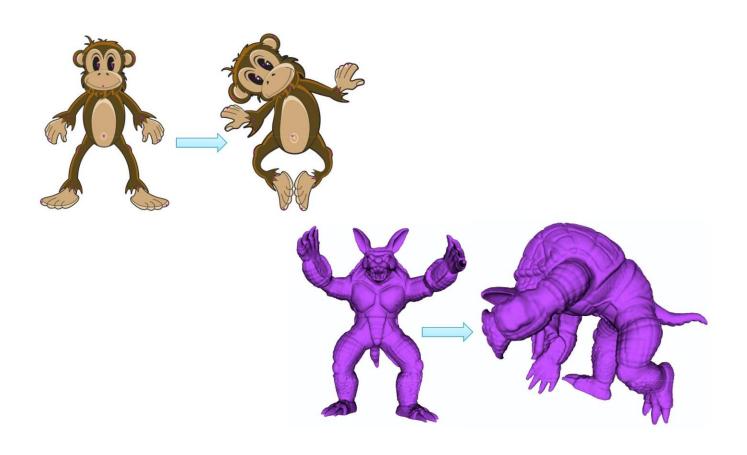
Deformation



Qixing Huang March. 9th 2017

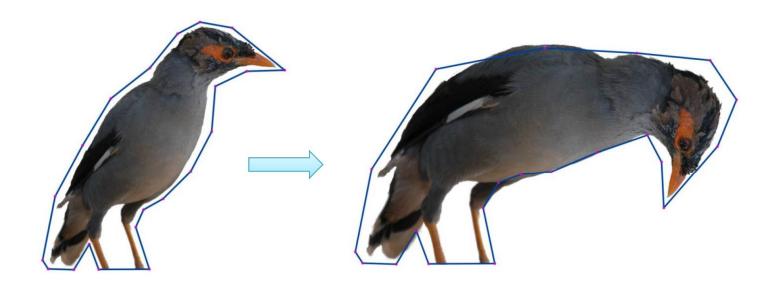


Deformation



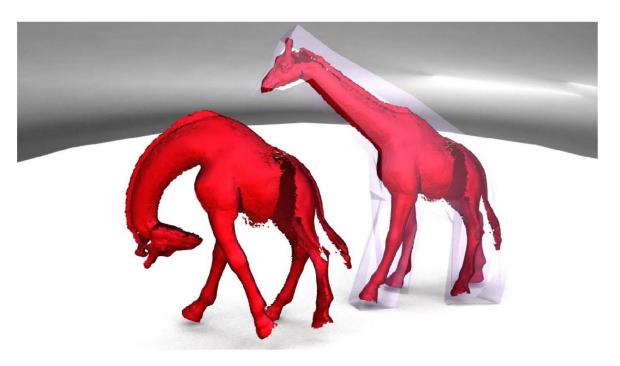
Motivation

Easy modeling – generate new shapes by deforming existing ones



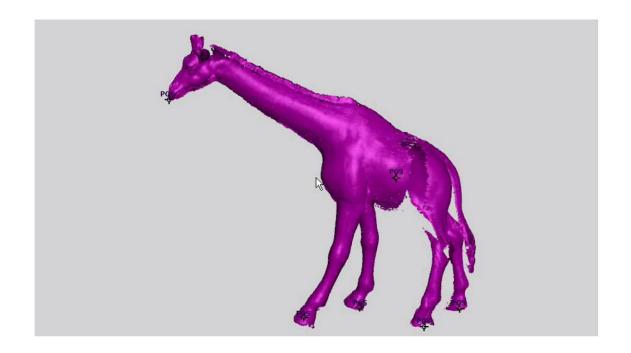
Motivation

Easy modeling – generate new shapes by deforming existing ones

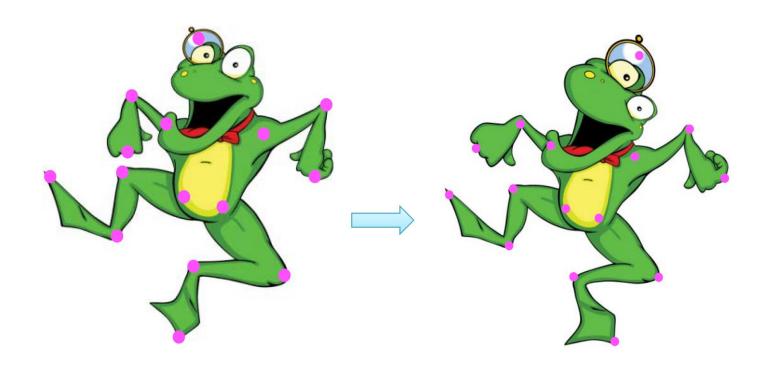


