

G22.3033-004, Spring 2009

# Interactive Shape Modeling

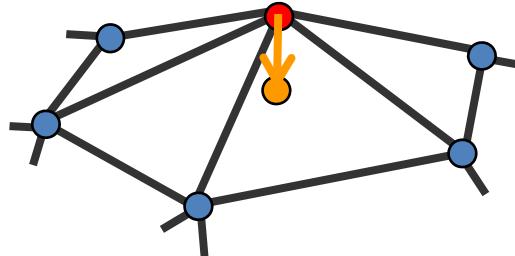
Digital Geometry Processing

$$\Delta_M \mathbf{p} = -H\mathbf{n}$$

# Recap

Laplace-Beltrami operator

- High-pass filter: extracts local surface detail
- Detail = *smooth*(surface) – surface
- Smoothing = averaging



First attempt at definition:  
uniform weighting

$$\delta_i = \frac{1}{d_i} \sum_{\mathbf{v}_j \in N_1(\mathbf{v}_i)} \mathbf{v}_j - \mathbf{v}_i$$

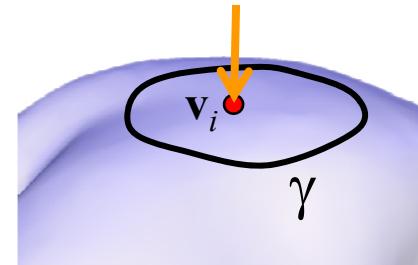
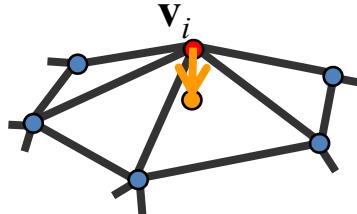
$$\delta_i = \sum_{\mathbf{v}_j \in N_1(\mathbf{v}_i)} \frac{1}{d_i} (\mathbf{v}_j - \mathbf{v}_i)$$

$$\Delta_M \mathbf{p} = -H\mathbf{n}$$

# Recap

Laplace-Beltrami operator

- The direction of  $\delta_i$  approximates the normal
- The size approximates the mean curvature



$$\delta_i = \frac{1}{d_i} \sum_{v_j \in N_1(v_i)} (v_j - v_i)$$

$$\delta_i = \frac{1}{len(\gamma)} \int_{s=a}^b (\gamma(s) - v_i) ds$$

$$\lim_{len(\gamma) \rightarrow 0} \frac{1}{len(\gamma)} \int_{s=a}^b (\gamma(s) - v_i) ds = -H(v_i) \mathbf{n}_i$$

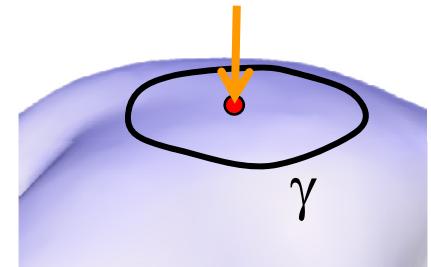
# Recap

## Discrete Laplace-Beltrami operator – weighting schemes

$$\delta_i = \frac{1}{A_i} \sum_{\mathbf{v}_j \in N_1(\mathbf{v}_i)} w_{ij} (\mathbf{v}_j - \mathbf{v}_i)$$

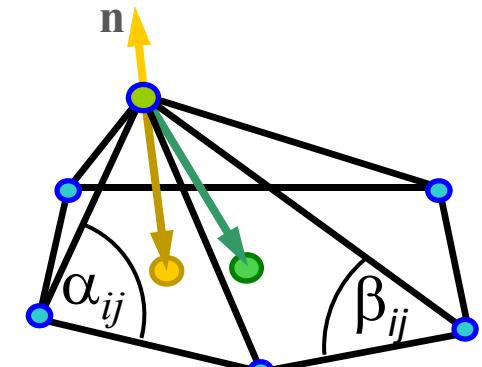
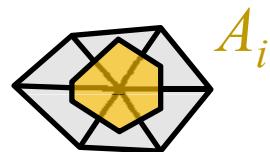
- Ignore geometry

$\delta_{\text{umbrella}} : A_i = 1, w_{ij} = 1/d_i$



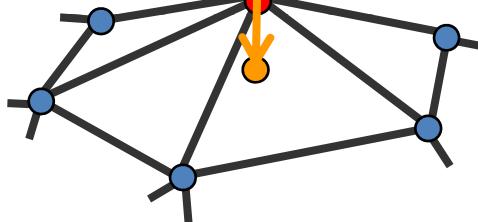
- Integrate over Voronoi region of the vertex

$\delta_{\text{cotangent}} : w_{ij} = \cot \alpha_{ij} + \cot \beta_{ij}$



# Laplacian matrix

- The transition between the  $\delta$  and  $xyz$  is linear:



$$\delta_i = \frac{1}{A_i} \sum_{v_j \in N_1(v_i)} w_{ij} (v_j - v_i)$$

$$\begin{array}{ccc} L & x & = \delta_x \\ L & y & = \delta_y \\ L & z & = \delta_z \end{array}$$

