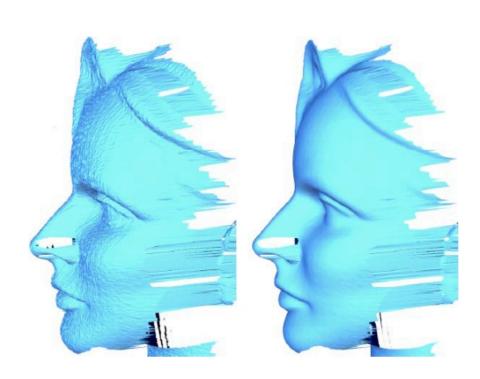
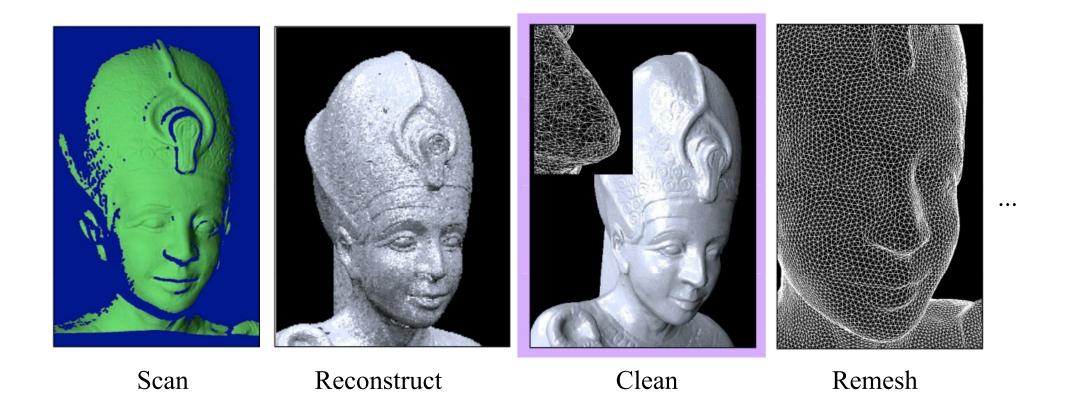
Mesh Smoothing