Motivation

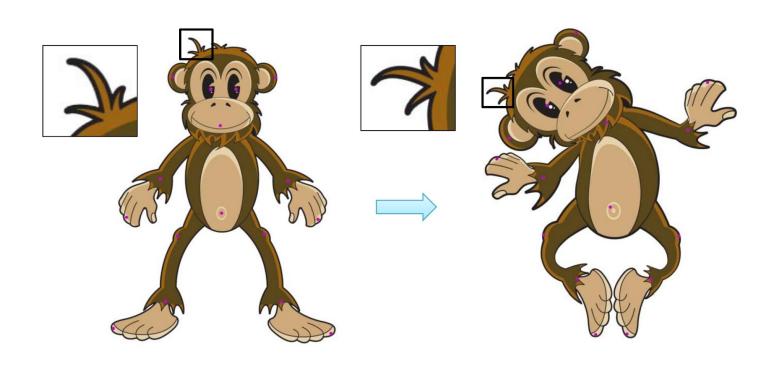
Character posing for animation



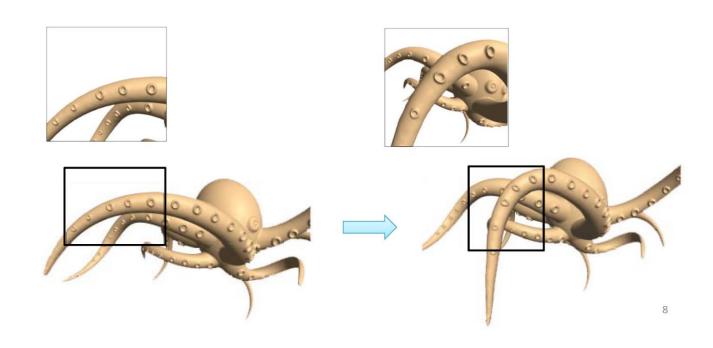
User says as little as possible, and algorithm deduces the rest



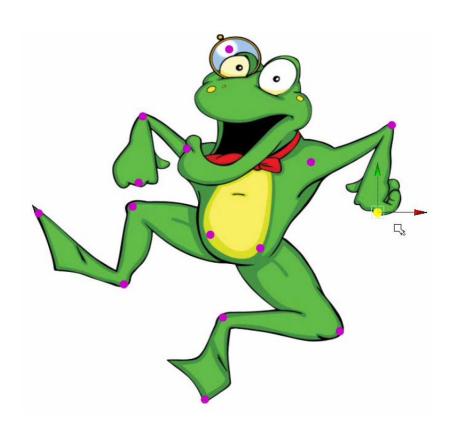
"Intuitive deformation" global change + local detail preservation



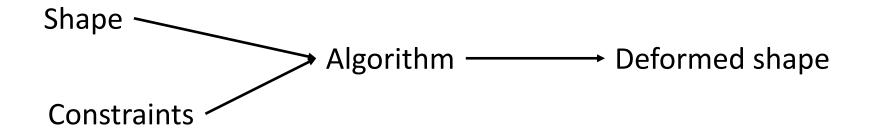
"Intuitive deformation" global change + local detail preservation



Efficient!