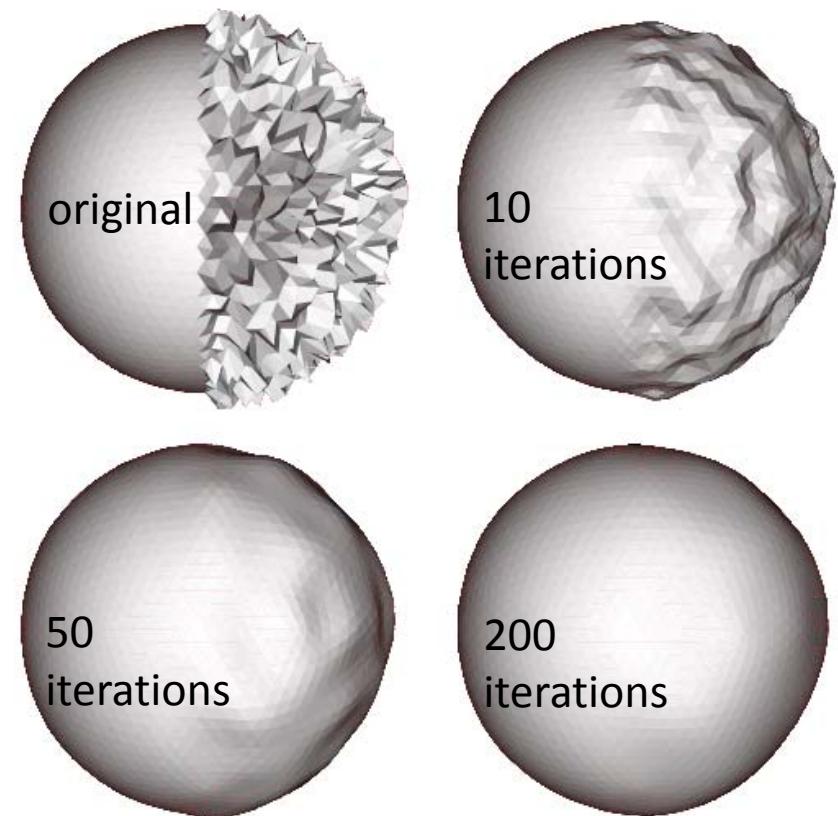
# Taubin smoothing

## Explicit steps

- Iterate:

$$\mathbf{x}' = \mathbf{x} + \lambda L\mathbf{x} = (I + \lambda L)\mathbf{x}$$

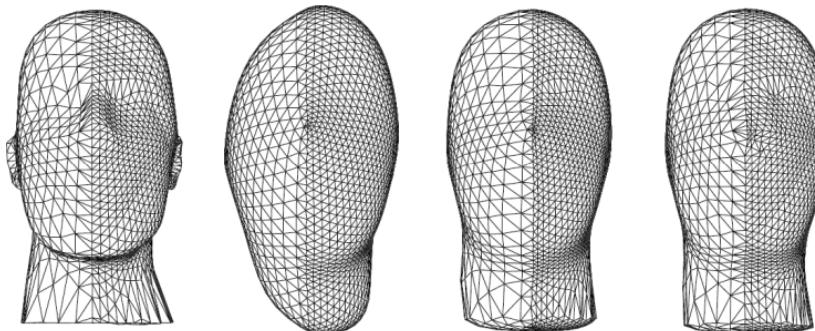
- $\lambda > 0$  to smooth;  
 $\lambda < 0$  to inflate
- Originally proposed  
with uniform Laplacian  
weights



# Implicit fairing

## Implicit Euler steps

- Use cotangent instead of uniform Laplacian



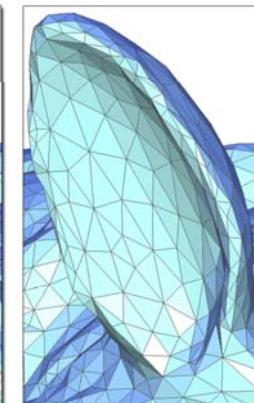
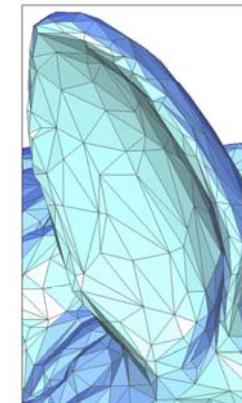
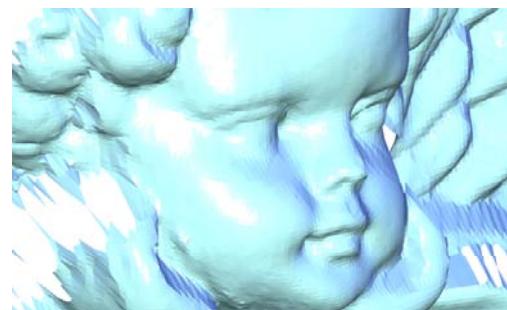
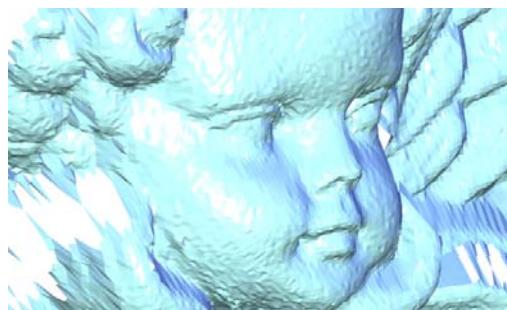
- In each iteration, solve for the smoothed  $\mathbf{x}'$ :

