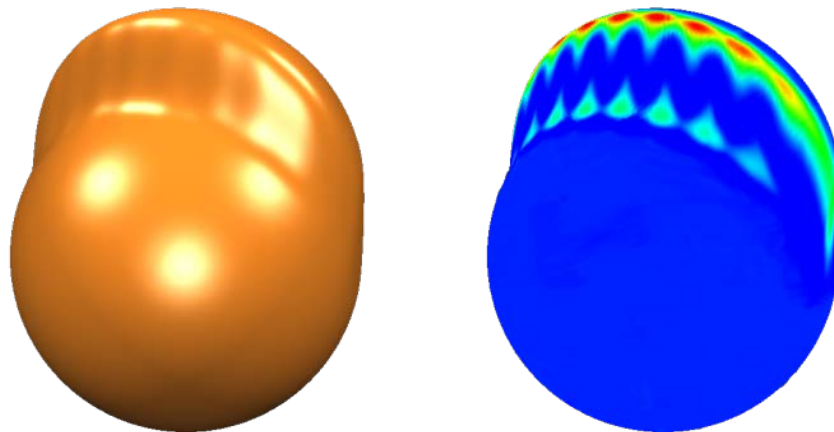


$$\mathbf{d}(x, y, z) = \sum_{i=0}^l \sum_{j=0}^m \sum_{k=0}^n \mathbf{d}_{ijk} N_i(x) N_j(y) N_k(z)$$



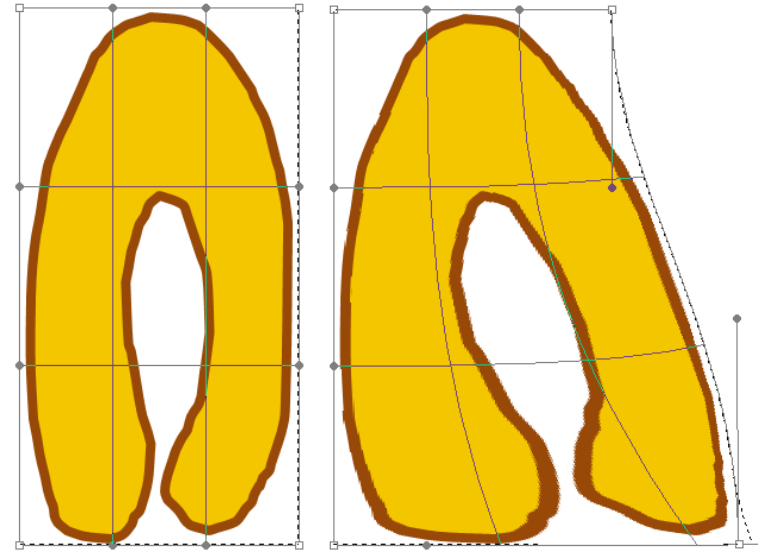
Trivariate Tensor Product Bases

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



Lattice as Control Object

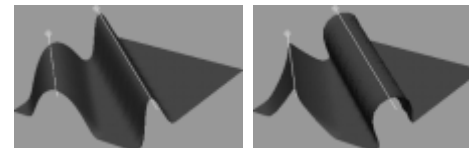
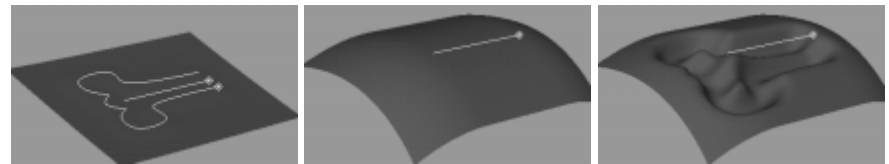
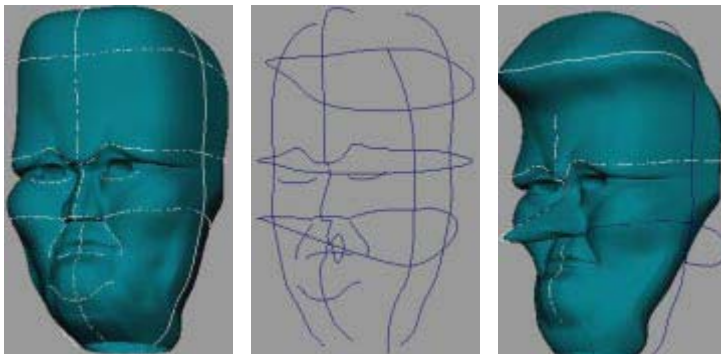
- Difficult to manipulate
- The control object is not related to the shape of the edited object
- Part of the shape in close Euclidean distance always deform similarly, even if geodesically far



Wires

[Singh and Fiume 98]

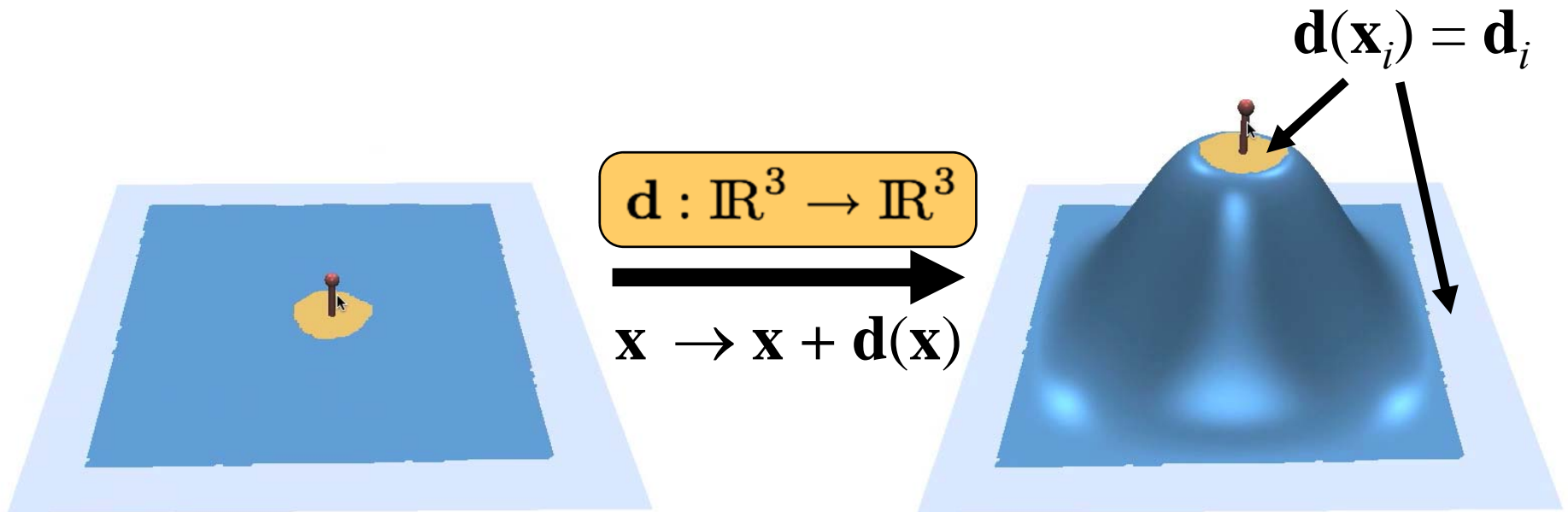
- Control objects are arbitrary space curves
- Can place curves along meaningful features of the edited object
- Smooth deformations around the curve with decreasing influence



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

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

- 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

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

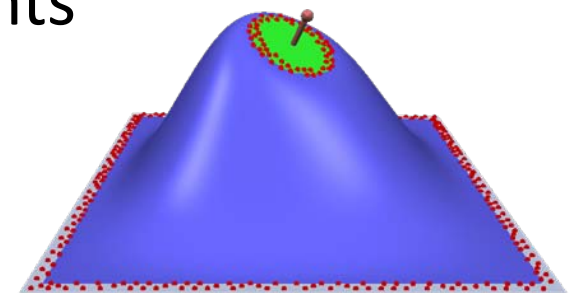
RBF Fitting

[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}



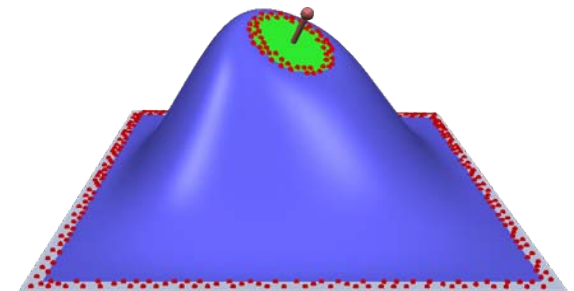
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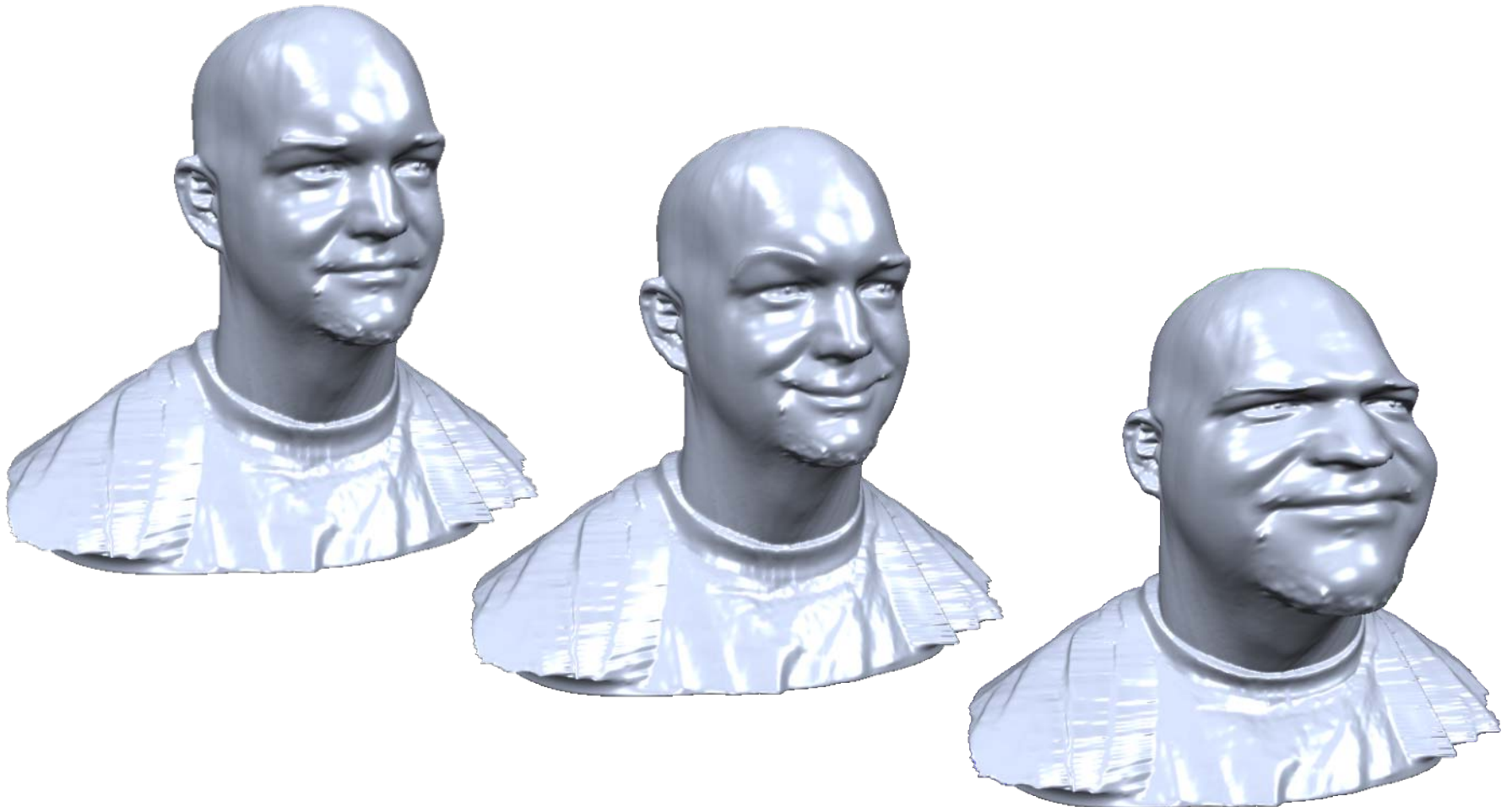
$$\mathbf{d}(\mathbf{x}) = \sum_j \mathbf{w}_j \cdot \varphi(\|\mathbf{c}_j - \mathbf{x}\|) + \mathbf{p}(\mathbf{x})$$

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



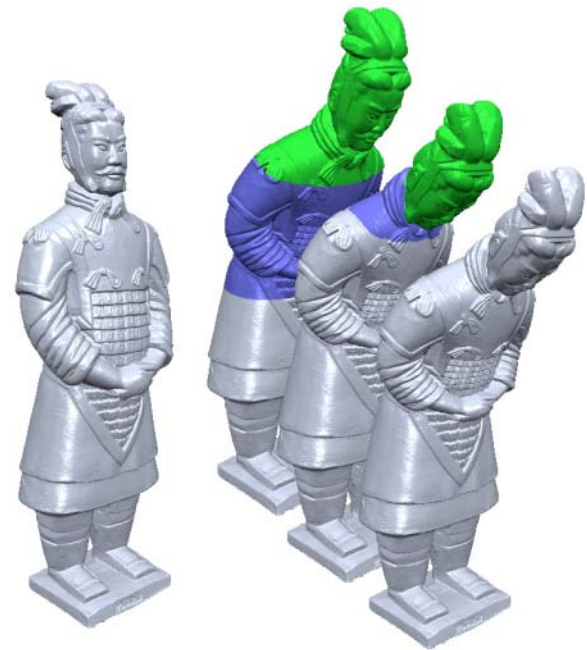
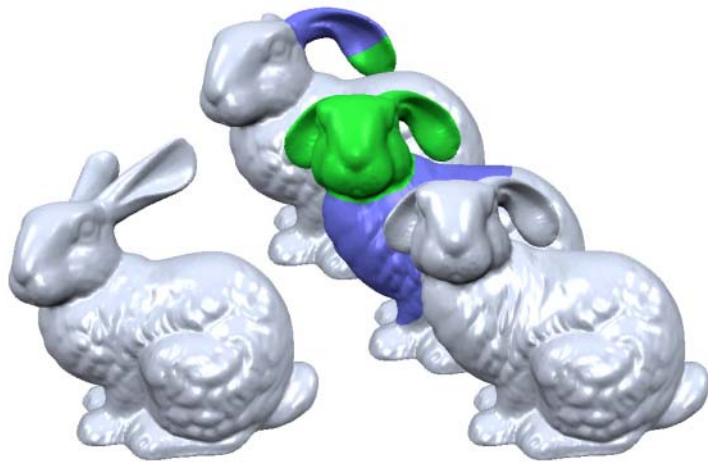
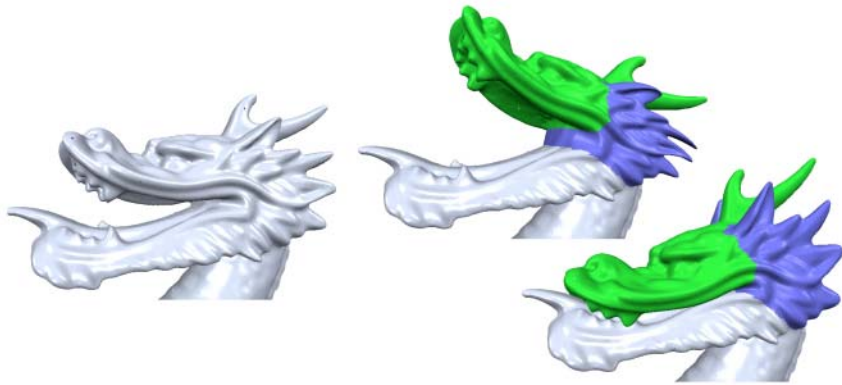
Local & Global Deformations

[RBF, Botsch and Kobbelt 05]



Local & Global Deformations

[RBF, Botsch and Kobbelt 05]

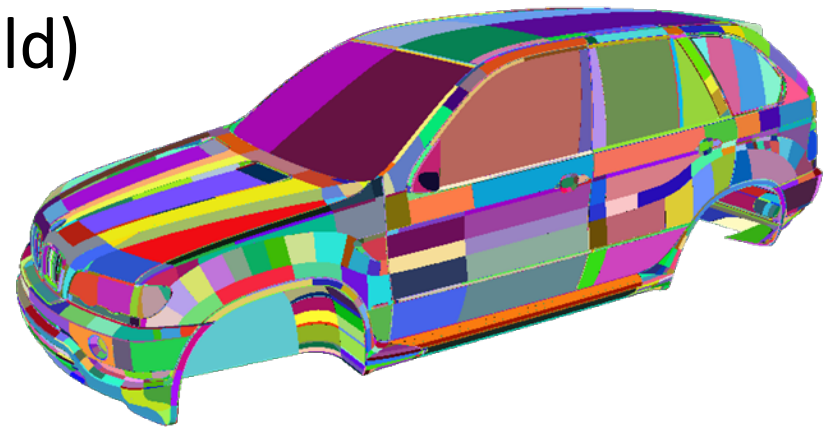


1M vertices
movie

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
 - ...