Problem Statement



Position

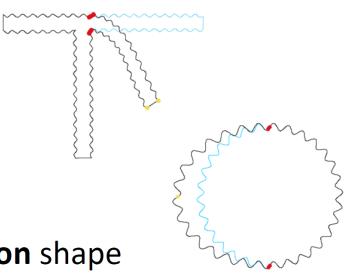
Orientation/Scale

Other shape property

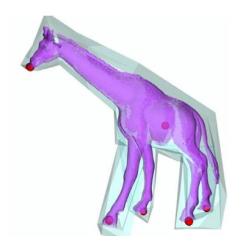
- Surface deformation
 - Shape is empty shell
 - Curve for 2D deformation
 - Surface for 3D deformation

Deformation only defined on shape

Deformation coupled with shape representation



- Space deformation
 - Shape is volumetric
 - Planar domain in 2D
 - Polyhedral domain in 3D



- Deformation defined in neighborhood of shape
- Can be applied to any shape representation

- Surface deformation
 - Find alternative representation which is "deformation invariant"

- Space deformation
 - Find a space map which has "nice properties"

- Surface deformation
 - Find alternative representation which is "deformation invariant"

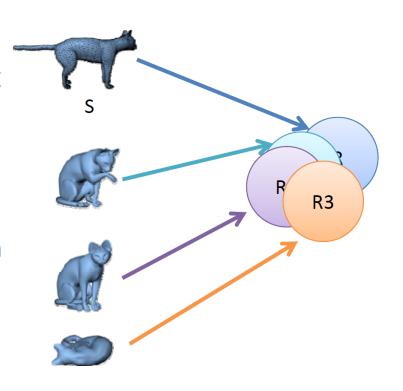
- Space deformation
 - Find a space map which has "nice properties"

Surface Deformation

Setup:

- Choose alternative representation f(S)

- Given S find S' such that
 - Constraints(S') are true
 - f(S') = f(S) (or close)
 - An optimization problem



Shape Representation

Robustness

- How hard is it to solve the optimization problem?
- Can we find the global minimum?
- Small change in constraints → similar shape?

Efficiency

— Can it be solved at interactive rates?

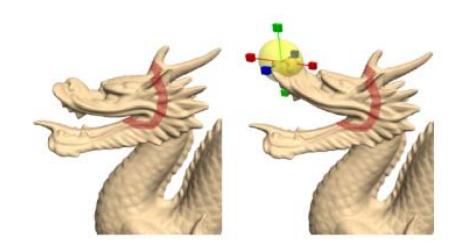
Surface Representations

- Laplacian coordinates
- Edge lengths + dihedral angles
- Pyramid coordinates
- Local frames

•

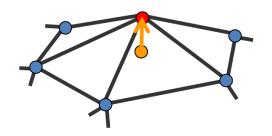
Laplacian Coordinates [Sorkine et al. 04]

- Control mechanism
 - Handles (vertices) moved by user
 - Region of influence (ROI)



Movie

Laplacian Coordinates



$$\boldsymbol{\delta}_i = \mathbf{v}_i - \sum_{j \in N(i)} \frac{1}{d_i} \mathbf{v}_j = \sum_{j \in N(i)} \frac{1}{d_i} (\mathbf{v}_i - \mathbf{v}_j)$$

$$\delta = LV = (I - D^{-1}A)V$$

I = Identity matrix

 \boldsymbol{D} = Diagonal matrix $[d_{ii} = deg(\boldsymbol{v}_i)]$

A = Adjacency matrix

V = Vertices in mesh

Approximation to normals - unique up to translation

Reconstruct by solving $LV = \delta$ for V, with one constraint

$$LV = \delta$$

Poisson equation

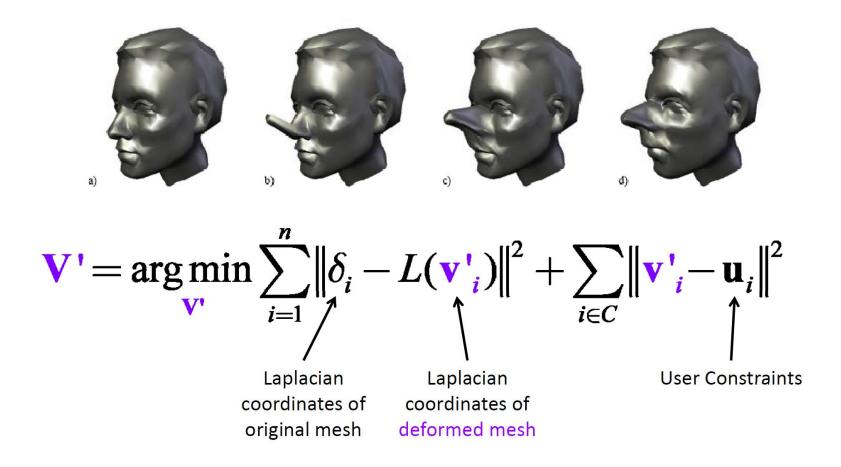
Deformation

• Pose modeling constraints for vertices $C \subset V$ $-v'_{i} = u_{i} \ i \in C$