$$(I - \lambda L)\mathbf{x}' = \mathbf{x}$$

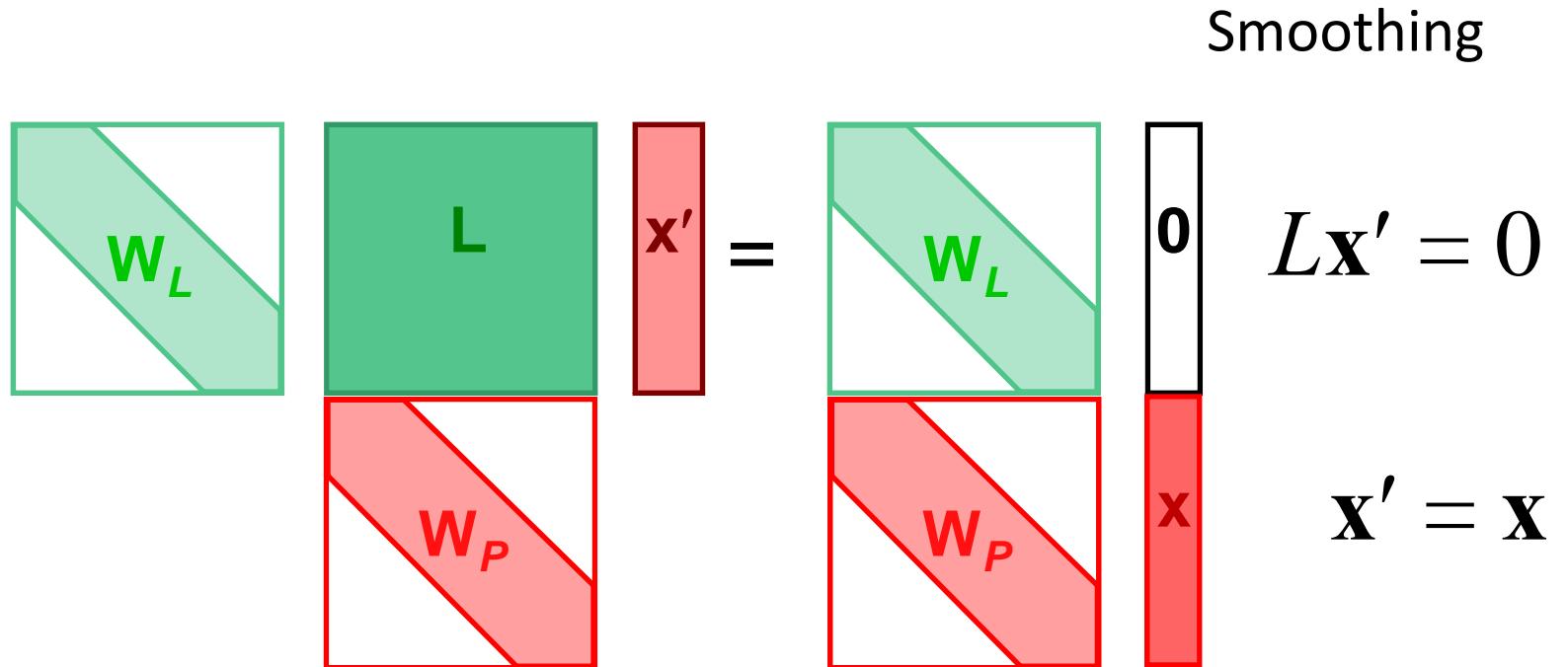
# Laplacian mesh optimization

[Nealen et al. 2006]

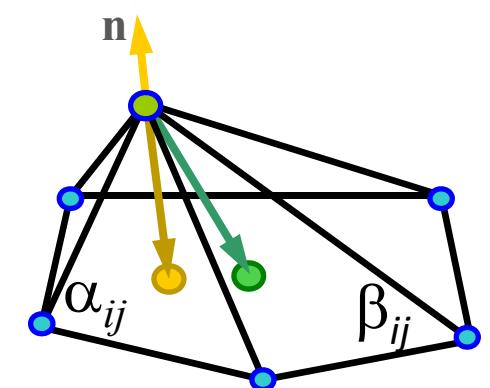
- Smoothing, improving of triangle shapes
- Basic idea: formulate a “shopping list”, solve with LS
  - Smooth mesh:  $L\mathbf{x}' = 0$
  - Passes close to input data set:  $\mathbf{x}' = \mathbf{x}$
  - Well-shaped triangles:  $L_{\text{uni}} \mathbf{x}' = L_{\text{cot}} \mathbf{x}$
  - The terms are weighted according to importance and geo.



# Laplacian mesh optimization

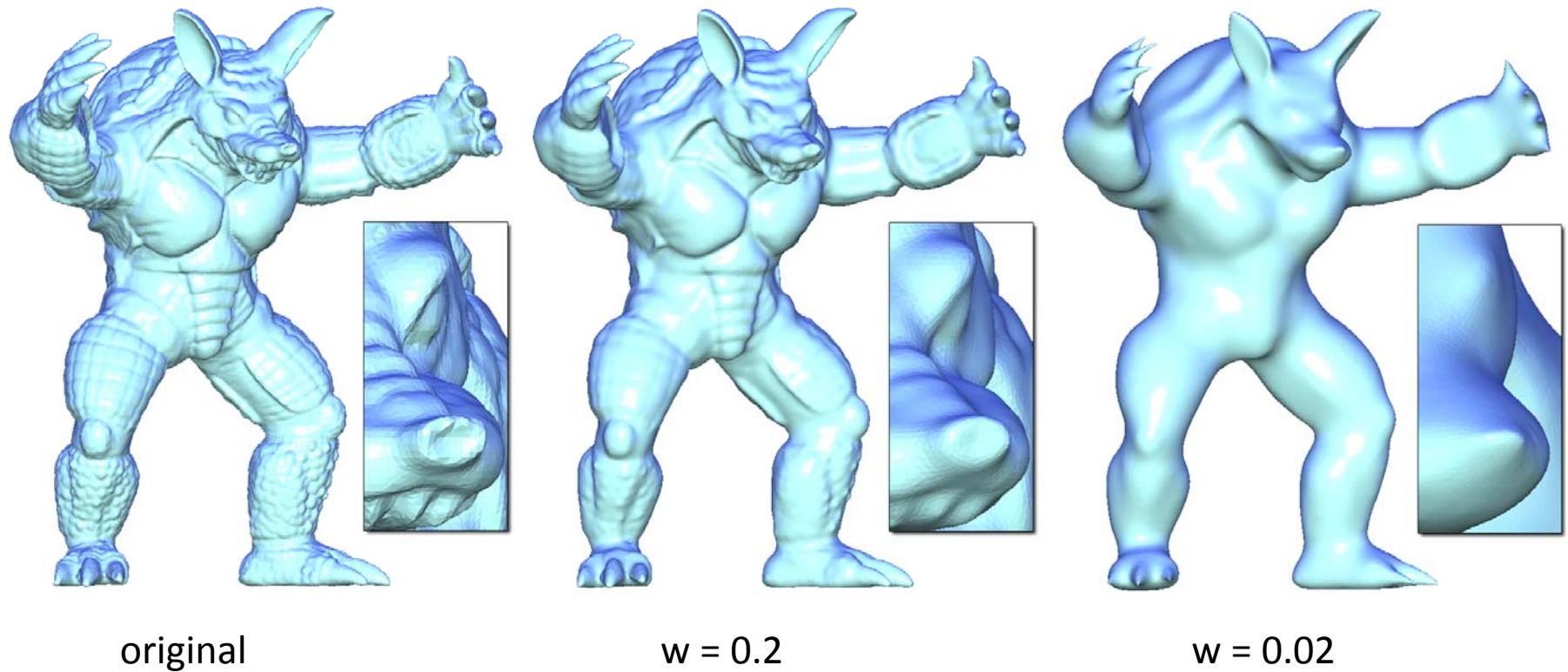


- **Mesh smoothing**  $L = L_{\text{cot}}$  (outer fairness) or  $L = L_{\text{uni}}$  (outer and inner fairness)
- Controlled by  $W_P$  and  $W_L$  (Intensity, Features)