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

then the Jacobian is the determinant

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

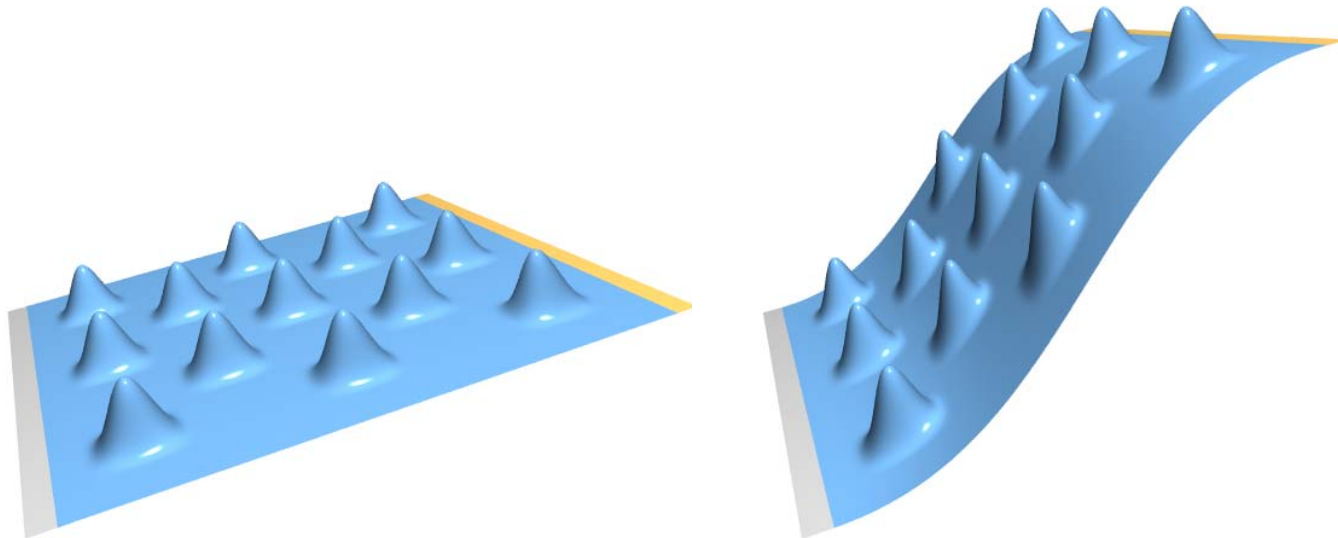
- Easier to analyze: functions on Euclidean domain

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

Space Deformations

Summary so far

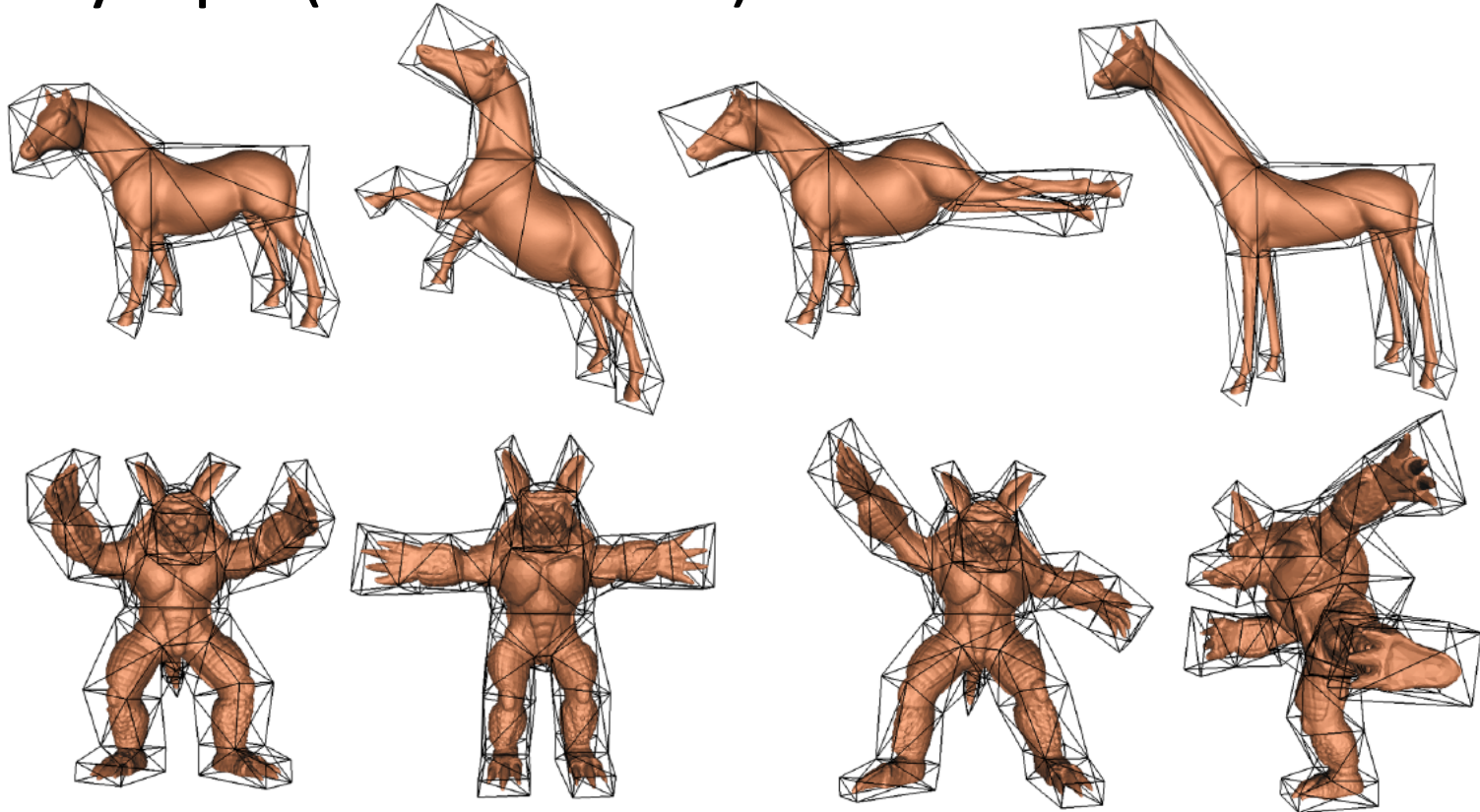
- The deformation is only loosely aware of the shape that is being edited
- Small Euclidean distance \rightarrow similar deformation
- Local surface detail may be distorted



Cage-based Deformations

[Ju et al. 2005]

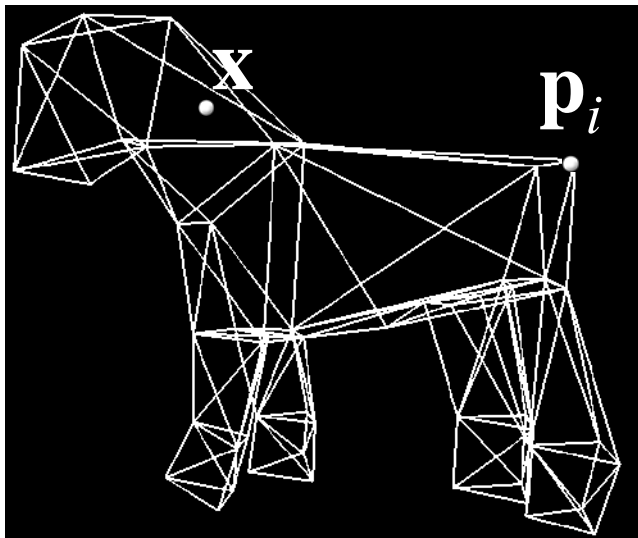
- Cage = crude version of the input shape
- Polytope (not a lattice)



Cage-based Deformations

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- Cage = crude version of the input shape
- Polytope (not a lattice)
- Each point \mathbf{x} in space is represented w.r.t. to the cage elements using coordinate functions

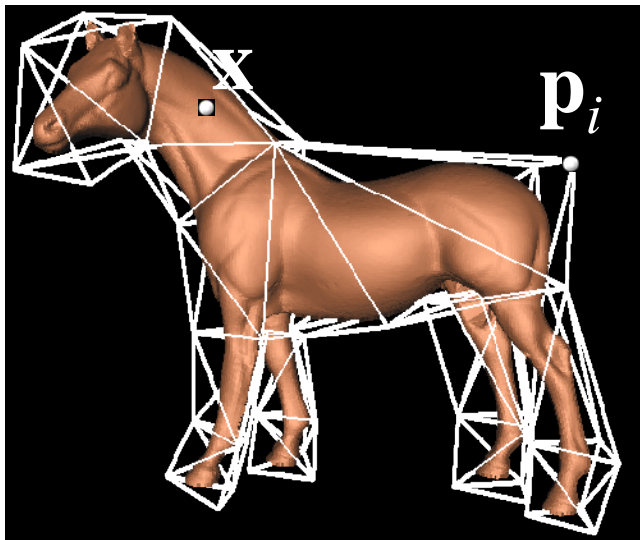


$$\mathbf{x} = \sum_{i=1}^k w_i(\mathbf{x}) \mathbf{p}_i$$

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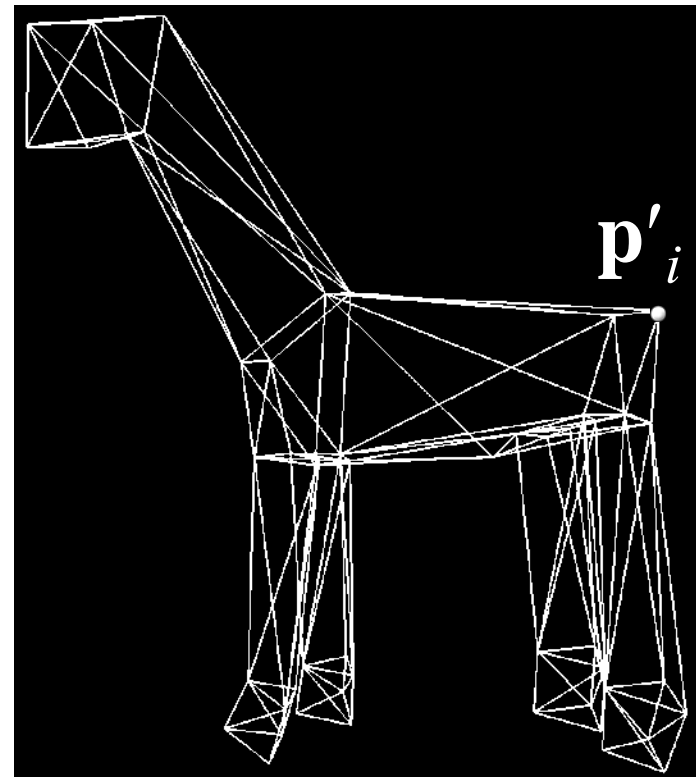
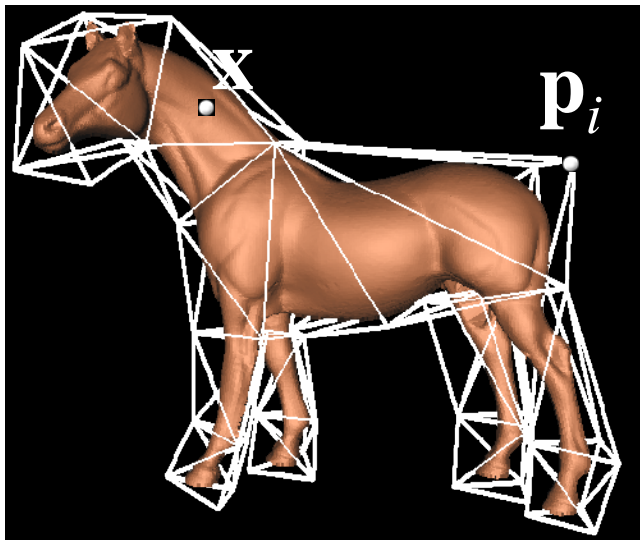


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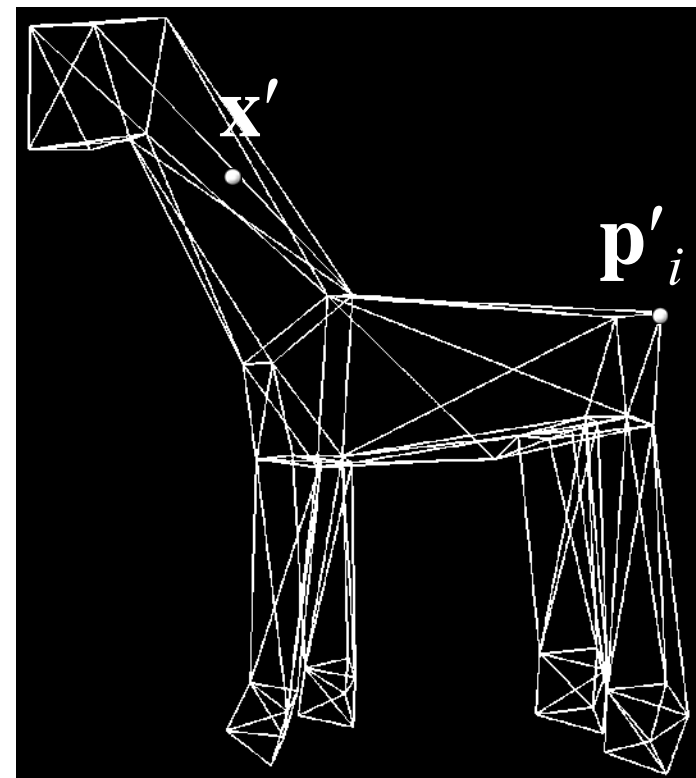
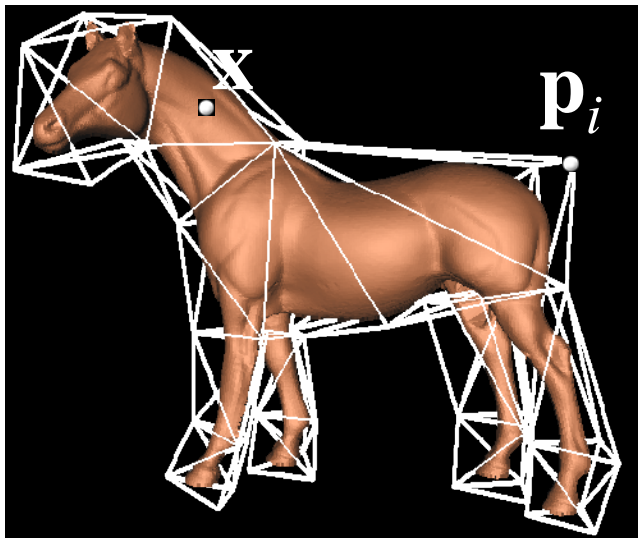


Cage-based Deformations

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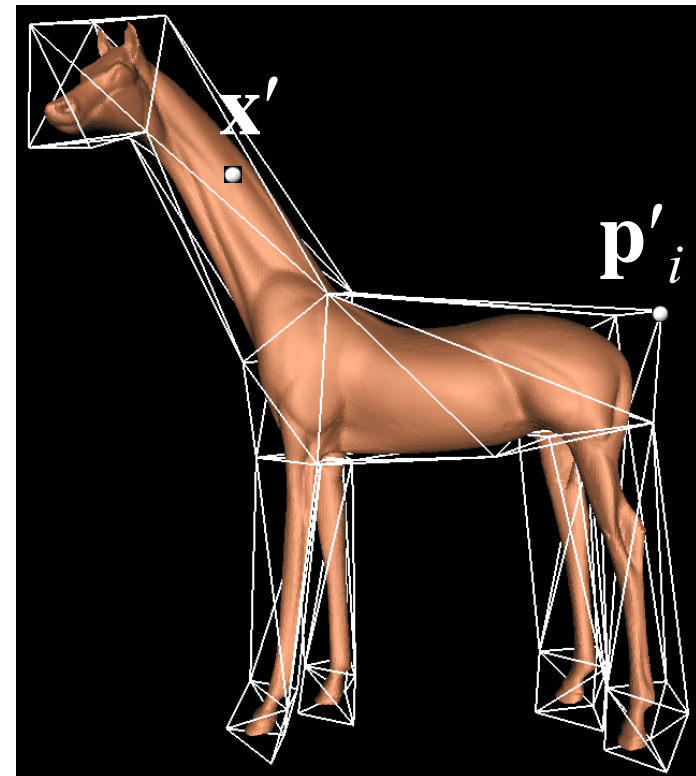
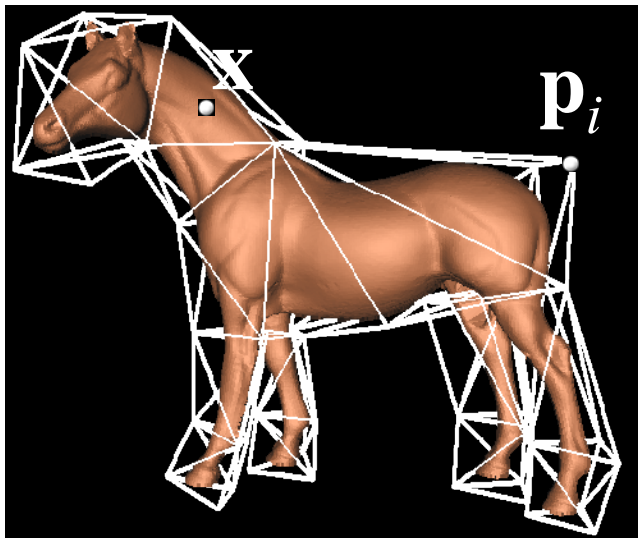