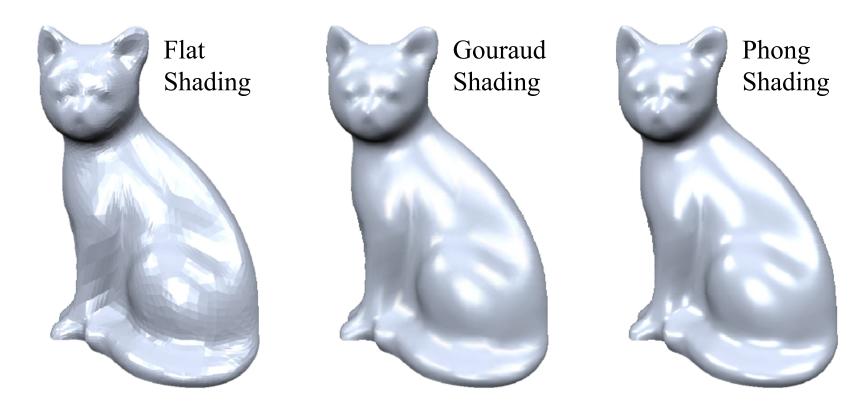


Mesh Processing Pipeline



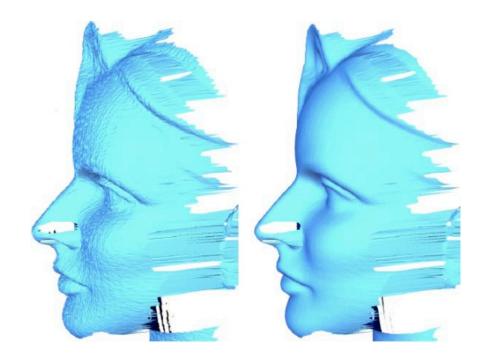
Mesh Quality

Visual inspection of "sensitive" attributes Specular shading



Motivation

Filter out high frequency noise



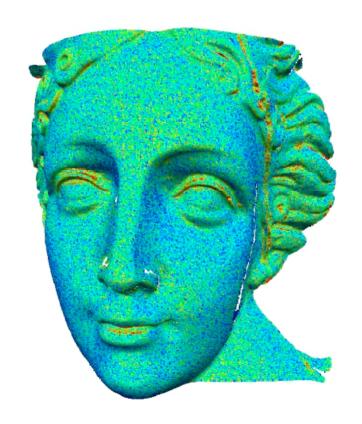
Mesh Smoothing

(aka Denoising, Filtering, Fairing)

Input: Noisy mesh (scanned or other)