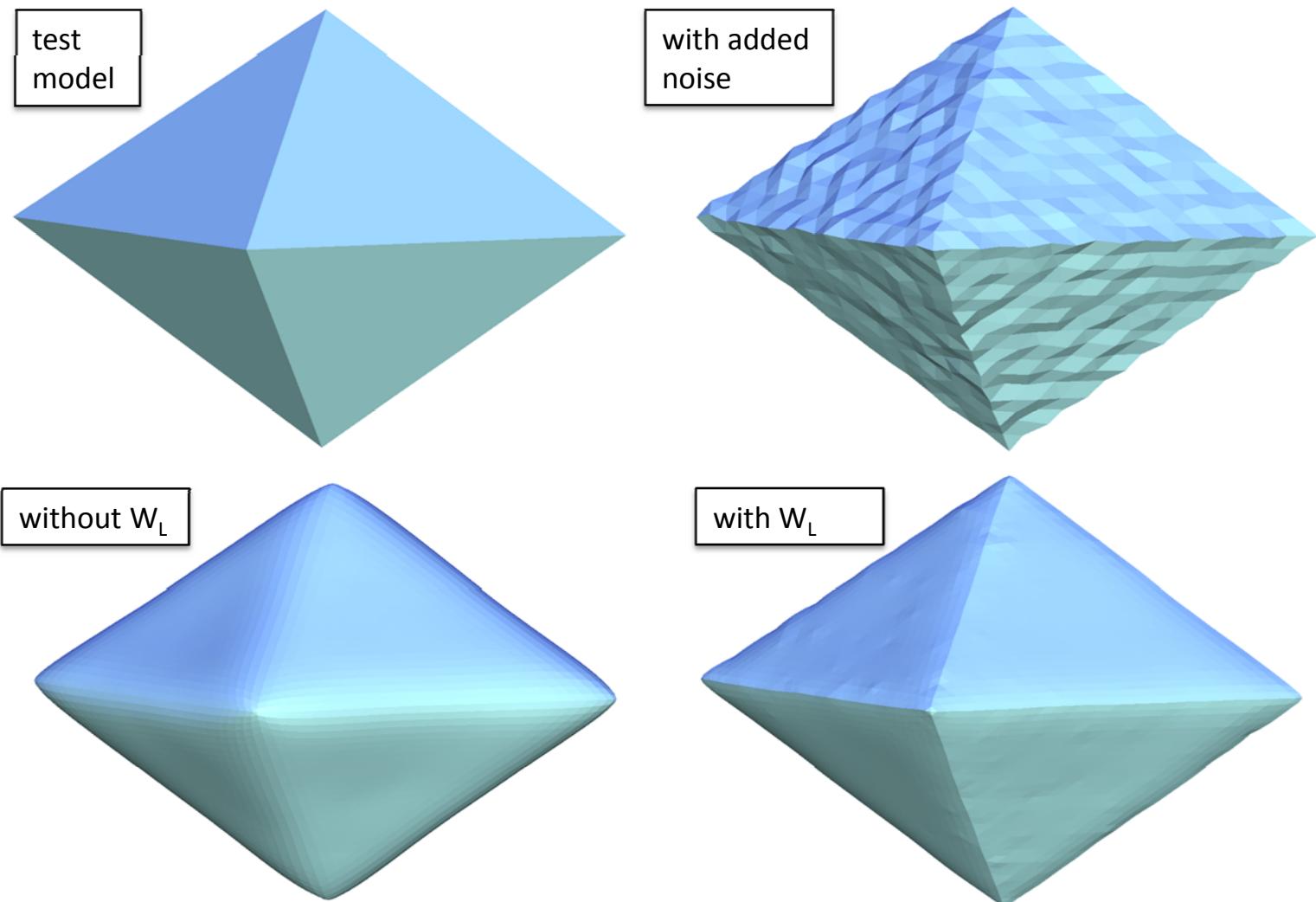
$$\begin{matrix} W_L \\ L \\ W_P \end{matrix} \quad x' = \begin{matrix} W_L \\ W_P \\ x \end{matrix} \quad \begin{matrix} 0 \\ \vdots \\ x \end{matrix}$$

Using  $W_P$



$$\begin{matrix} W_L \\ L \\ W_P \end{matrix} \quad x' = \begin{matrix} W_L \\ W_P \\ x \end{matrix} \quad \begin{matrix} 0 \\ \vdots \\ x \end{matrix}$$