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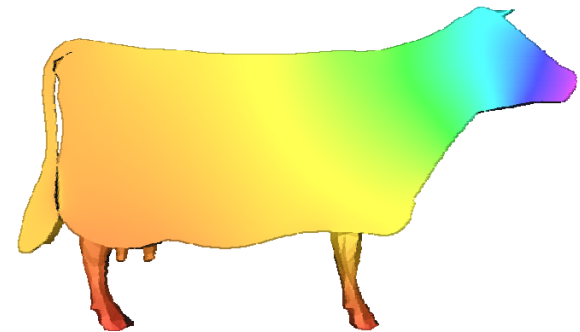
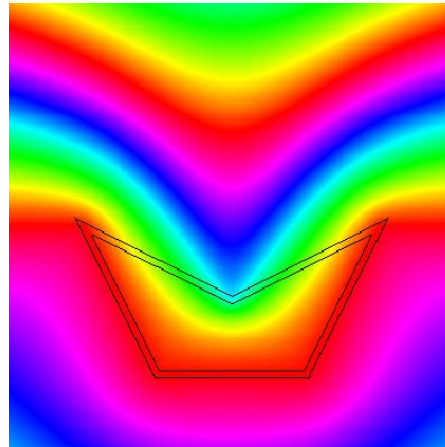
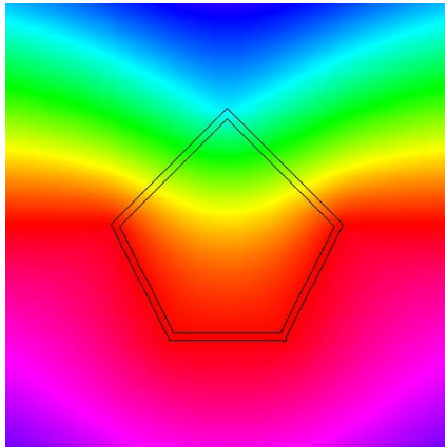
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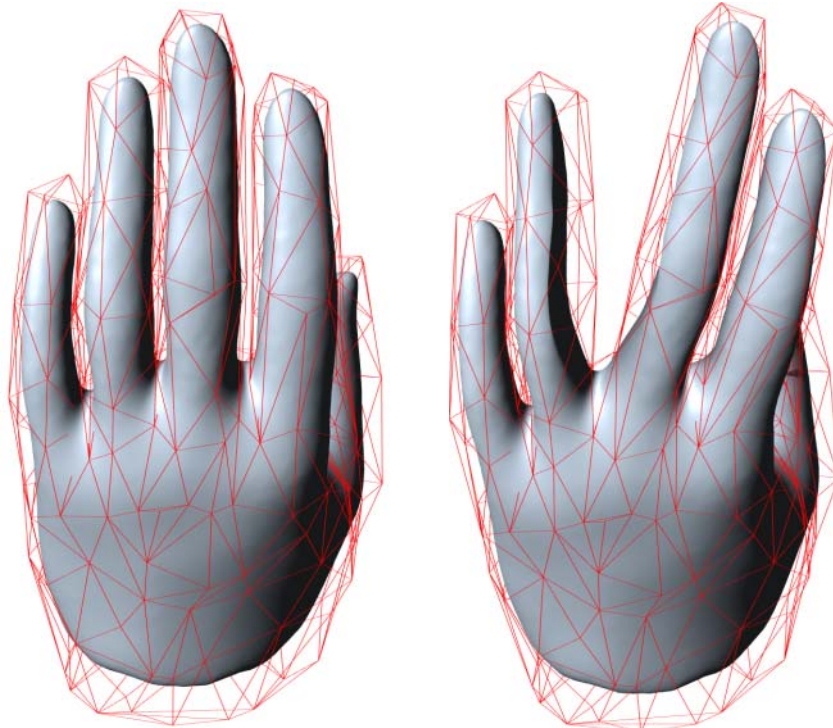
Coordinate Functions

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



Coordinate Functions

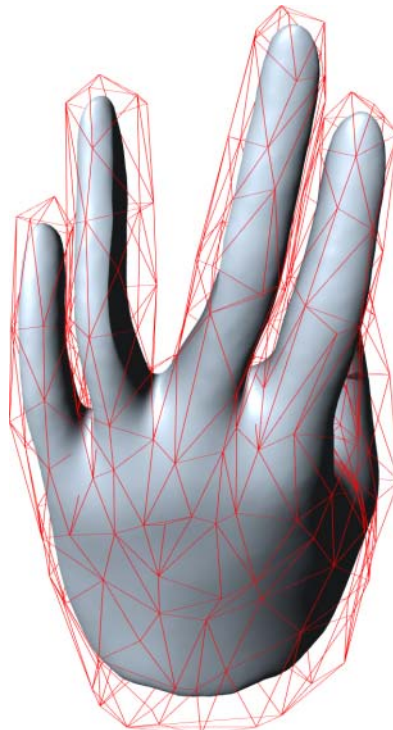
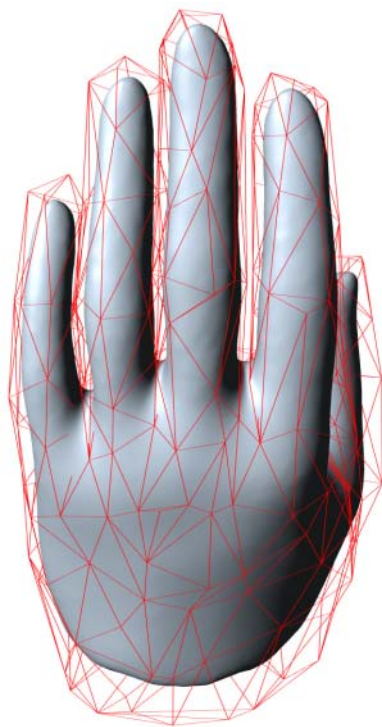
- Mean-value coordinates (Floater, Ju et al. 2005)
 - Not necessarily positive on non-convex domains



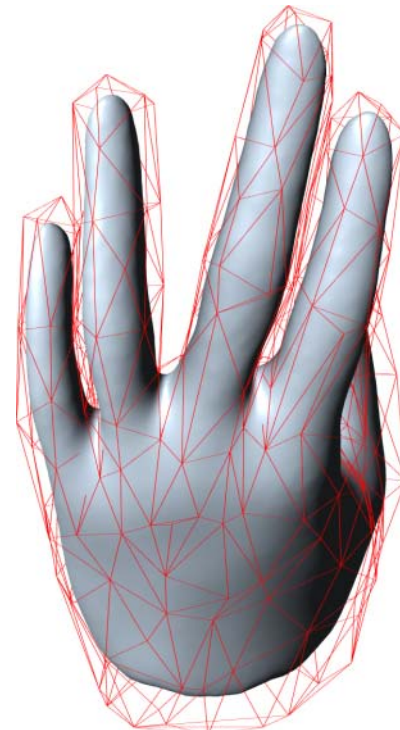
MVC

Coordinate Functions

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



MVC



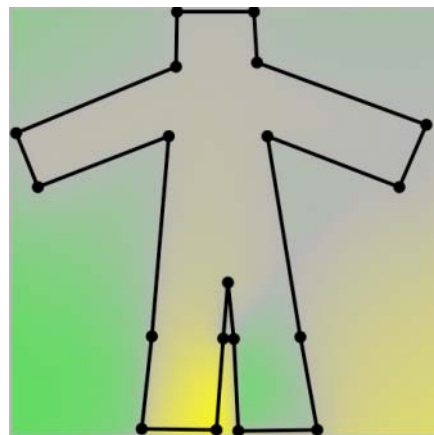
PMVC

Coordinate Functions

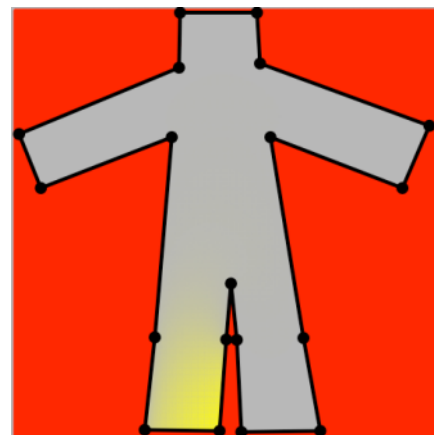
- Harmonic coordinates (Joshi et al. 2007)
 - Harmonic functions $h_i(\mathbf{x})$ for each cage vertex \mathbf{p}_i
 - Solve

$$\Delta h = 0$$

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



MVC



HC

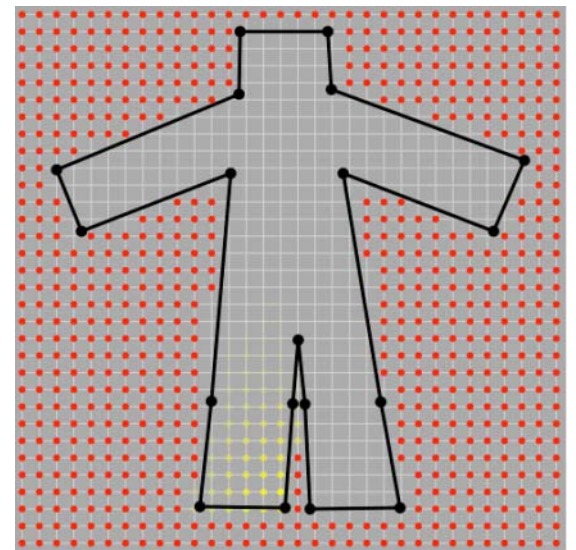
Coordinate Functions

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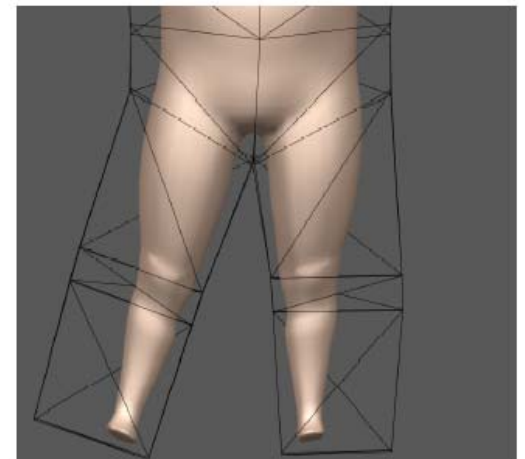
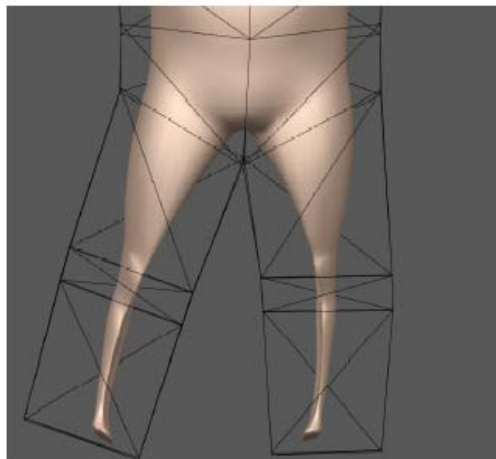
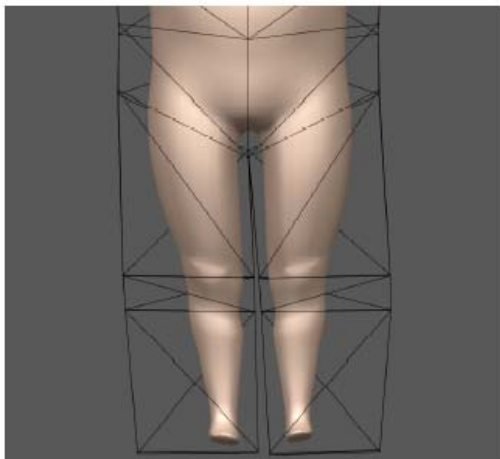
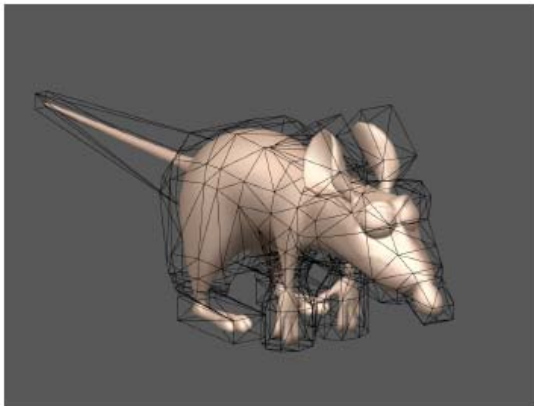
subject to: h_i linear on the boundary s.t. $h_i(\mathbf{p}_i) = \delta_{ij}$

- Volumetric Laplace equation
- Discretization, no closed-form



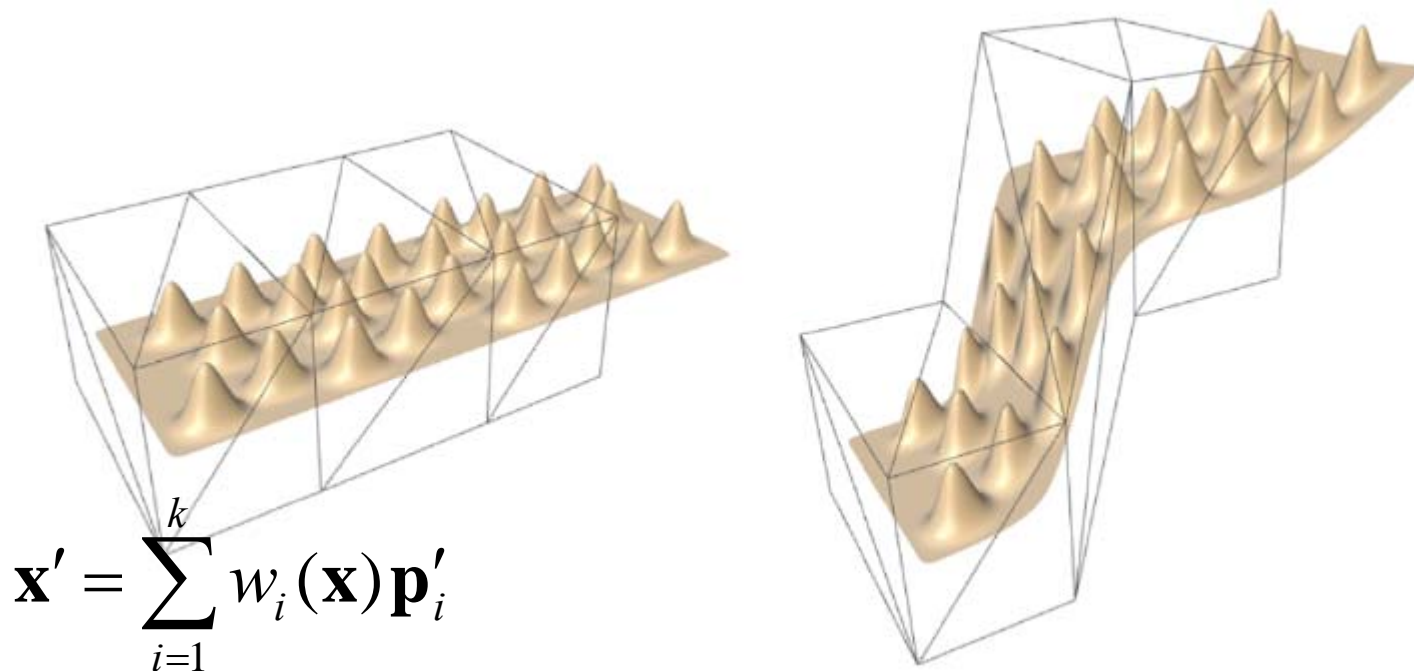
Coordinate Functions

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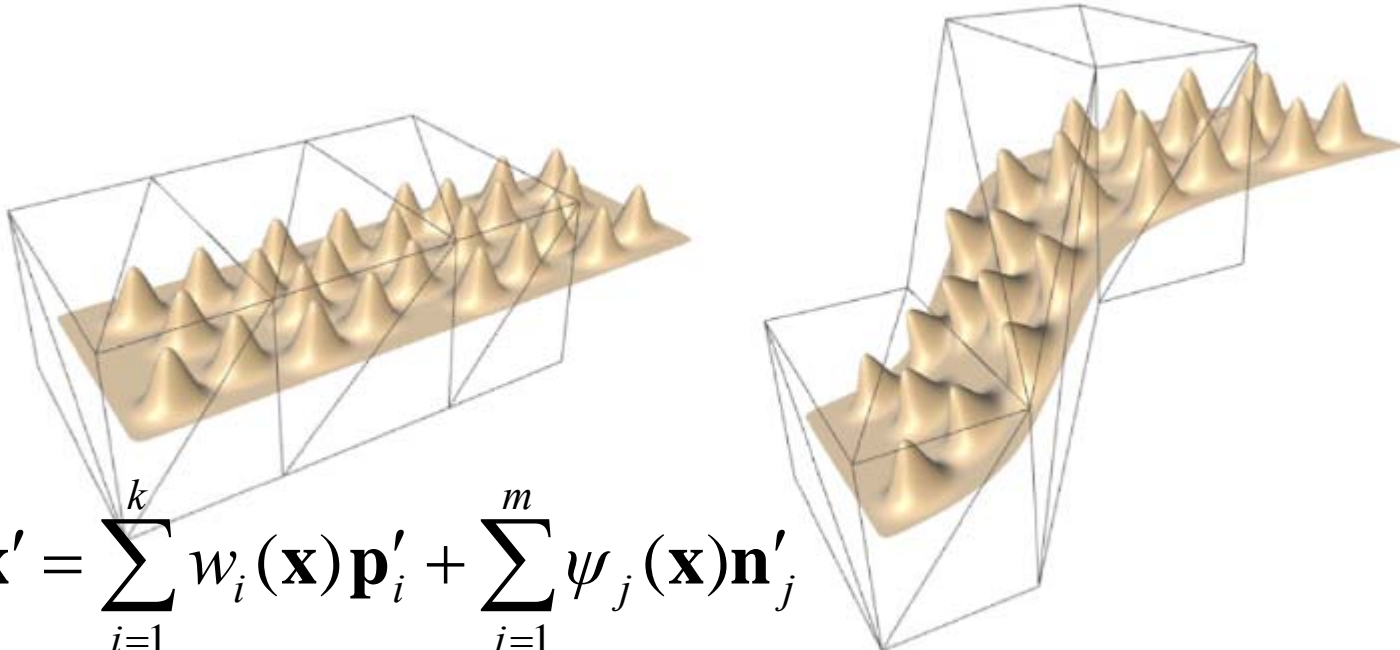
Coordinate Functions