Outut: Smooth mesh

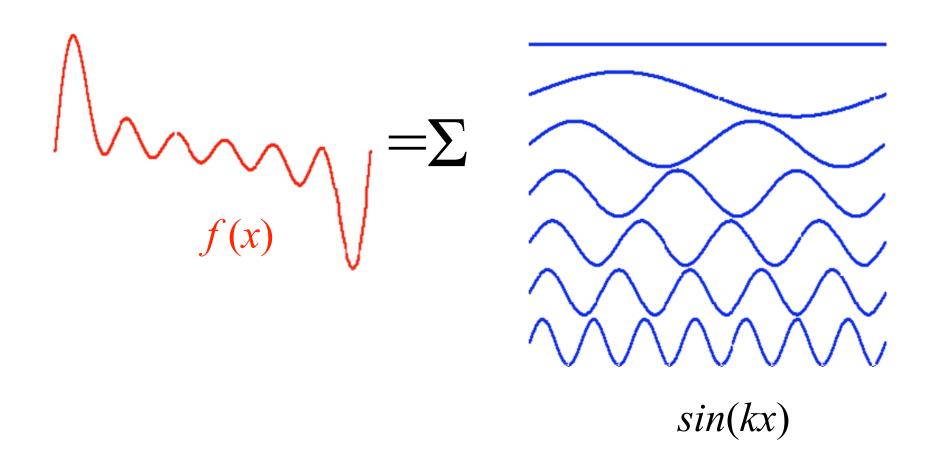
How: Filter out high frequency noise





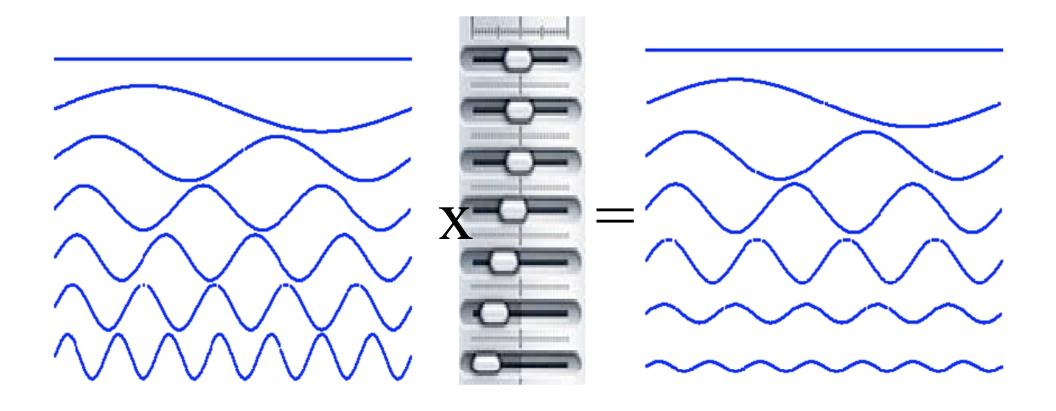
Smoothing Filtering

Fourier Transform



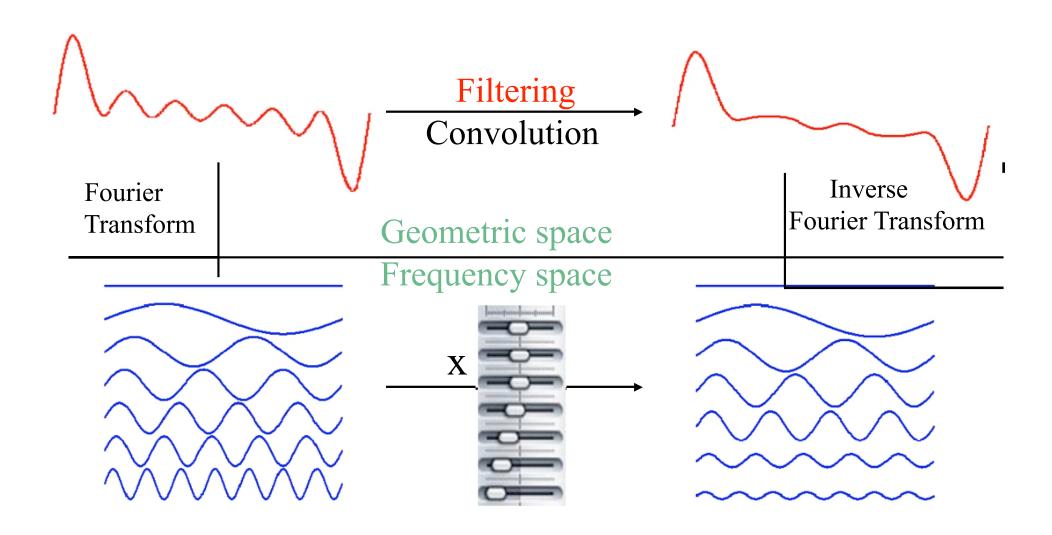
Smoothing Filtering

Fourier Transform

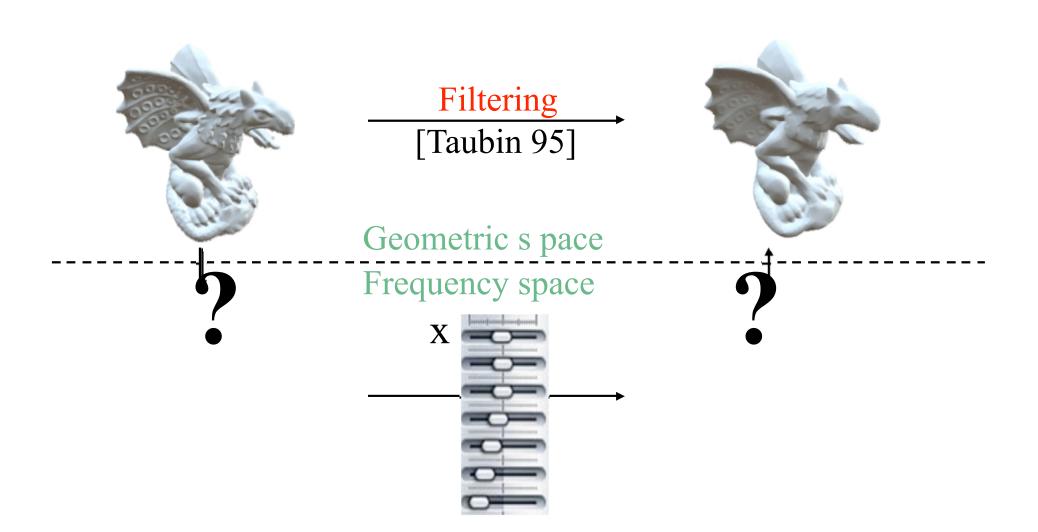


Smoothing Filtering

Fourier Transform

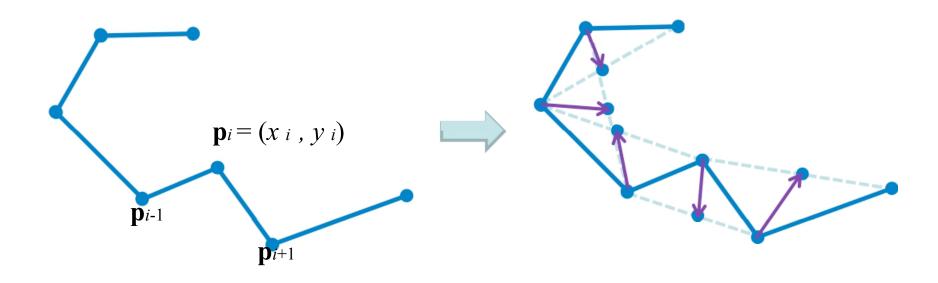


Filtering on a Mesh



Laplacian Smoothing

An easier problem: How to smooth a curve?

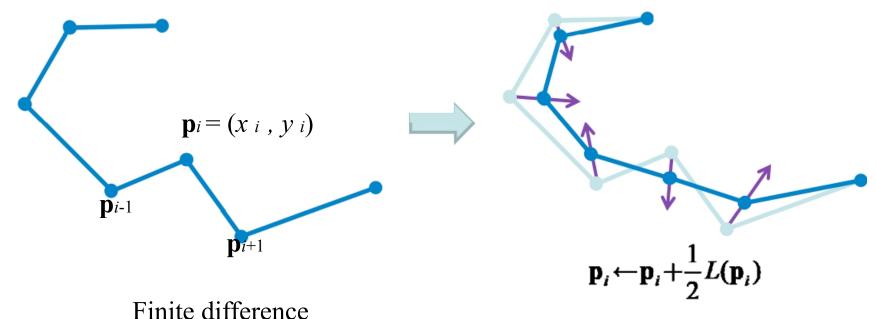


$$(\mathbf{p}_{i-1} + \mathbf{p}_{i+1})/2 - \mathbf{p}_i$$

$$L(\mathbf{p}_{i}) = \frac{1}{2} (\mathbf{p}_{i+1} - \mathbf{p}_{i}) + \frac{1}{2} (\mathbf{p}_{i-1} - \mathbf{p}_{i})$$

Laplacian Smoothing

An easier problem: How to smooth a curve?



discretization of second derivative = Laplace operator in one dimension

Laplacian Smoothing

Algorithm:

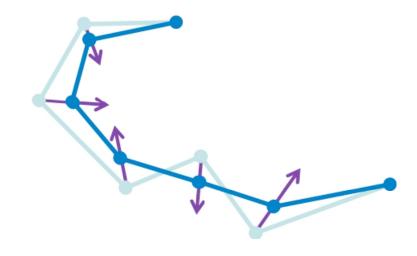
Repeat for *m* iterations (for non boundary points):

$$\mathbf{p}_i \leftarrow \mathbf{p}_i + \lambda L(\mathbf{p}_i)$$

For which λ ?

$$0 < \lambda < 1$$

Closed curve conver ges to?
Single point



Spectral Analysis

Closed Curve

Re-write
$$\mathbf{p}_{i}^{(t+1)} = \mathbf{p}_{i}^{(t)} + \lambda L(\mathbf{p}_{i}^{(t)})$$

$$L(\mathbf{p}_{i}) = \frac{1}{2}(\mathbf{p}_{i+1} - \mathbf{p}_{i}) + \frac{1}{2}(\mathbf{p}_{i-1} - \mathbf{p}_{i})$$

in matrix notation: $\mathbf{P}^{(t+1)} = \mathbf{P}^{(t)} - \lambda \mathbf{L} \mathbf{P}^{(t)}$

$$\mathbf{P} = \begin{pmatrix} x_1 & y_2 \\ \dots & \dots \\ x_n & y_n \end{pmatrix} \in \mathbb{R}^{n \times 2} \quad \mathbf{L} = \frac{1}{2} \begin{pmatrix} 2 & -1 & & & -1 \\ -1 & 2 & -1 & & & \\ & & \dots & & -1 & 2 & -1 \\ -1 & & & & -1 & 2 \end{pmatrix} \in \mathbb{R}^{n \times n}$$