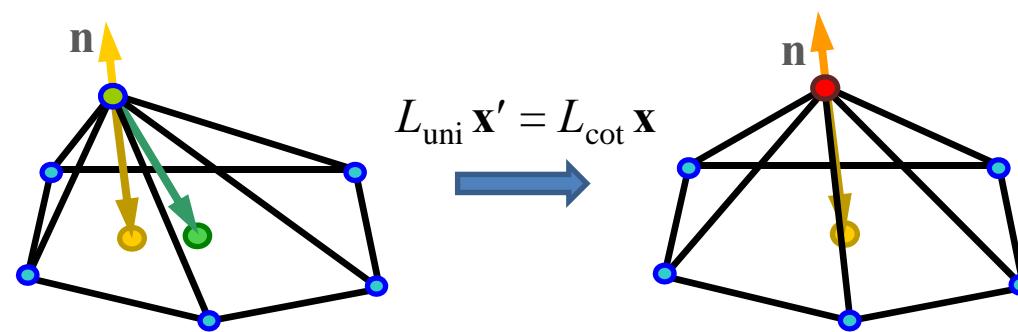
# Using $W_P$ and $W_L$



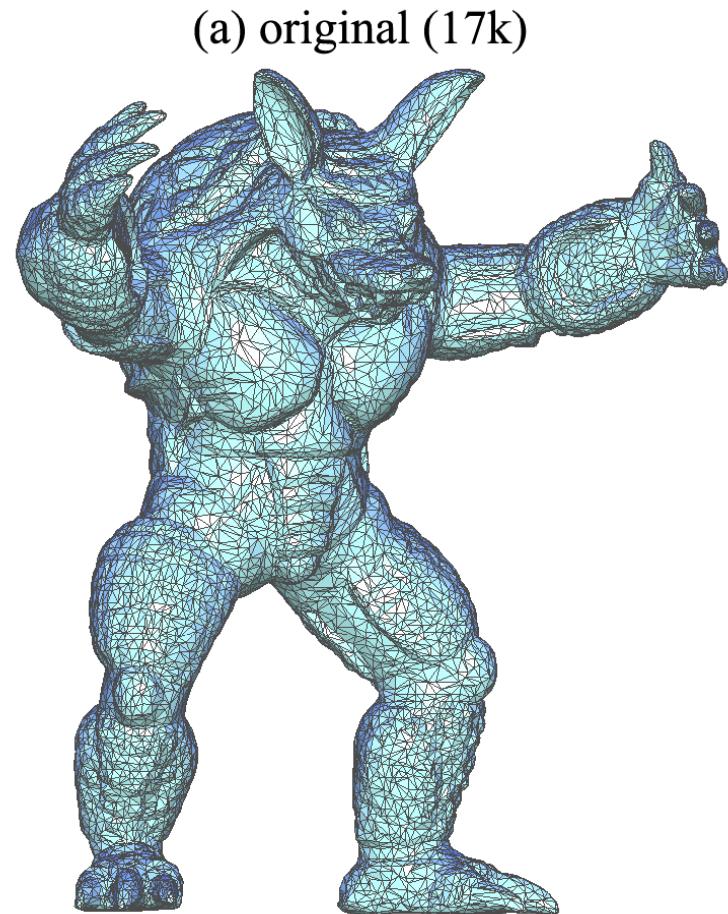
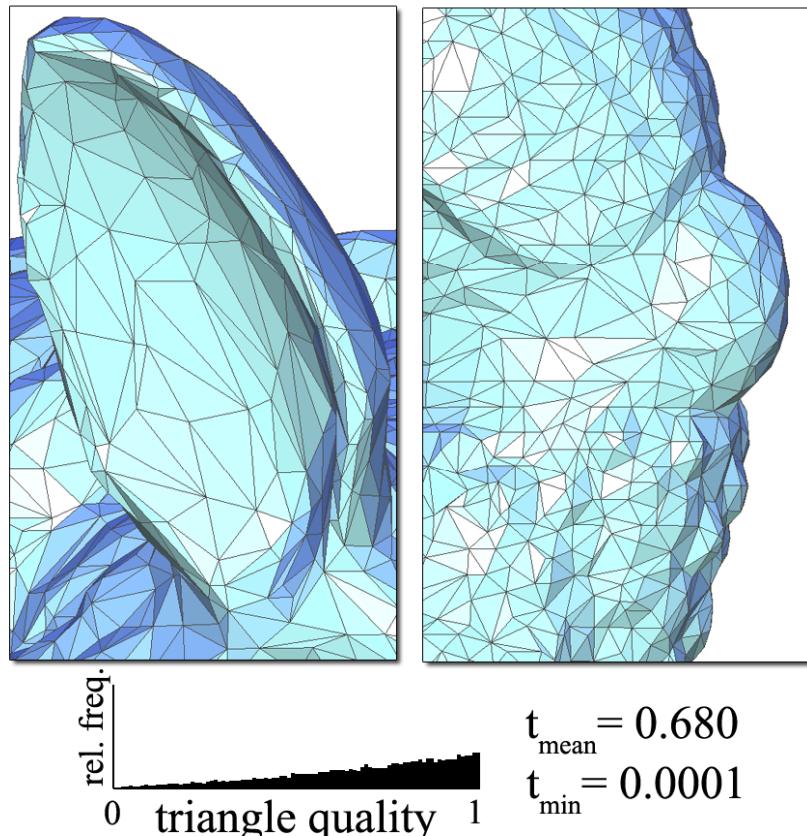
# Triangle shape Optimization

By global vertex relocation

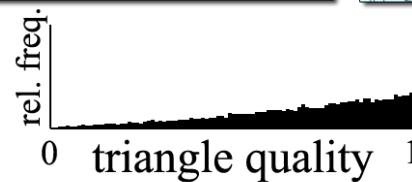
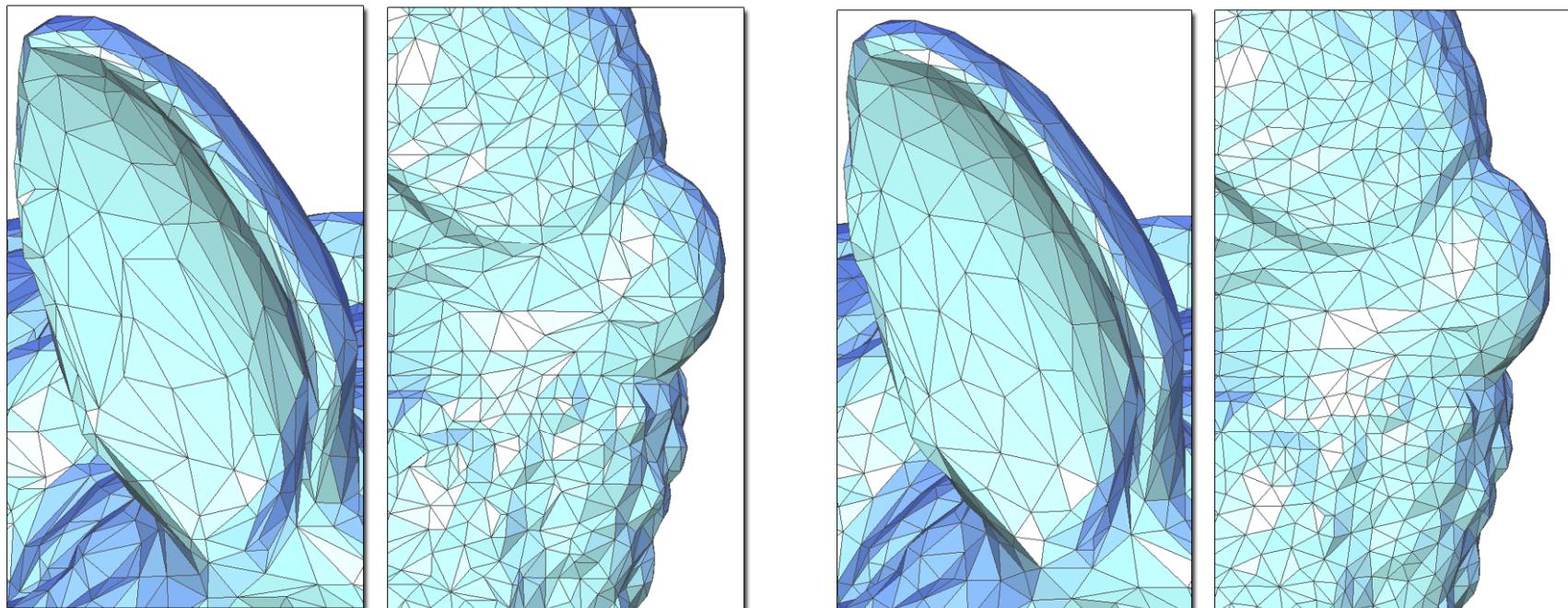
$$\begin{matrix} L_{\text{uni}} \\ W_P \end{matrix} \quad \begin{matrix} x' \\ = \end{matrix} \quad \begin{matrix} \delta_{\text{cot}} \\ x \end{matrix}$$