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



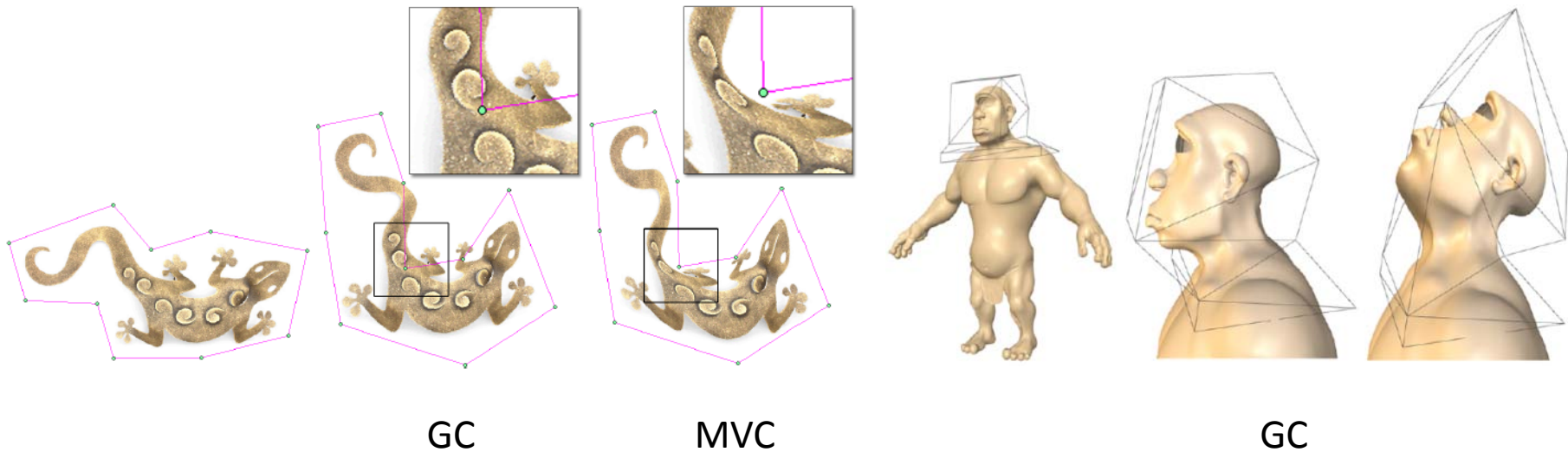
Coordinate Functions

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


$$\mathbf{x}' = \sum_{i=1}^k w_i(\mathbf{x}) \mathbf{p}'_i + \sum_{j=1}^m \psi_j(\mathbf{x}) \mathbf{n}'_j$$

Coordinate Functions

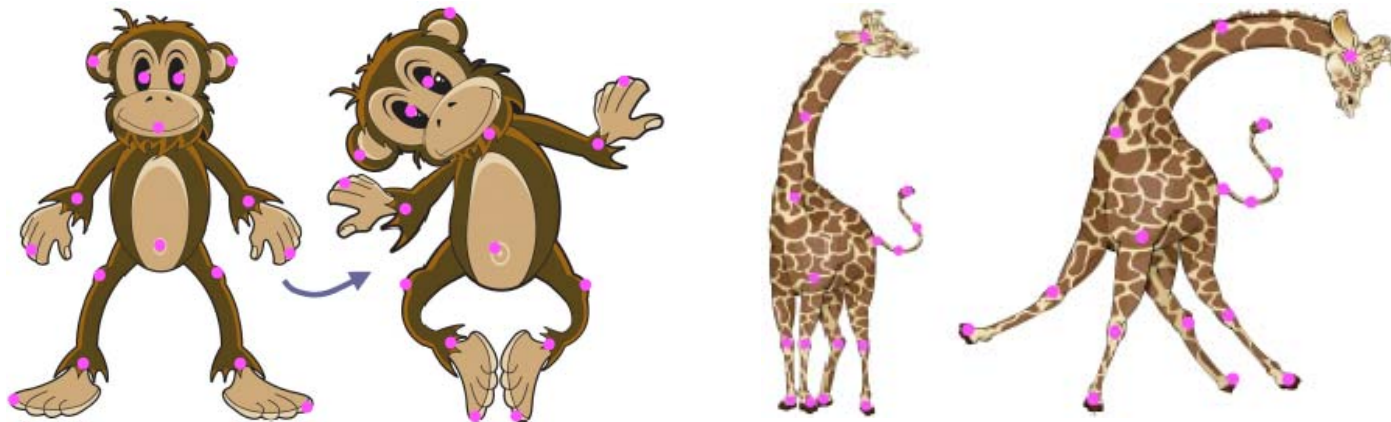
- Green coordinates (Lipman et al. 2008)
- Closed-form solution
- Conformal in 2D, quasi-conformal in 3D



Coordinate Functions

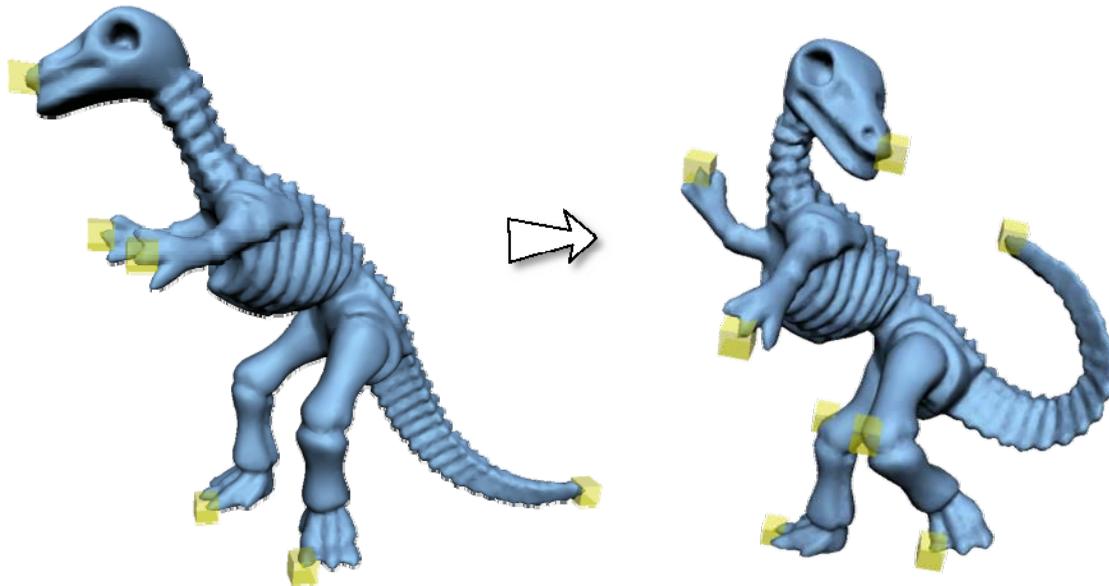
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Alternative interpretation in 2D via holomorphic functions and extension to point handles : **Weber et al. Eurographics 2009**



Nonlinear Space Deformations

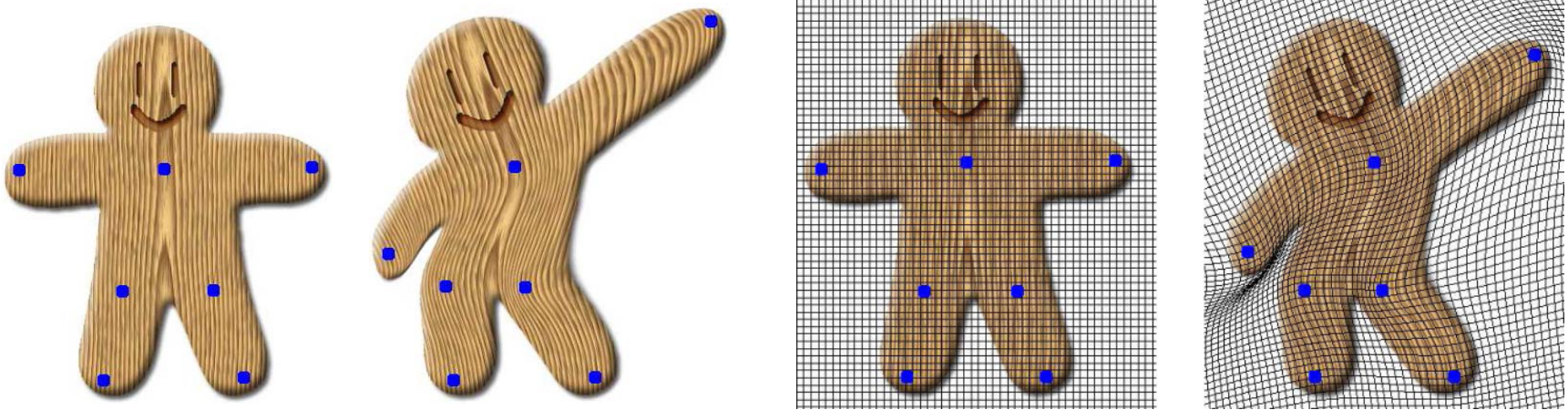
- Involve nonlinear optimization
- Enjoy the advantages of space warps
- Additionally, have shape-preserving properties



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

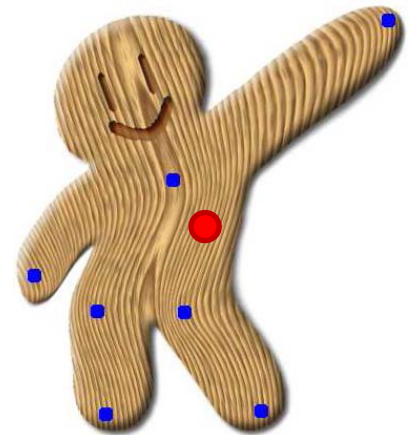
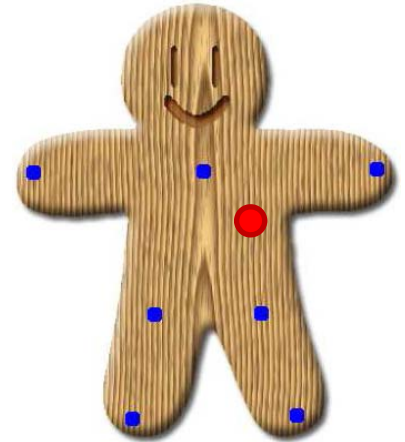
Moving-Least-Squares (MLS) approach [Schaefer et al. 2006]

- 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]

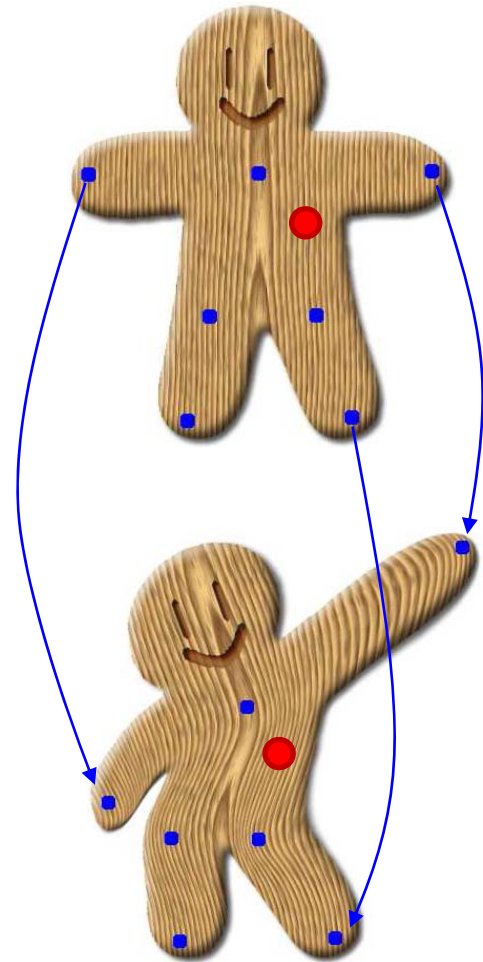
- Handles \mathbf{p}_i are displaced to \mathbf{q}_i
- The local transformation at \mathbf{x} :

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

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

- The weights depend on \mathbf{x} :

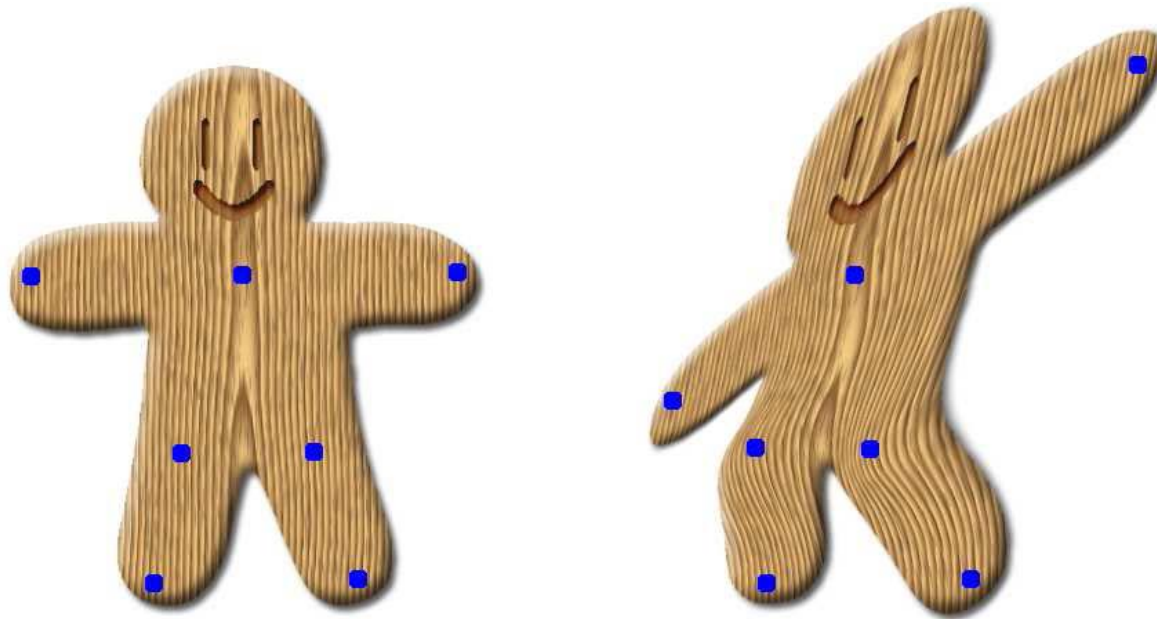
$$w_i(\mathbf{x}) = \|\mathbf{p}_i - \mathbf{x}\|^{-2\alpha}$$