No exact solution, minimize error

$$\begin{aligned} \mathbf{V'} &= \underset{\mathbf{v'}}{\operatorname{arg\,min}} \sum_{i=1}^{n} \left\| \delta_i - L(\mathbf{v'_i}) \right\|^2 + \sum_{i \in C} \left\| \mathbf{v'_i} - \mathbf{u_i} \right\|^2 \\ & \underset{\text{coordinates of original mesh}}{\operatorname{Laplacian}} & \underset{\text{coordinates of original mesh}}{\operatorname{Laplacian}} & \underset{\text{coordinates of deformed mesh}}{\operatorname{Laplacian}} & \underset{\text{coordinates of deformed mesh}}{\operatorname{Laplacian}} \end{aligned}$$

Deformation



Laplacian Coordinates

Sanity Check

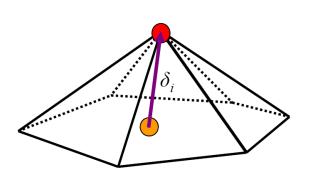
Translation invariant?

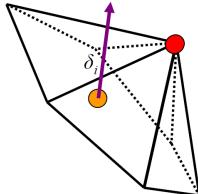


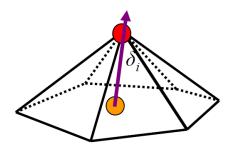
$$\delta_i = L(\mathbf{v}_i) = L(\mathbf{v}_i + \mathbf{t}) \quad \forall \mathbf{t} \in \mathbb{R}^3$$

Rotation/scale invariant?