# Positional Weights



# Constant Weights



$$\begin{aligned} t_{\text{mean}} &= 0.680 \\ t_{\min} &= 0.0001 \end{aligned}$$

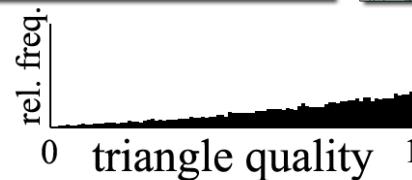
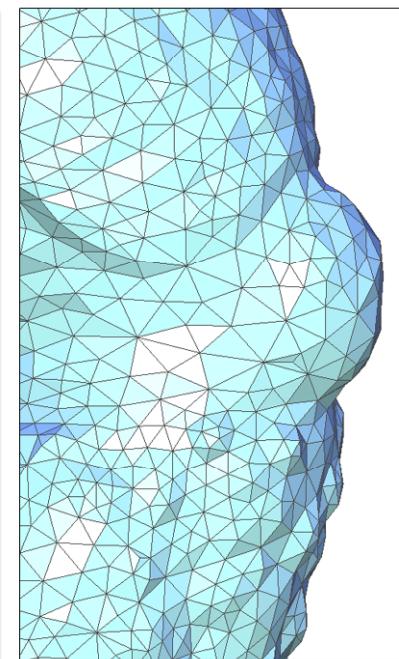
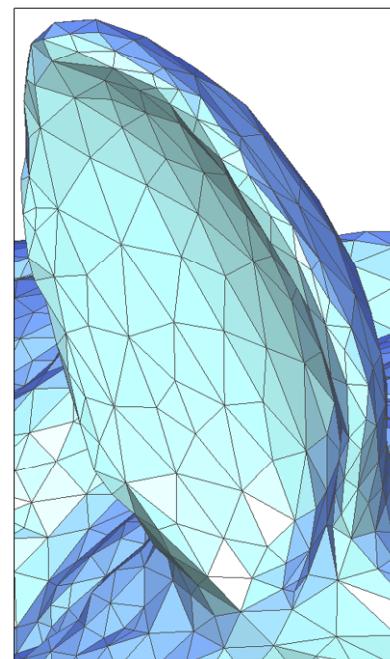
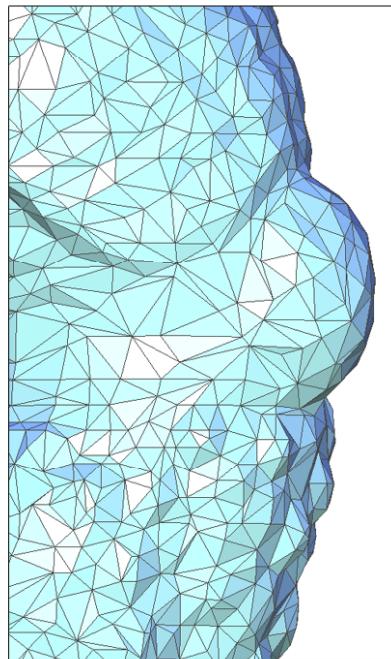
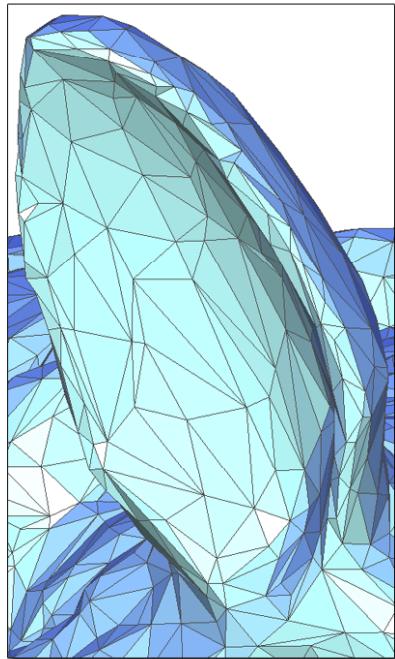