As-Rigid-As-Possible Deformation

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

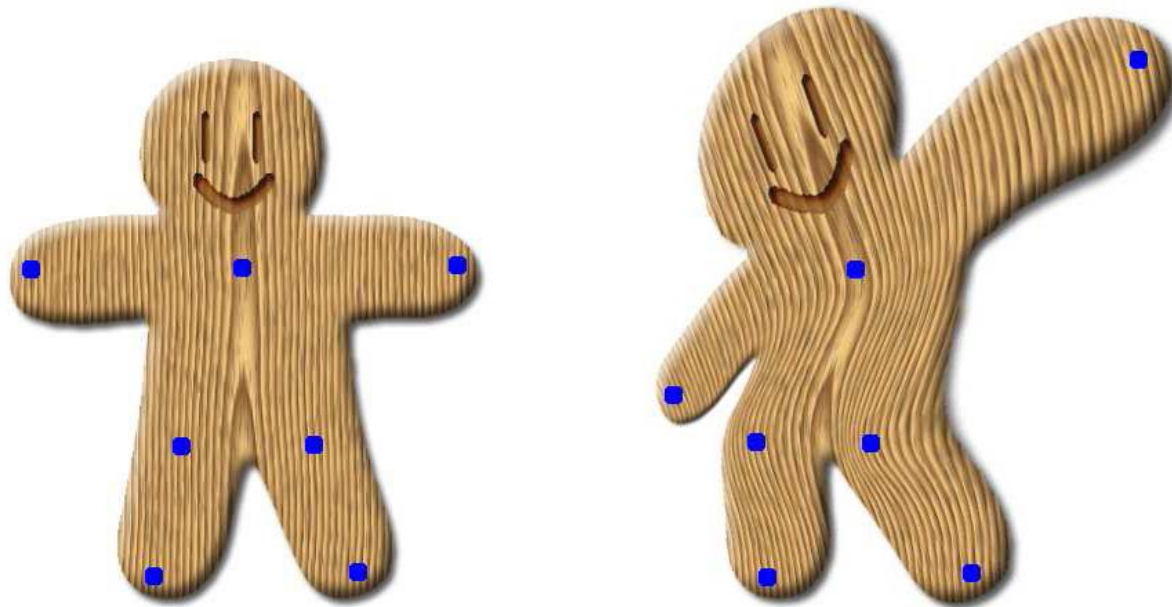
- No additional restriction on $A_x(\cdot)$ – affine local transformations



As-Rigid-As-Possible Deformation

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

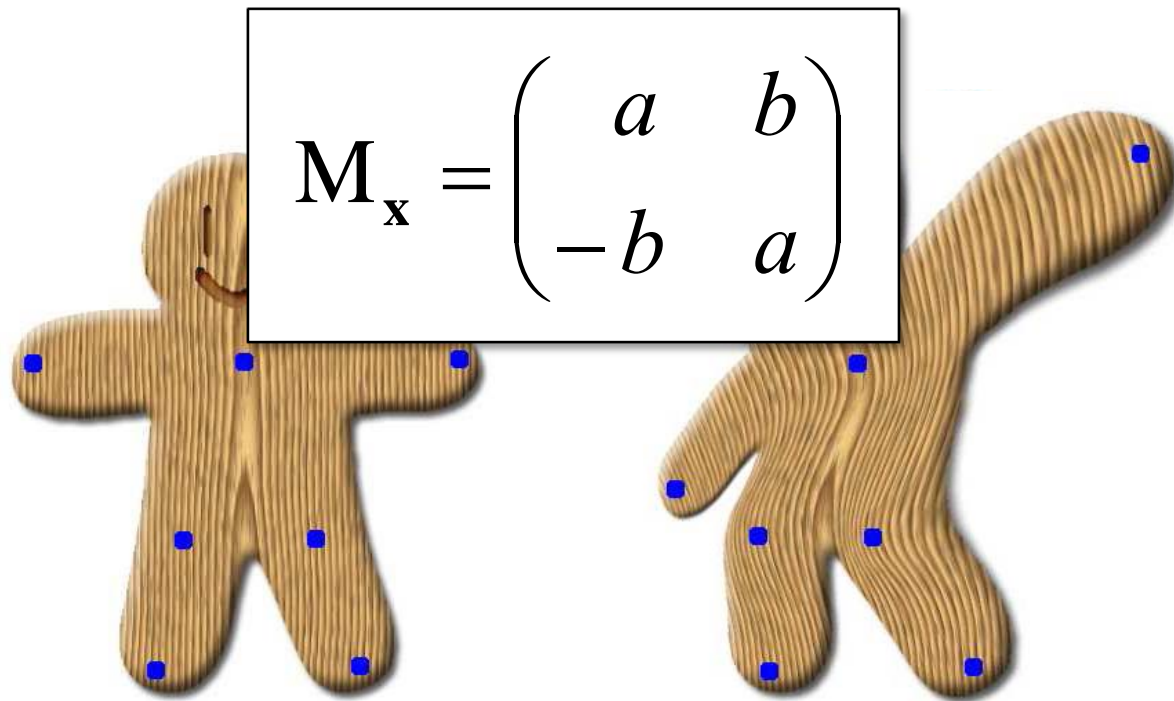
- Restrict $A_x(\cdot)$ to similarity



As-Rigid-As-Possible Deformation

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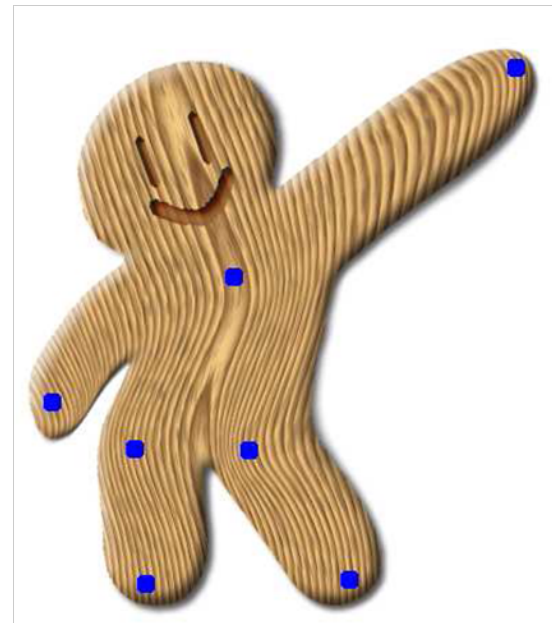
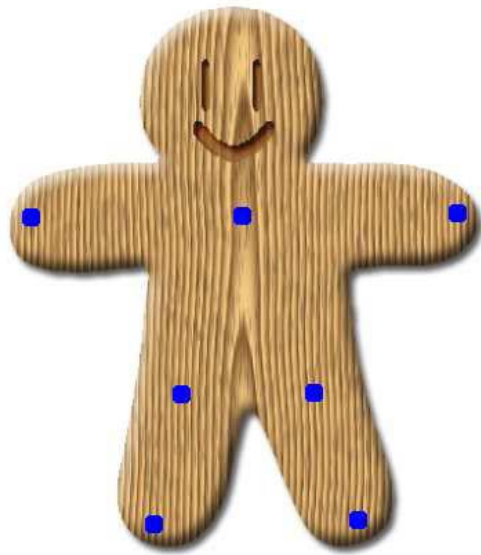
- Restrict $A_{\mathbf{x}}(\cdot)$ to similarity



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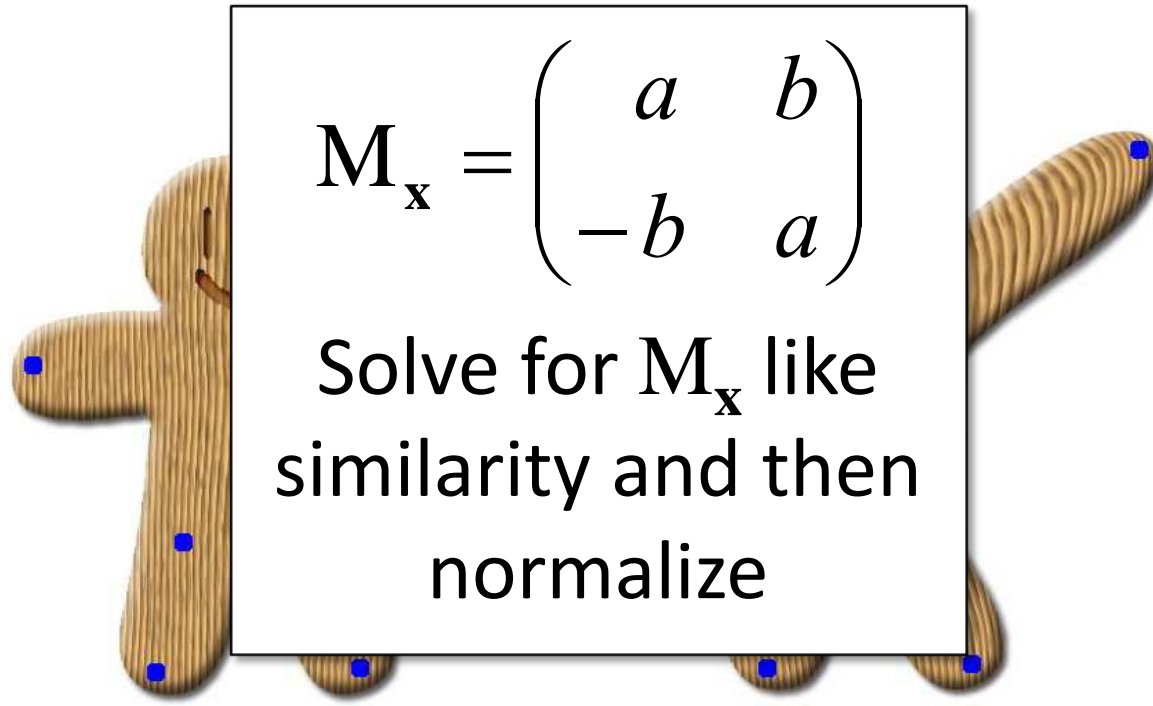
- Restrict $A_x(\cdot)$ to rigid



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

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

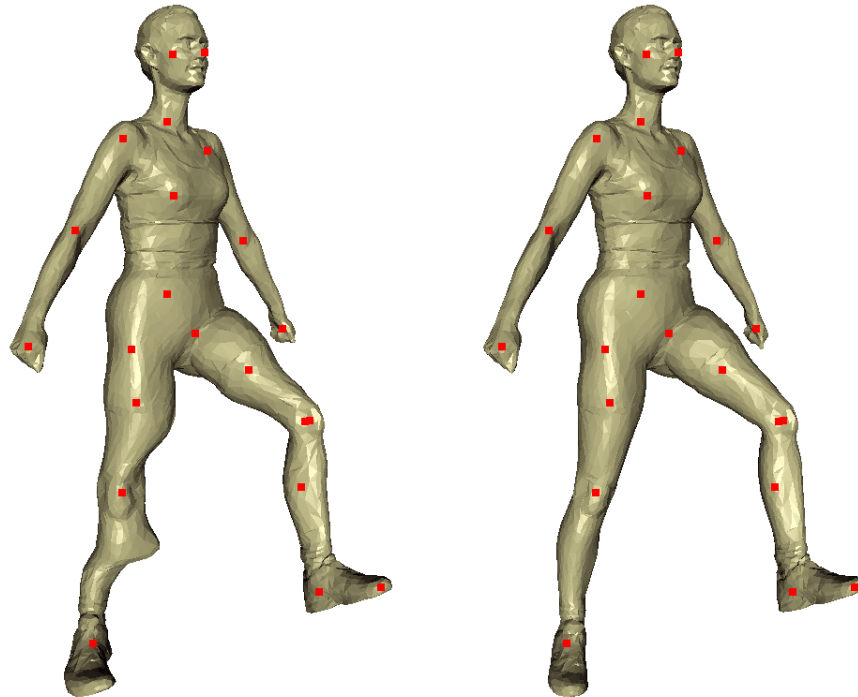
by polar decomposition of the 3×3 covariance matrix

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”

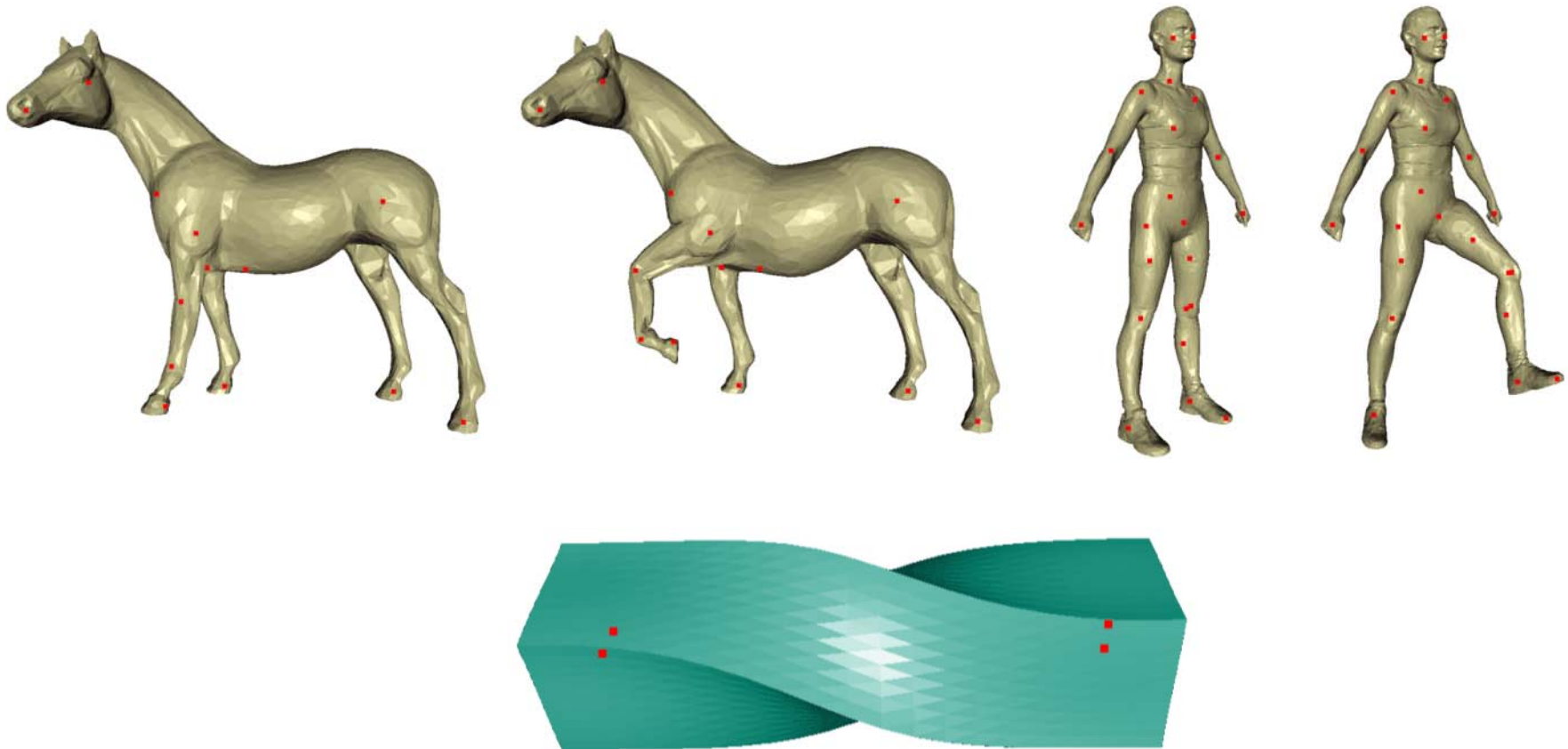
$$w_i(\mathbf{x}) = d(\mathbf{p}_i, \mathbf{x})^{-2\alpha}$$