The Eigen vectors of L

$$\mathbf{L} = \mathbf{V} \mathbf{D} \mathbf{V}^{T} \qquad \begin{pmatrix} \begin{vmatrix} & & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\$$

Spectral Analysis

Then:
$$\mathbf{P}^{(t+1)} = \mathbf{P}^{(t)} - \lambda \mathbf{L} \mathbf{P}^{(t)} = (\mathbf{I} - \lambda \mathbf{L}) \mathbf{P}^{(t)}$$

After *m* iterations:
$$\mathbf{P}^{(m)} = (\mathbf{I} - \lambda \mathbf{L})^m \mathbf{P}^{(0)}$$

Can be described using eigendecomposition of L

$$\begin{array}{c}
\mathbf{L} = \mathbf{V} \mathbf{D} \mathbf{V}^{T} \\
\mathbf{V} = \begin{bmatrix} 1 & 1 & 1 \\ \mathbf{v}_{1} & \mathbf{v}_{2} & \dots & \mathbf{v}_{n} \\ 1 & 1 & 1 \end{bmatrix}, \mathbf{D} = \begin{bmatrix} k_{1} & k_{2} & \dots & k_{n} \\ k_{2} & \dots & k_{n} \\ \dots & \dots & \dots \end{bmatrix} \quad \Rightarrow \quad \mathbf{P}^{(m)} = \mathbf{V} \underbrace{\left(\mathbf{I} - \lambda \mathbf{D}\right)^{m}}_{\mathbf{V}} \mathbf{V}^{T} \mathbf{P}^{(0)}$$

Filtering high

frequencies

Laplacian Smoothing on Meshes

Same as for curves:

$$\mathbf{p}_{i}^{(t+1)} = \mathbf{p}_{i}^{(t)} + \lambda \Delta \mathbf{p}_{i}^{(t)}$$

 $N = \{ k, l, m, n \}$ $\mathbf{p}_i = (x_i, y_i, z_i)$

What is $\Delta \mathbf{p} i$?