$$w_{\text{const}}(\kappa)$$

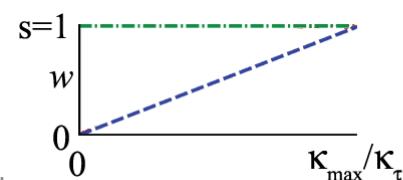
$$\text{dist} = 1.24 \cdot 10^{-3}$$

$$\begin{aligned} t_{\text{mean}} &= 0.791 \\ t_{\min} &= 0.024 \end{aligned}$$

# Linear Weights



$$t_{\text{mean}} = 0.680$$
$$t_{\min} = 0.0001$$

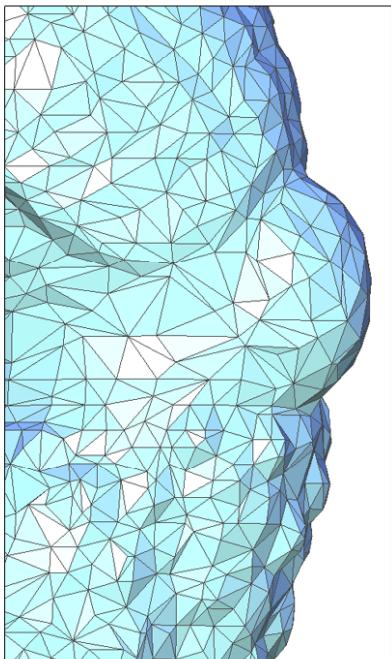
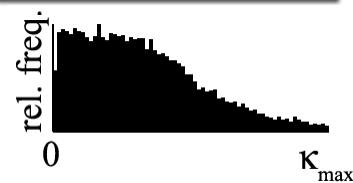
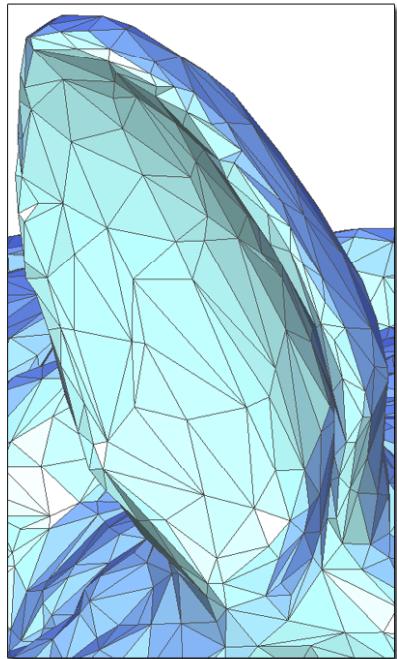


$$\text{dist} = 2.53 \cdot 10^{-3}$$

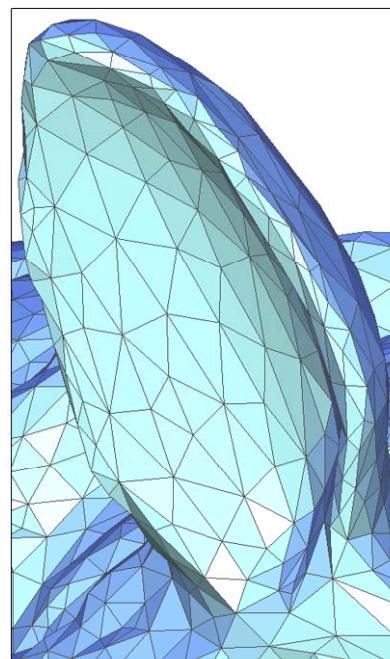


$$t_{\text{mean}} = 0.842$$
$$t_{\min} = 0.040$$

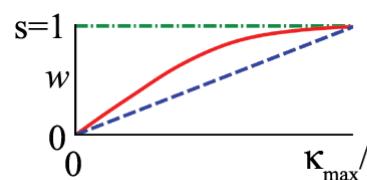
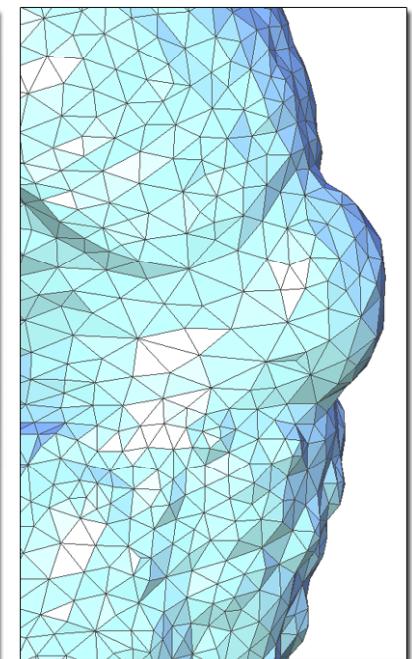
# CDF Weights