As-Rigid-As-Possible Deformation

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

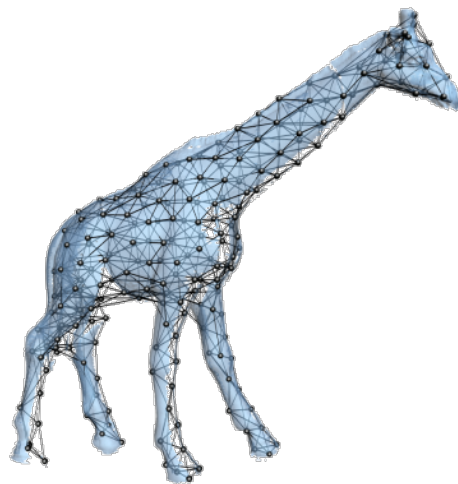
- More results



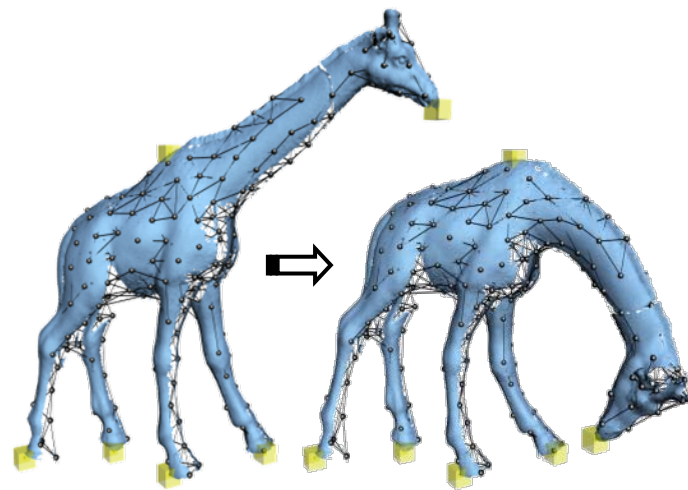
As-Rigid-As-Possible Deformation

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



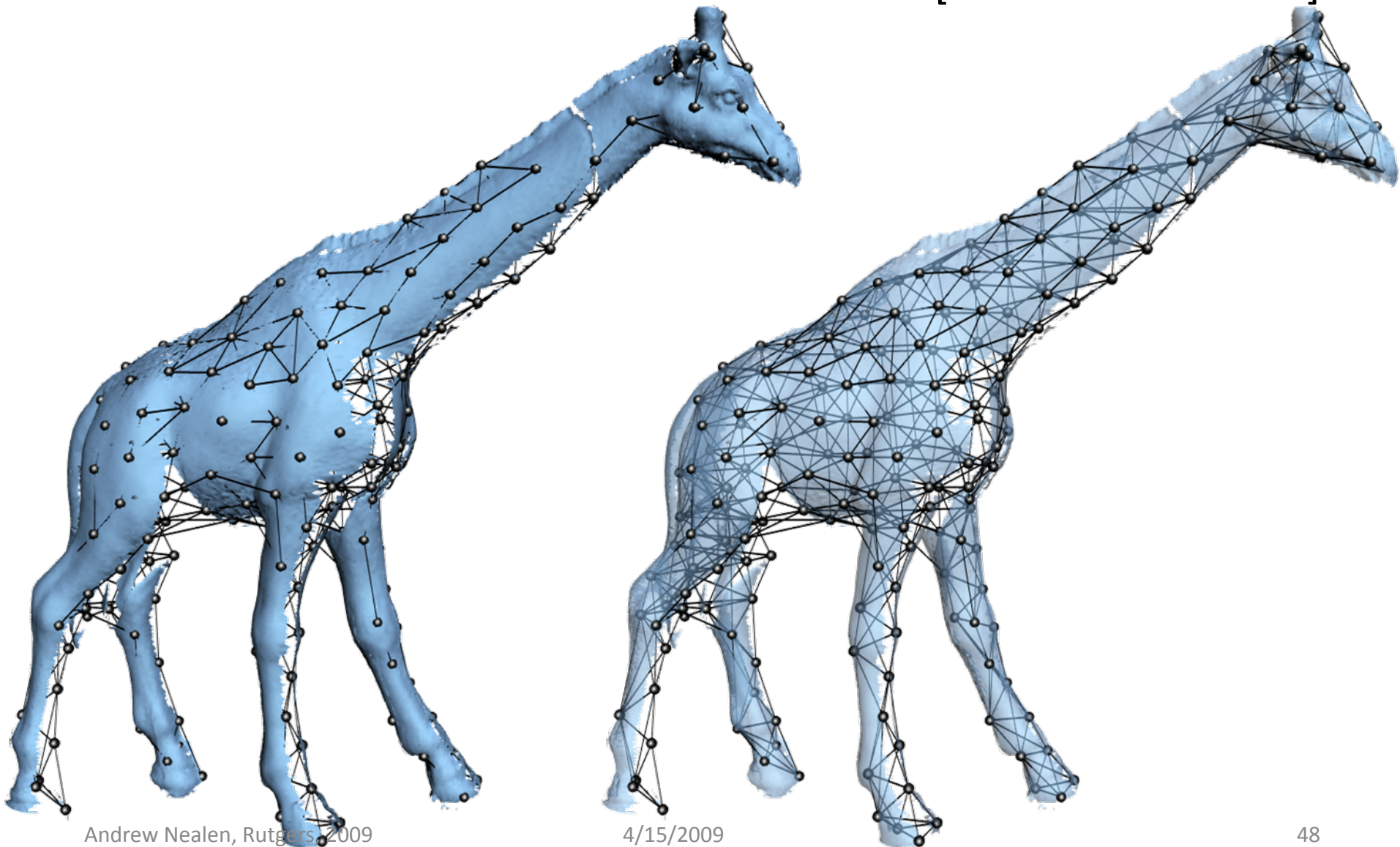
Deformation Graph



Optimization Procedure

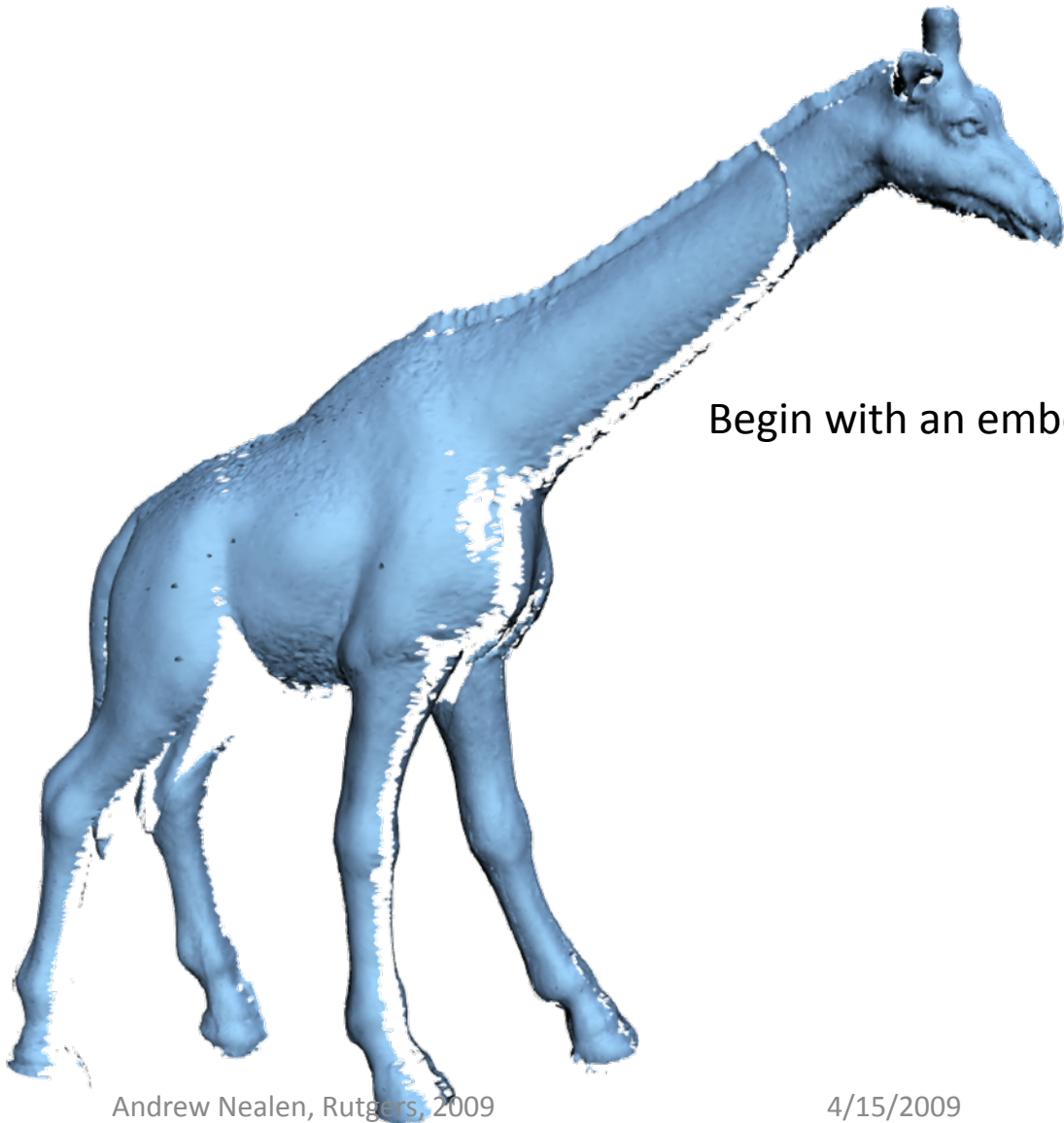
Deformation Graph

[Sumner et al. 2007]



Deformation Graph

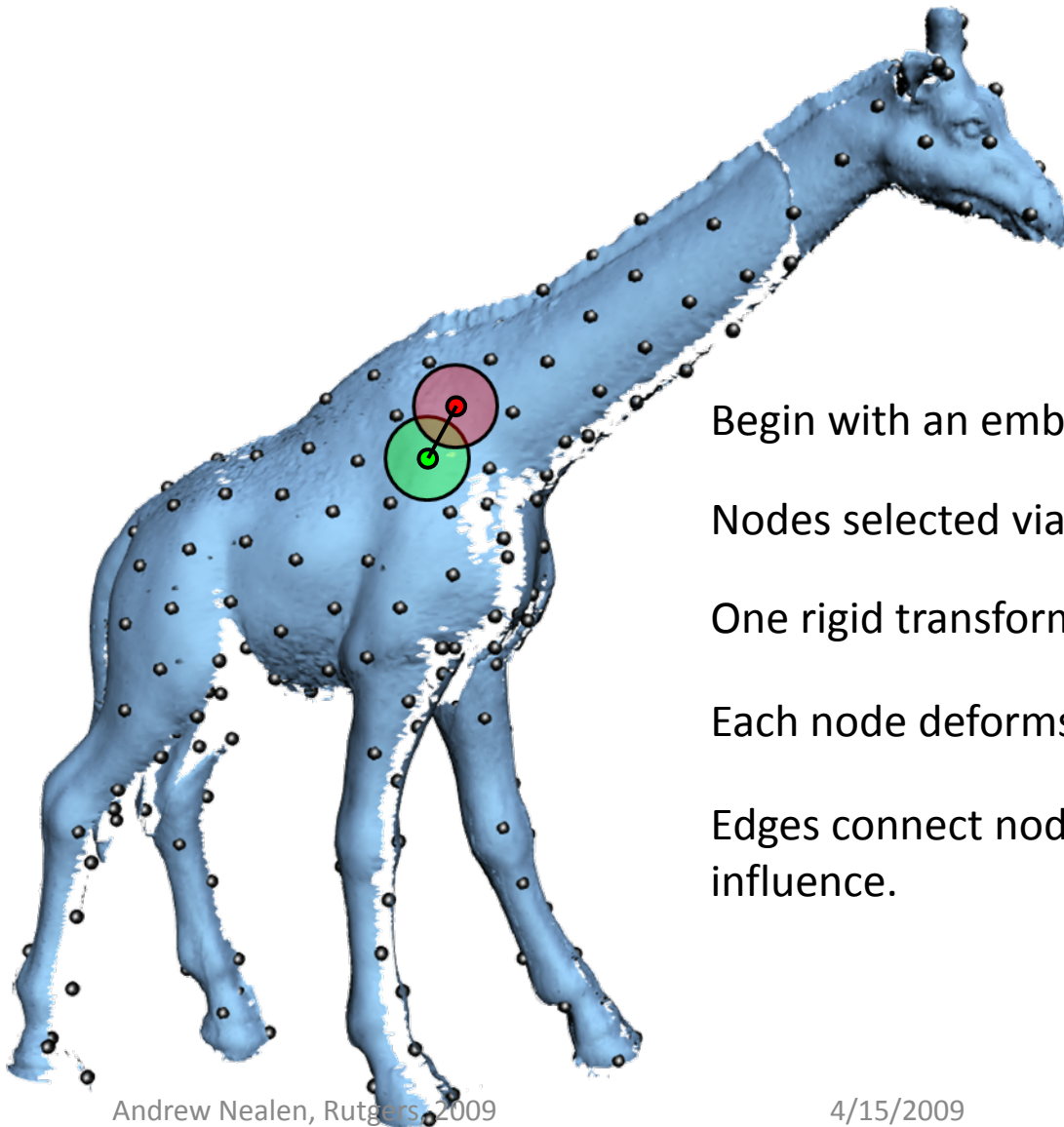
[Sumner et al. 2007]



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

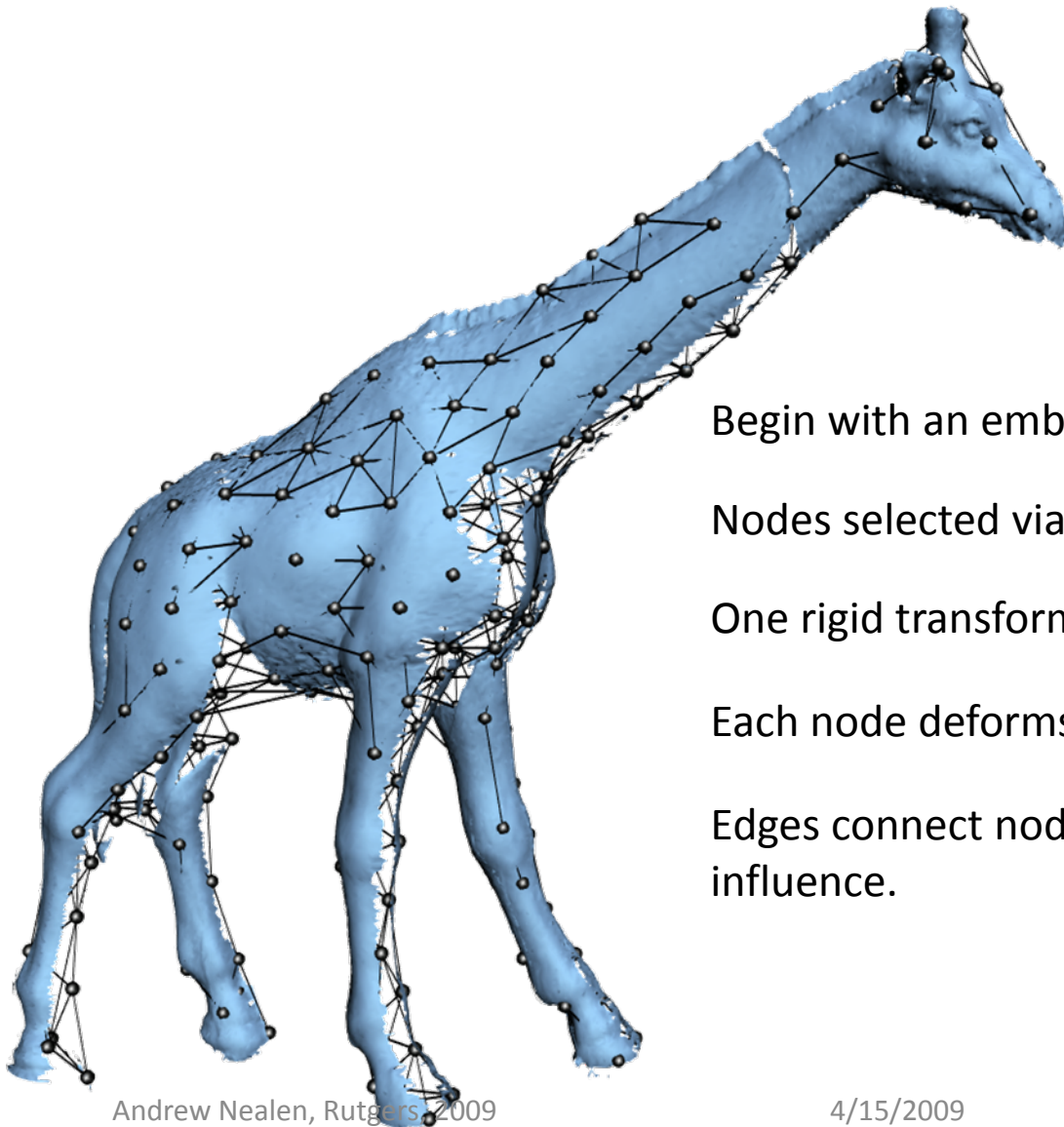
One rigid transformation for each node: $\mathbf{R}_j, \mathbf{t}_j$

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

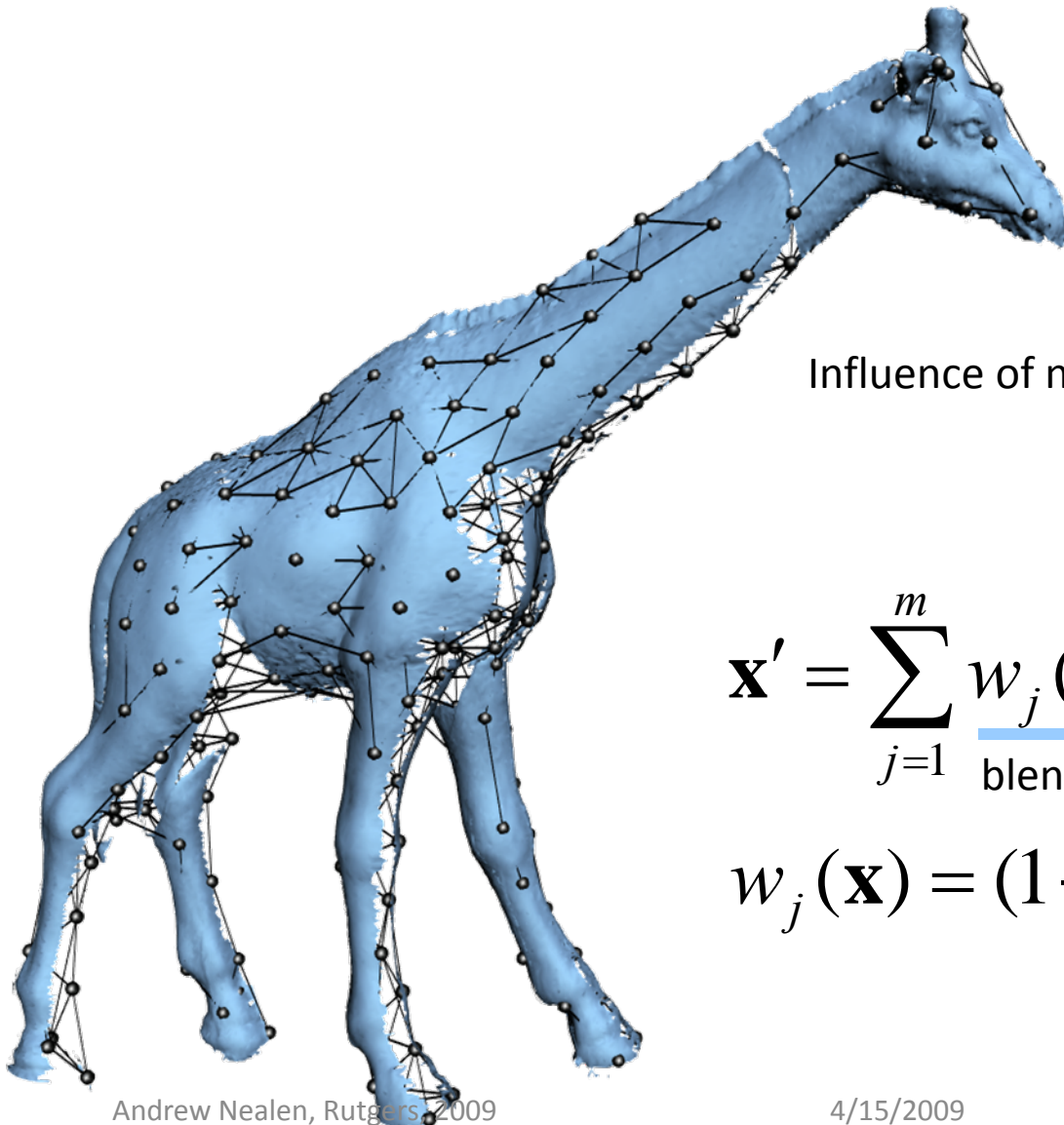
One rigid transformation for each node: $\mathbf{R}_j, \mathbf{t}_j$

Each node deforms nearby space.