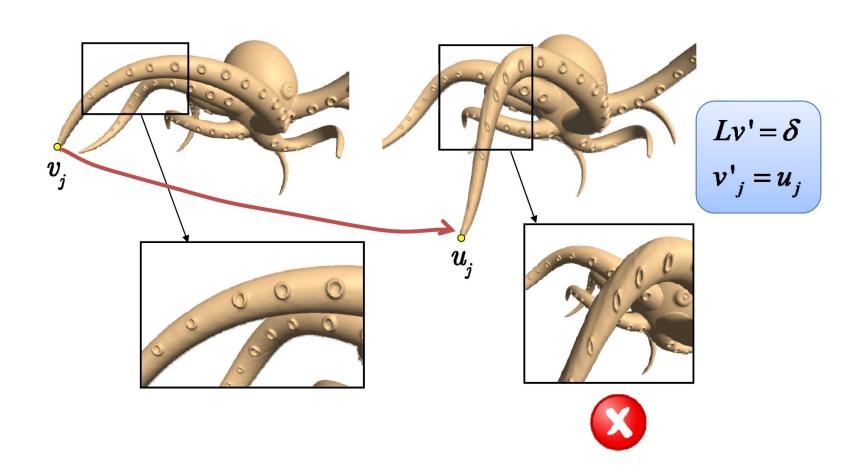


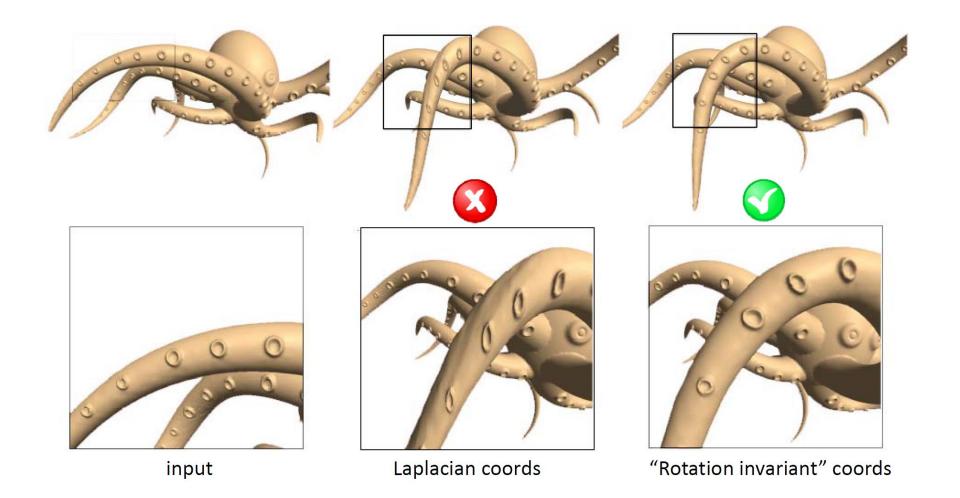






Problem

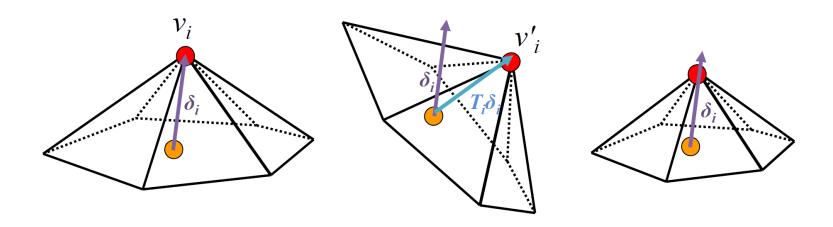




"Rotation Invariant" Coords

The representation should take into account local rotations + scale

$$\delta_i = L(v_i)$$
 $T_i \delta_i = L(v_i)$



Solution: Implicit Transformations

<u>Idea:</u> solve for local transformation and deformed surface simultaneously

$$V' = \arg\min_{v'} \left(\sum_{i=1}^{n} \left\| L(\mathbf{v}_i') - \mathbf{T}_i(\boldsymbol{\delta}_i) \right\|^2 + \sum_{j \in C} \left\| \mathbf{v}_j' - \mathbf{u}_j \right\|^2 \right)$$

$$\text{Transformation of the local frame }$$

Similarities

Restrict T_i to "good" transformations = rotation + scale \rightarrow similarity transformation

$$V' = \arg\min_{v'} \left(\sum_{i=1}^{n} ||L(\mathbf{v}_i')| - \left(\sum_{i=1}^{n} (\mathbf{\delta}_i) ||^2 + \sum_{j \in C} ||\mathbf{v}_j' - \mathbf{u}_j||^2 \right)$$
Similarity Transformation

Similarities

• Conditions on T_i to be a similarity matrix?

• Linear in 2D:

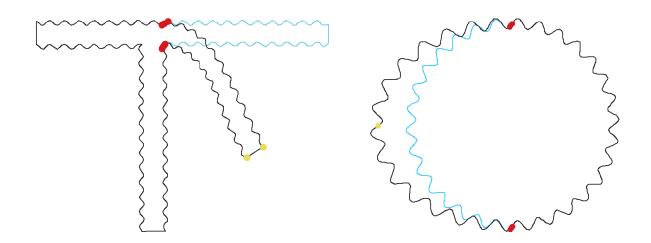
Auxiliary variables

$$T_{i} = \begin{pmatrix} s & 0 & 0 \\ 0 & s & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \cos \theta & \sin \theta & d_{x} \\ -\sin \theta & \cos \theta & d_{y} \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} w & a & t_{x} \\ -a & w & t_{y} \\ 0 & 0 & 1 \end{pmatrix}$$