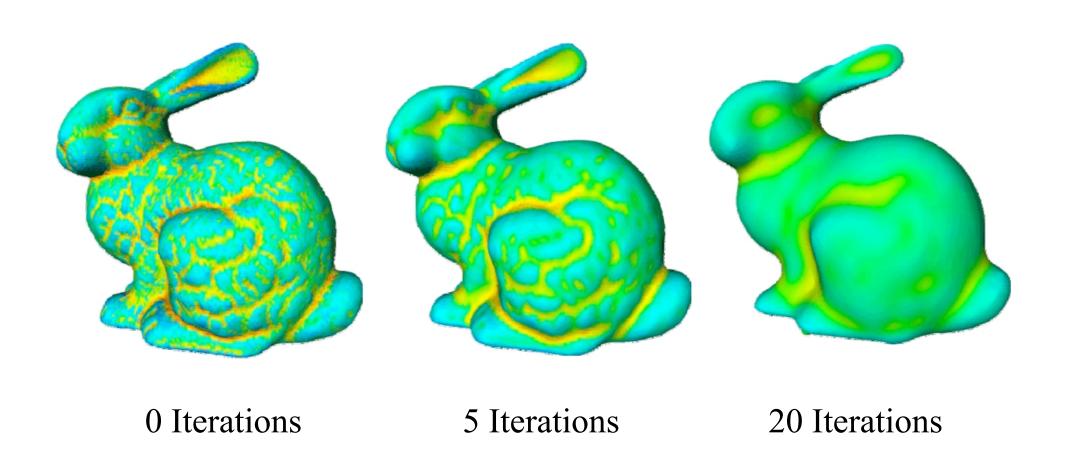


$$\mathbf{p}_{n}$$
 \mathbf{p}_{k}

$$\frac{1}{2}(\mathbf{p}_{i+1}+\mathbf{p}_{i-1})-\mathbf{p}_i \qquad \frac{1}{|N_i|}\left[\sum_{j\in N_i}\mathbf{p}_j\right]$$

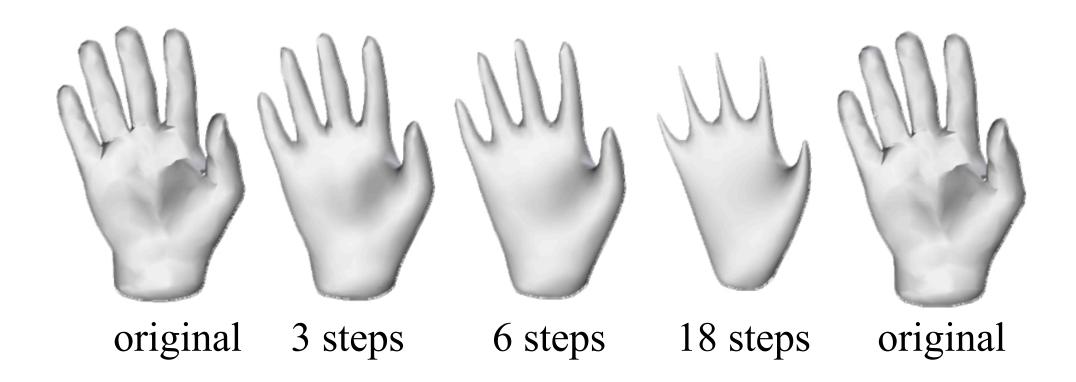
$$\frac{1}{|N_i|} \left(\sum_{j \in N_i} \mathbf{p}_j \right) - \mathbf{p}_i$$

Laplacian Smoothing on Meshes



Problem - Shrinkage

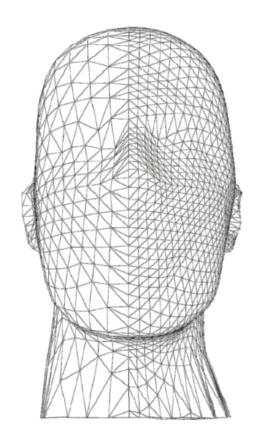
Repeated iterations of Laplacian smoothing shrinks the mesh

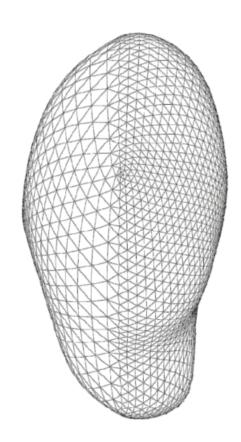


Lalace Operator Discretization

The Problem

Not good – The result should not depend on triangle si zes

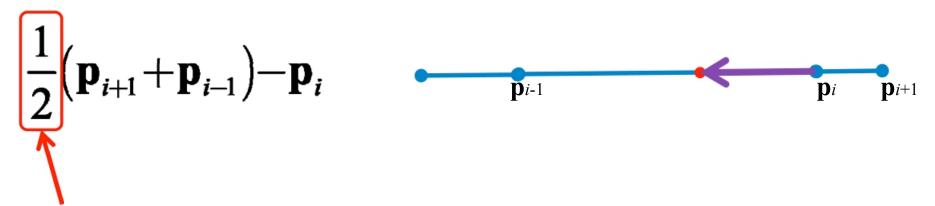




Laplace Operator Discretization

What Went Wrong?