Edges connect nodes of overlapping influence.

Deformation Graph

[Sumner et al. 2007]



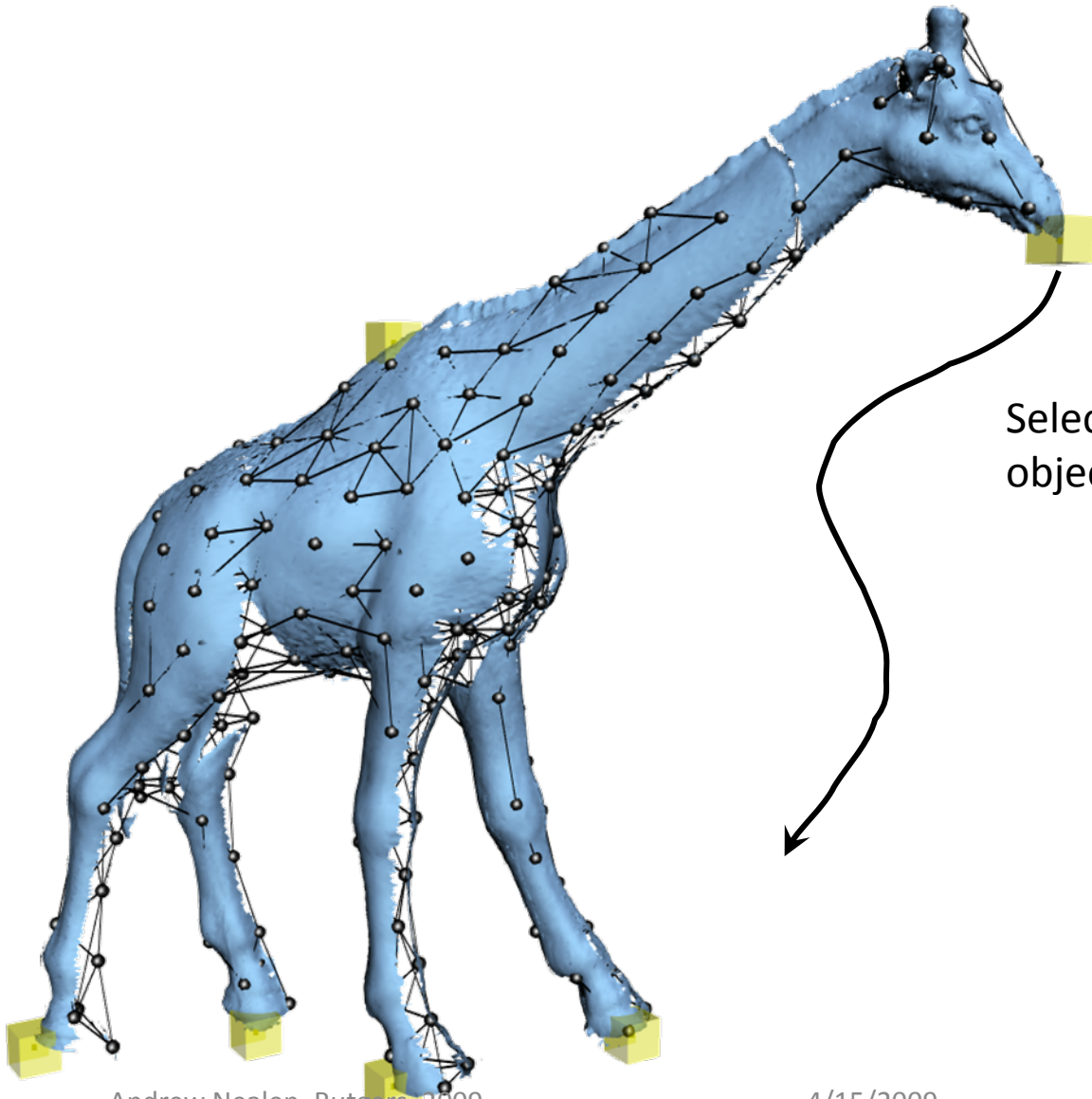
Influence of nearby transformations is blended.

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

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

Optimization

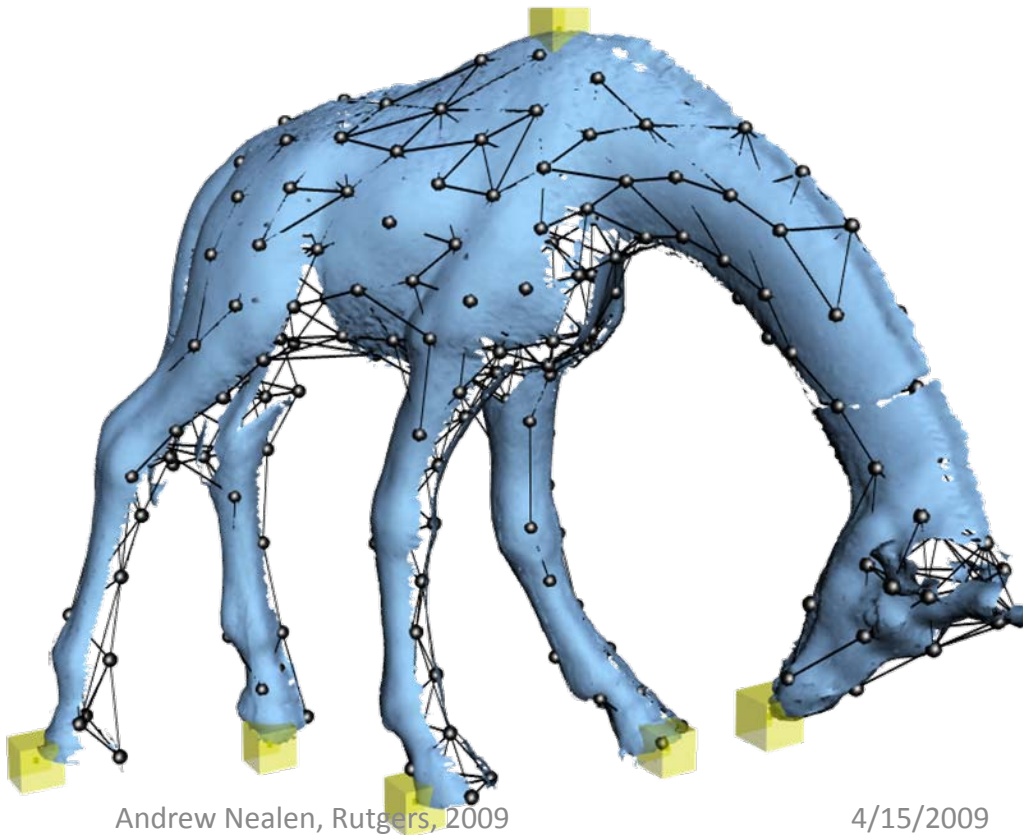
[Sumner et al. 2007]



Select & drag vertices of embedded object.

Optimization

[Sumner et al. 2007]



Select & drag vertices of embedded object.

Optimization finds

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

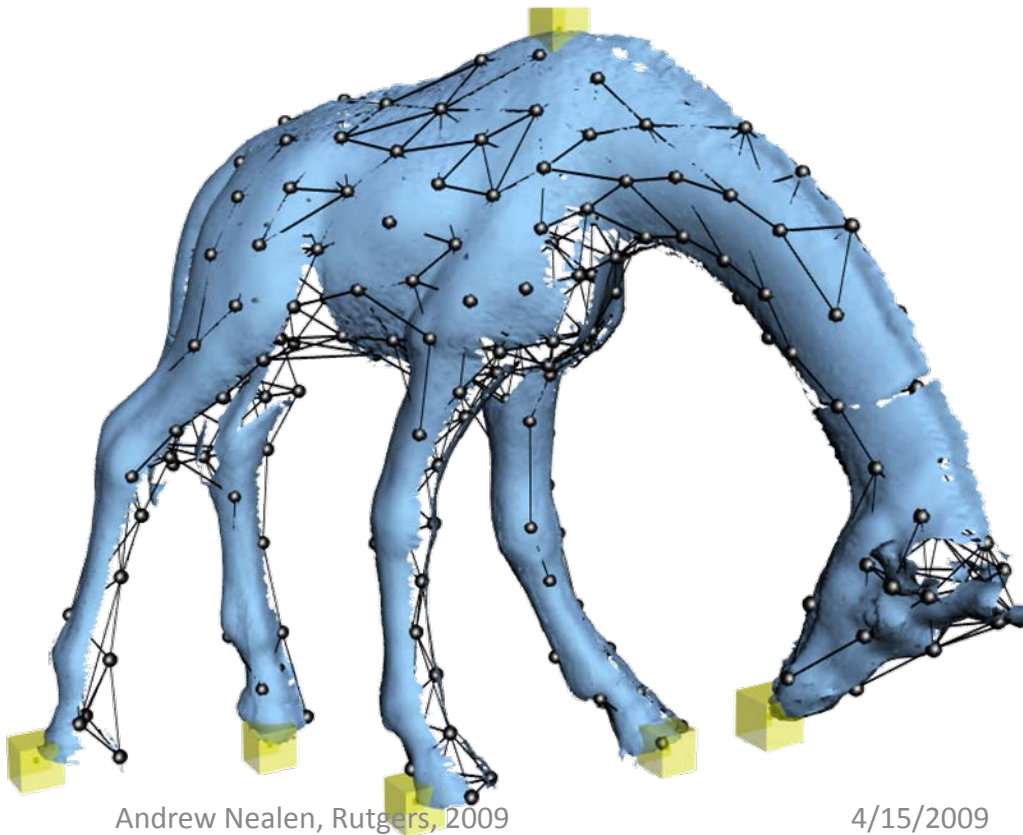
$$\min_{\mathbf{R}_1, \mathbf{t}_1, \dots, \mathbf{R}_m, \mathbf{t}_m} \underbrace{w_{\text{rot}} \mathbf{E}_{\text{rot}}}_{\text{Rotation term}} + \underbrace{w_{\text{reg}} \mathbf{E}_{\text{reg}}}_{\text{Regularization term}} + \underbrace{w_{\text{con}} \mathbf{E}_{\text{con}}}_{\text{Constraint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

Graph
parameters

Rotation
term

Regularization
term

Constraint
term

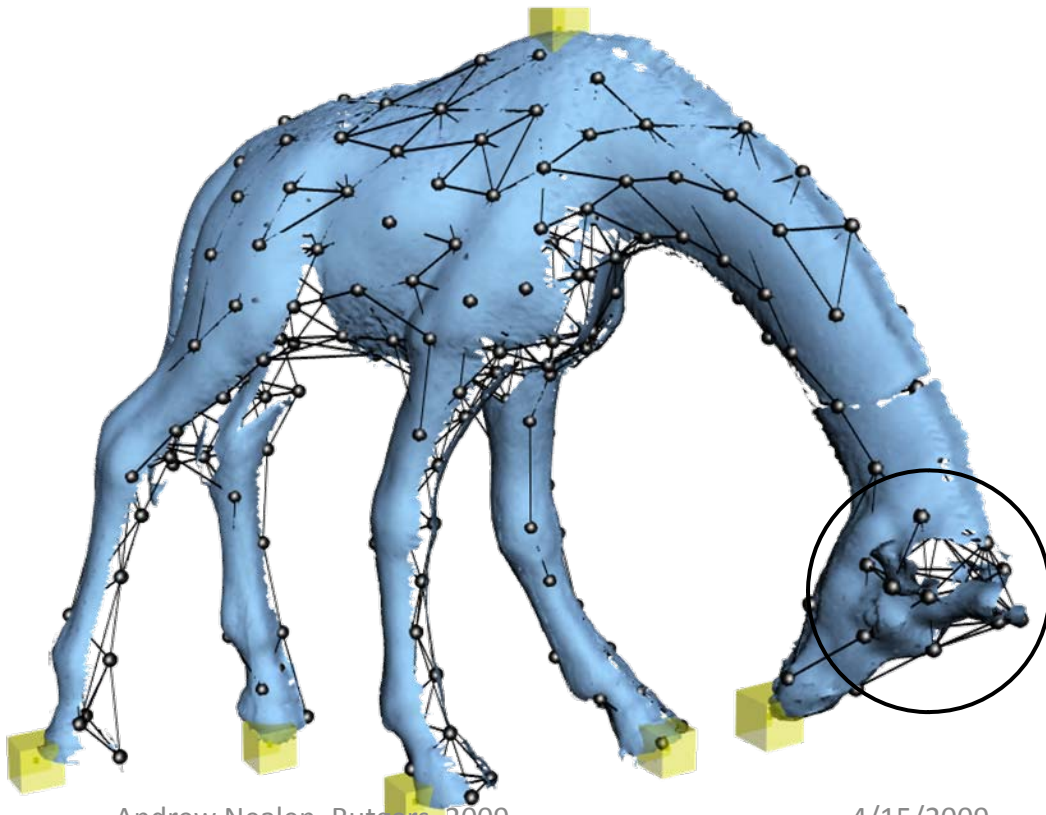


Select & drag vertices of embedded object.

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

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

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



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

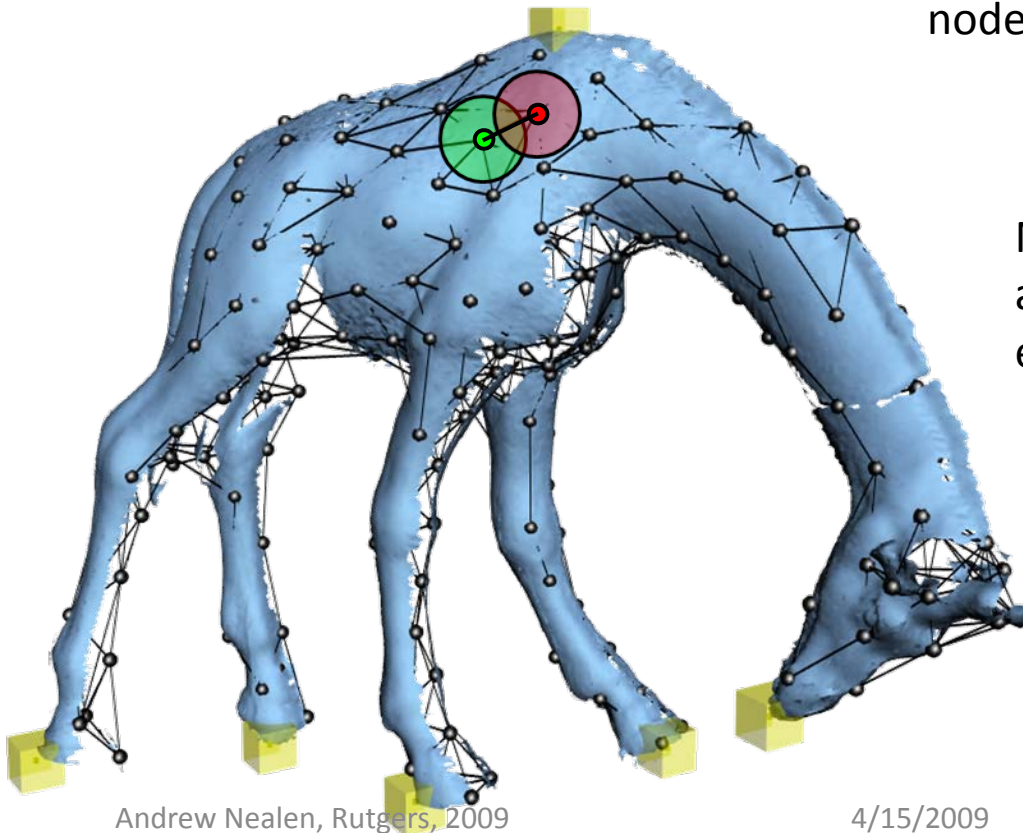
For detail preservation,
features should rotate and
not scale or skew.