Uniform scale

Rotation + translation

Similarities 2D



Similarities – 3D case

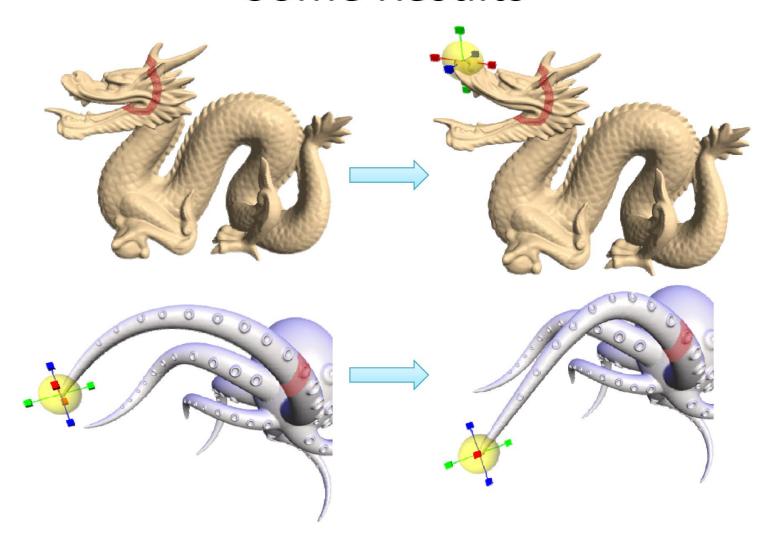
Not linear in 3D:

$$\begin{pmatrix}
\text{rotation} + \\
\text{uniform scale}
\end{pmatrix} = s \exp H = s \left(\alpha I + \beta H + \mathbf{h}^T \mathbf{h}\right)$$

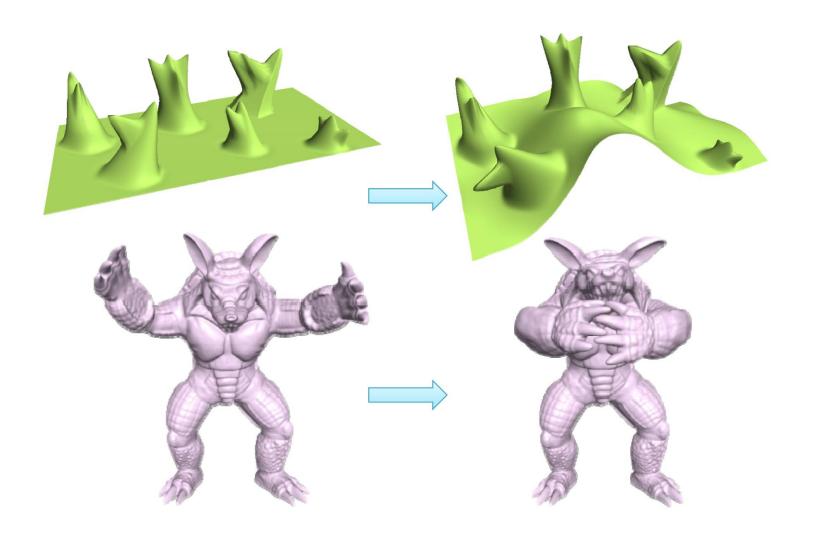
$$H \text{ is } 3 \times 3 \text{ skew-symmetric, } H\mathbf{x} = \mathbf{h} \times \mathbf{x}$$

- Linearize by dropping the quadratic term
 - Effectively: only small rotations are handled

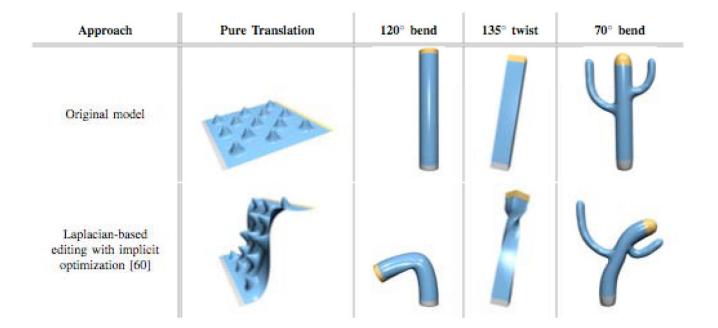
Some Results



Some Results



Limitations: Large Rotations



How to Find the Rotations?