Back to curves:



Same weight for both neighbors, although one is closer

Laplace Operator Discretization

The Solution

Use a weighted average to define Δ

$$w_{ij} = \frac{1}{l_{ij}} \qquad w_{ik} = \frac{1}{l_{ik}} \qquad L(\mathbf{p}_i) = \frac{w_{ij}\mathbf{p}_j + w_{ik}\mathbf{p}_k}{w_{ij} + w_{ik}} - \mathbf{p}_i$$

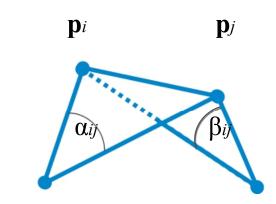
Straight curves will be invariant to smoothing

Laplace Operator Discretiztion

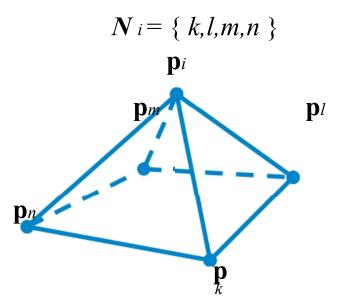
Cotangent Weights

Use a weighted average to define Δ

Which weights?



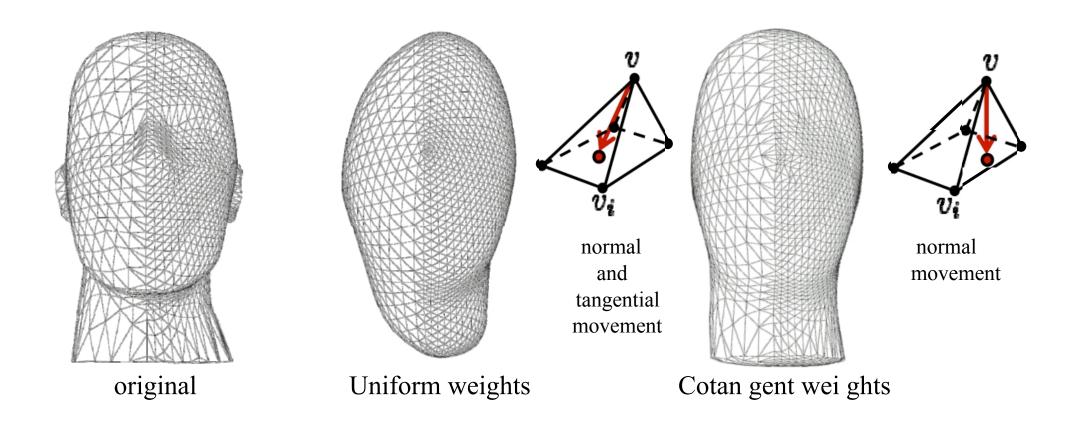
$$w_{ij} = \frac{h_{1+} h_{2}}{l_{ij}} = \frac{1}{2} \left(\cot \alpha_{ij} + \cot \beta_{ij} \right) \qquad L(\mathbf{p}) = \frac{1}{\sum_{i} w_{ij}} \sum_{j \in N_{i}} w_{j} \left(\mathbf{p}_{j} - \mathbf{p} \right)$$



$$L(\mathbf{p}) = \frac{1}{\sum_{ij} w_{ij}} \sum_{j \in N_i} w_j (\mathbf{p}_j - \mathbf{p})$$

Planar meshes will be invariant to smoothing

Smoothing with the Cotangent Laplacian



References

- "A Signal Processing Approach to Fair Surface Design", Taubin, Siggraph '95
- "Implicit Fairing of Irregular Meshes using Diffusion and Curvature Flow", Desbrun et al., Siggraph '99
- "An Intuitive Framework for Real -Time Freeform Modelin", Botsch et al., Siggraph'04
- "Spectral Geometry Processing with Manifold Harmonics", Vallet et al., E urograp hi cs '08