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

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

where node j thinks
node k should go

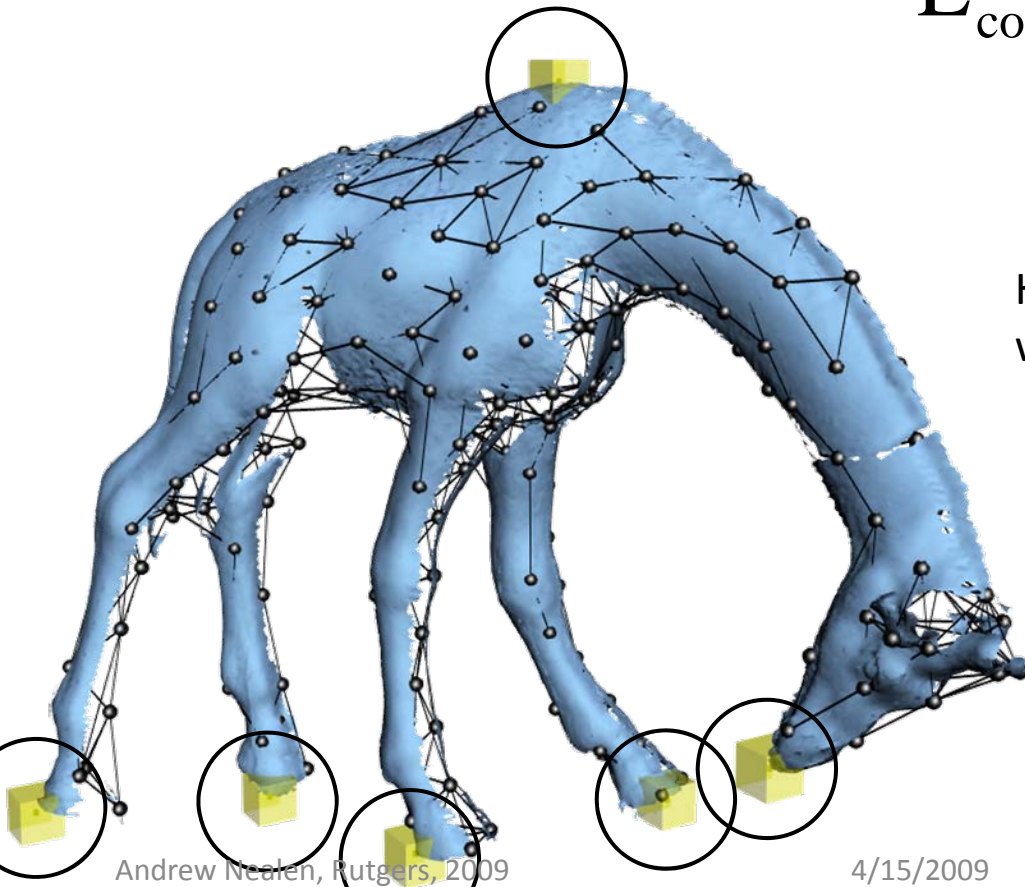
where node k
actually goes



Neighboring nodes should
agree on where they transform
each other.

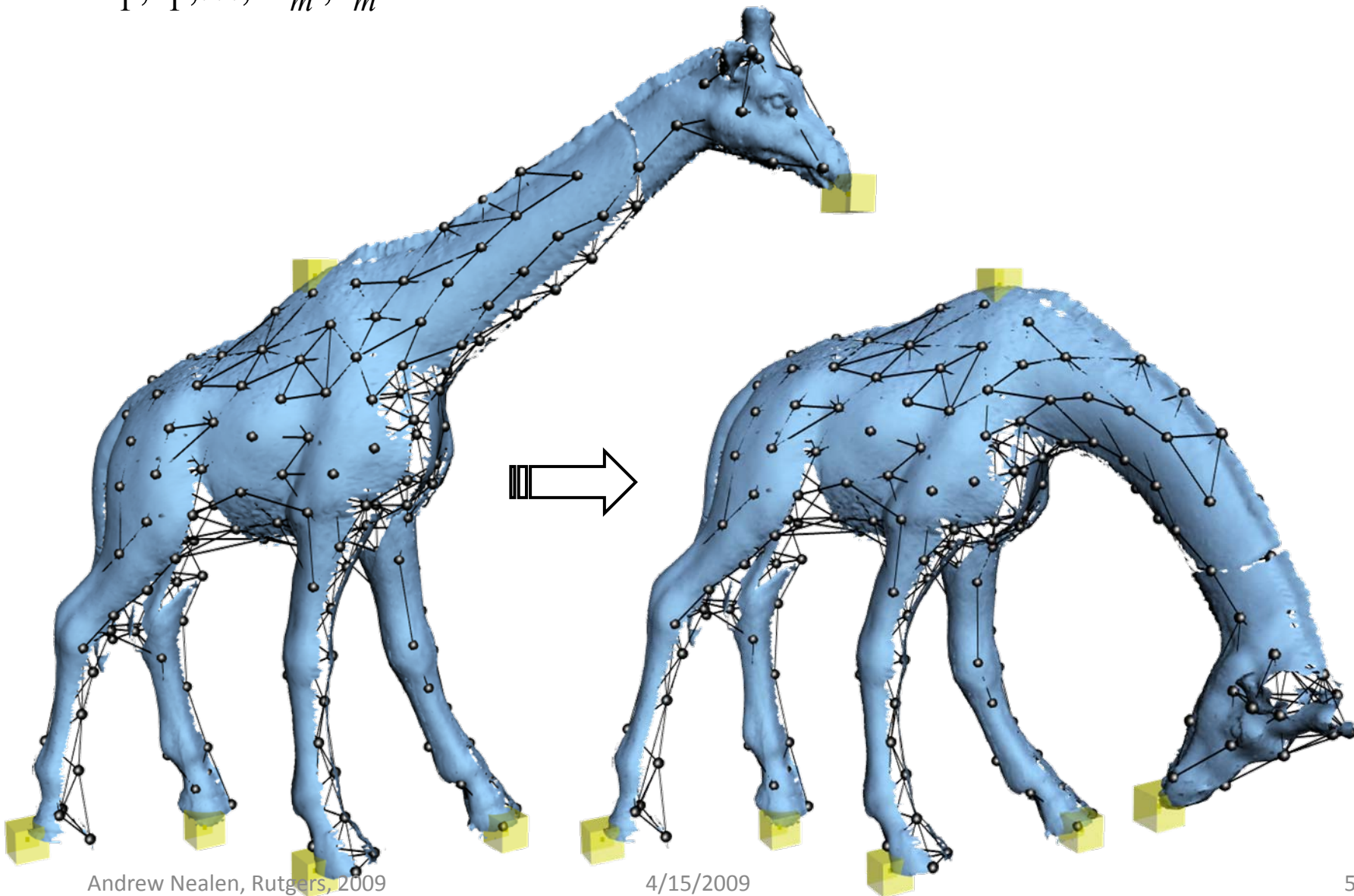
$$\min_{\mathbf{R}_1, \mathbf{t}_1, \dots, \mathbf{R}_m, \mathbf{t}_m} w_{\text{rot}} \mathbf{E}_{\text{rot}} + w_{\text{reg}} \mathbf{E}_{\text{reg}} + \underline{w_{\text{con}} \mathbf{E}_{\text{con}}}$$

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



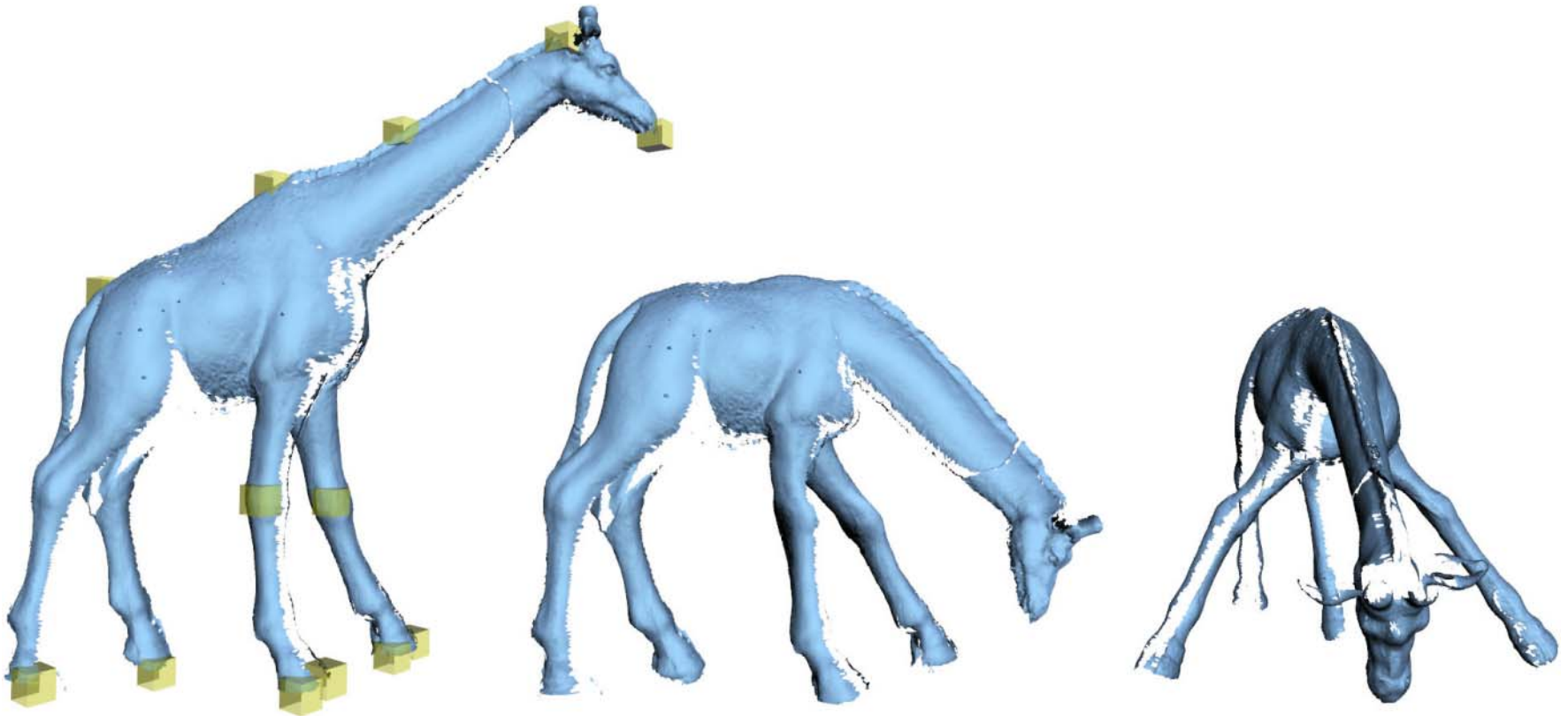
Handle vertices should go where the user puts them.

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



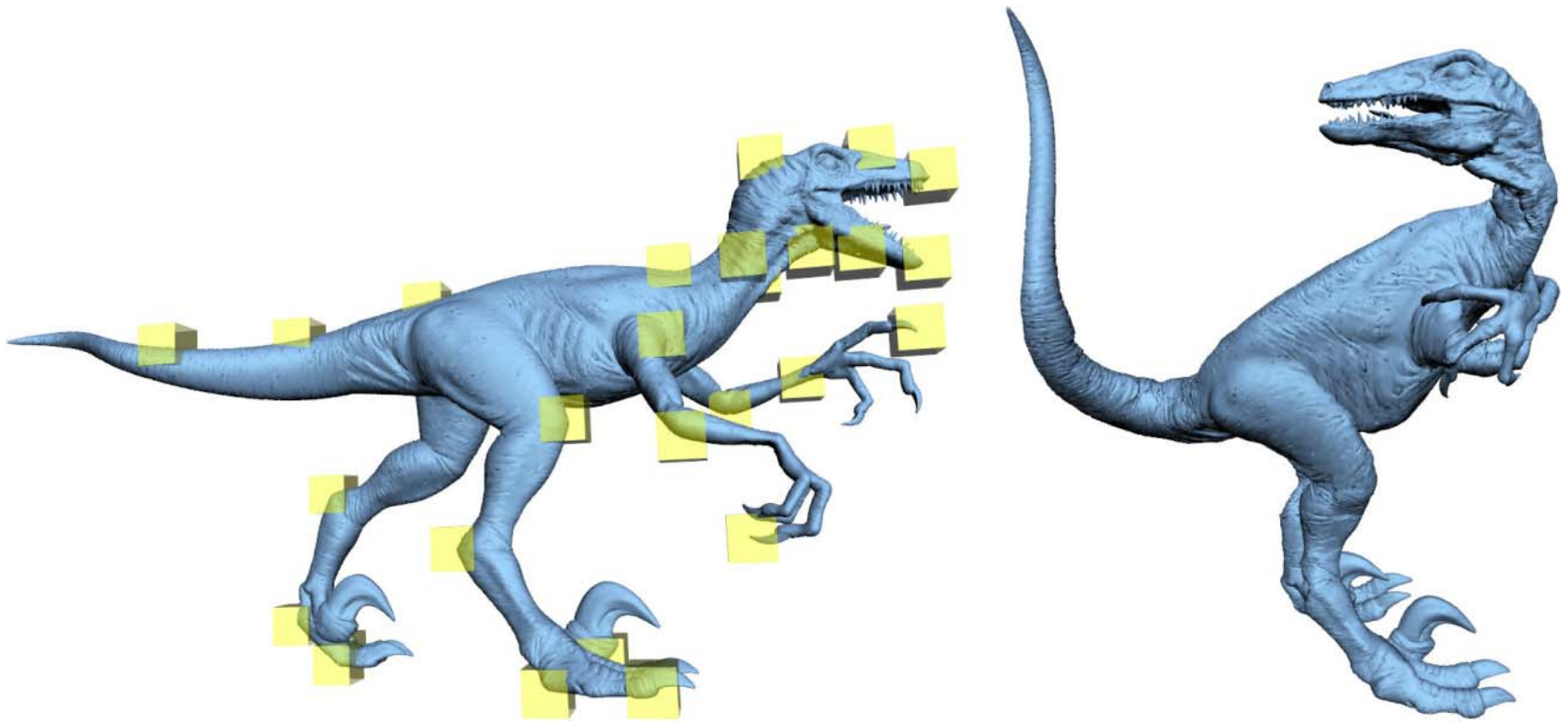
Results: Polygon Soup

[Sumner et al. 2007]



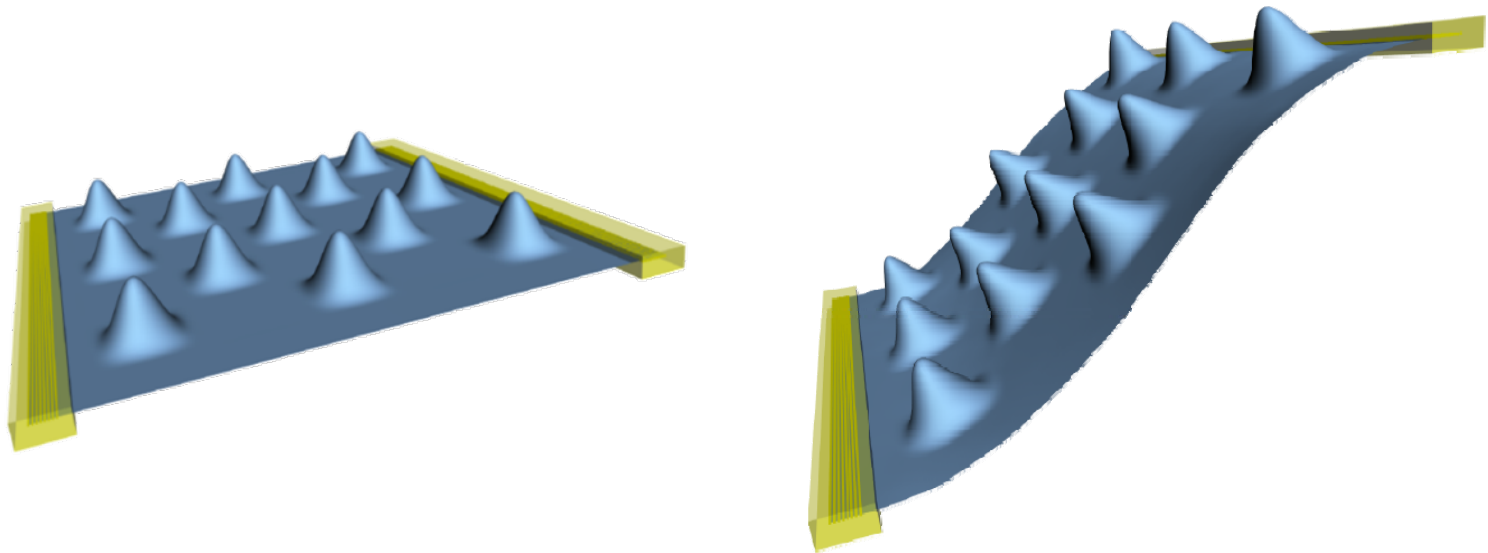
Results: Giant Mesh

[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