- Laplacian coordinates solve for them
 - Problem: not linear

Another approach: propagate rotations from handles

Rotation Propagation

- Compute handle's "deformation gradient"
- Extract rotation and scale/shear components
- Propagate damped rotations over ROI

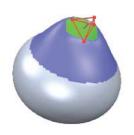




Deformation Gradient

Handle has been transformed affinely

$$\mathbf{T}(\mathbf{x}) = \mathbf{A}\mathbf{x} + \mathbf{t}$$



Deformation gradient is:

$$\nabla \mathbf{T}(\mathbf{x}) = \mathbf{A}$$

Extract rotation R and scale/shear S

$$\mathbf{A} = \mathbf{U} \mathbf{\Sigma} \mathbf{V}^T \quad \Rightarrow \quad \mathbf{R} = \mathbf{U} \mathbf{V}^T, \ \mathbf{S} = \mathbf{V} \mathbf{\Sigma} \mathbf{V}^T$$

Smooth Propagation

- Construct smooth scalar field [0,1]
 - $-\alpha(\mathbf{x})=1$ Full deformation (handle)
 - $-\alpha(\mathbf{x})=0$ No deformation (fixed part)
 - $-\alpha(\mathbf{x}) \in [0,1]$ Damp transformation (in between)
- Linearly damp scale/shear:

$$S(x) = \alpha(x)S(handle)$$

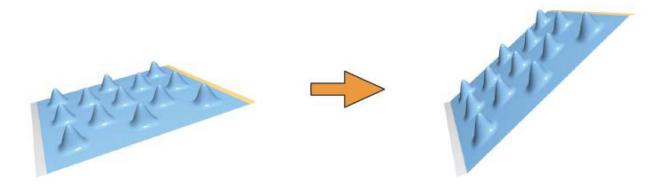
• Log scale damp rotation:

$$\mathbf{R}(\mathbf{x}) = \exp(\alpha(\mathbf{x})\log(\mathbf{R}(handle))$$

Limitations

Works well for rotations

- Translations don't change deformation gradient
 - "Translation insensitivity"



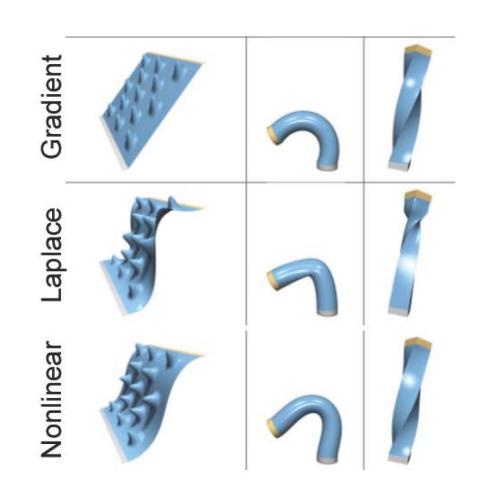
The Curse of Rotations

- Can't solve for them directly using a linear system
- Can't propagate if the handles don't rotate

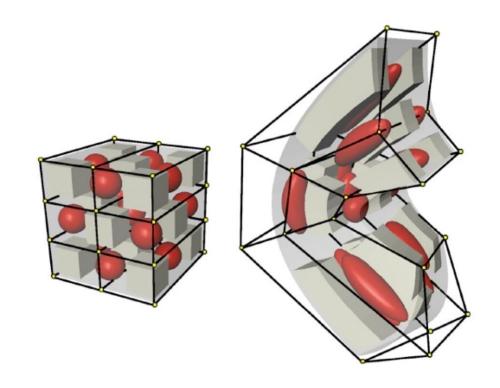
- Some linear methods work for rotations
- Some work for translations
- None work for both

The Curse of Rotations

- Non linear methods work for both large rotations and translation only
- No free lunch: much more expensive



- Deform object's bounding box
 - Implicitly deforms embedded objects



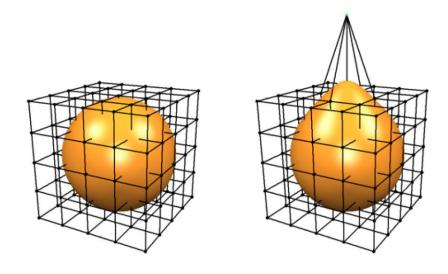
- Deform object's bounding box
 - Implicitly deforms embedded objects

Tri-variate tensor-product spline

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

- Deform object's bounding box
 - Implicitly deforms embedded objects

Tri-variate tensor-product spline



- Deform object's bounding box
 - Implicitly deforms embedded objects

- Tri-variate tensor-product spline
 - Aliasing artifacts

- Interpolate deformation constraints
 - Only in least squares sense