mean curvature  
distribution  $c(\kappa)$



$$\text{dist} = 2.04 \cdot 10^{-3}$$

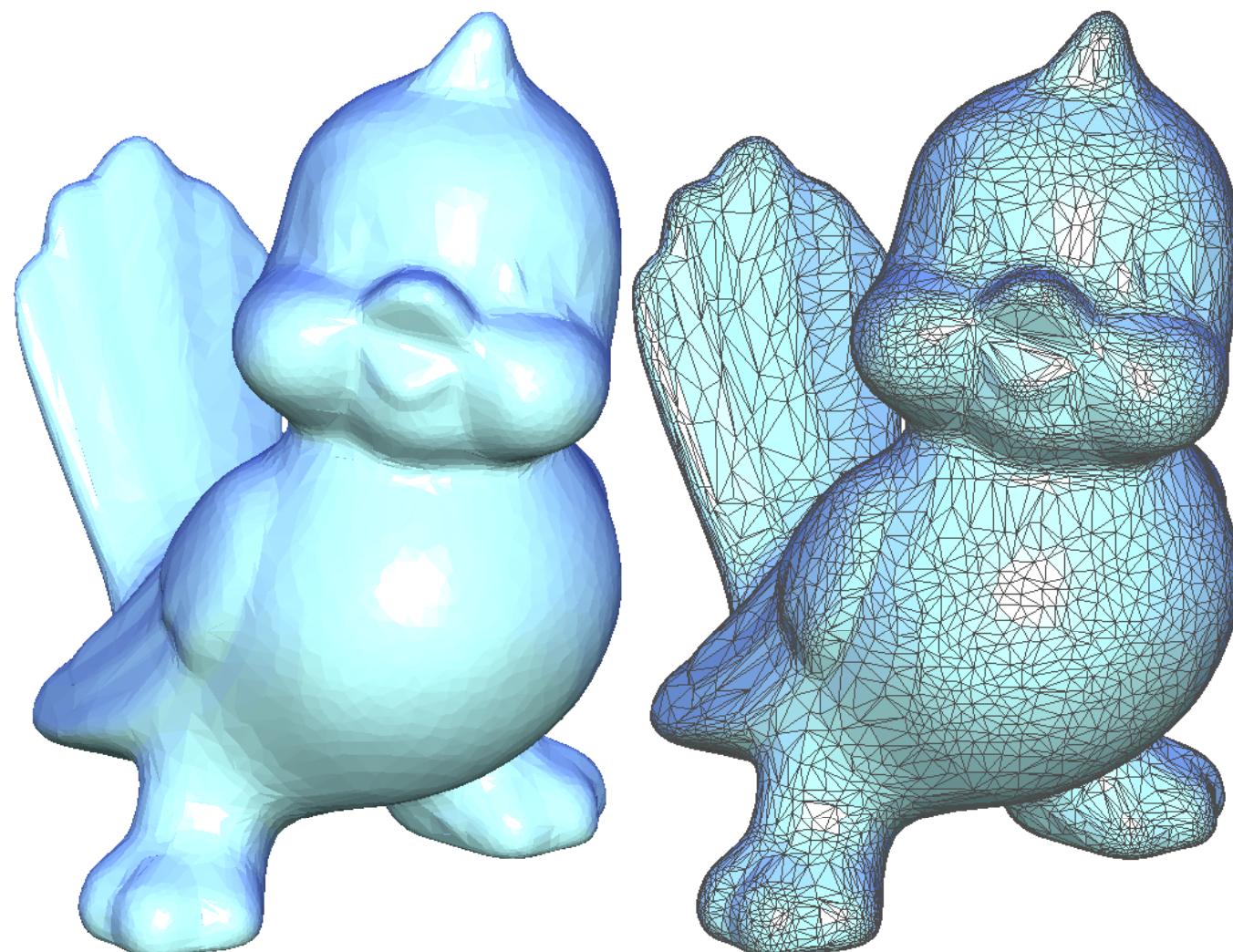


$w_{\text{const}}(\kappa)$   
 $w_{\text{linear}}(\kappa)$   
 $w_{\text{cdf}}(\kappa)$

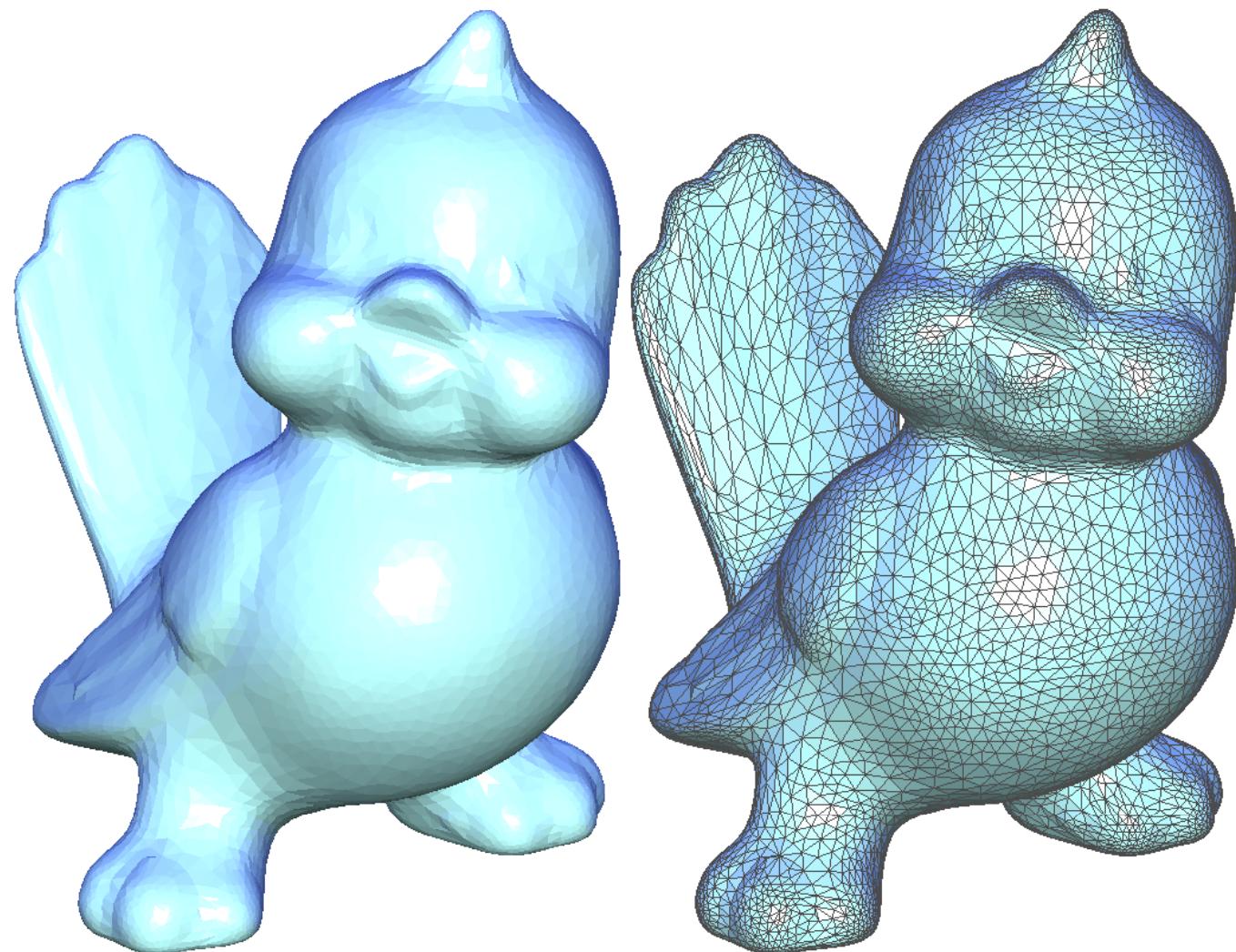


$$t_{\text{mean}} = 0.826$$
$$t_{\text{min}} = 0.034$$

# Original



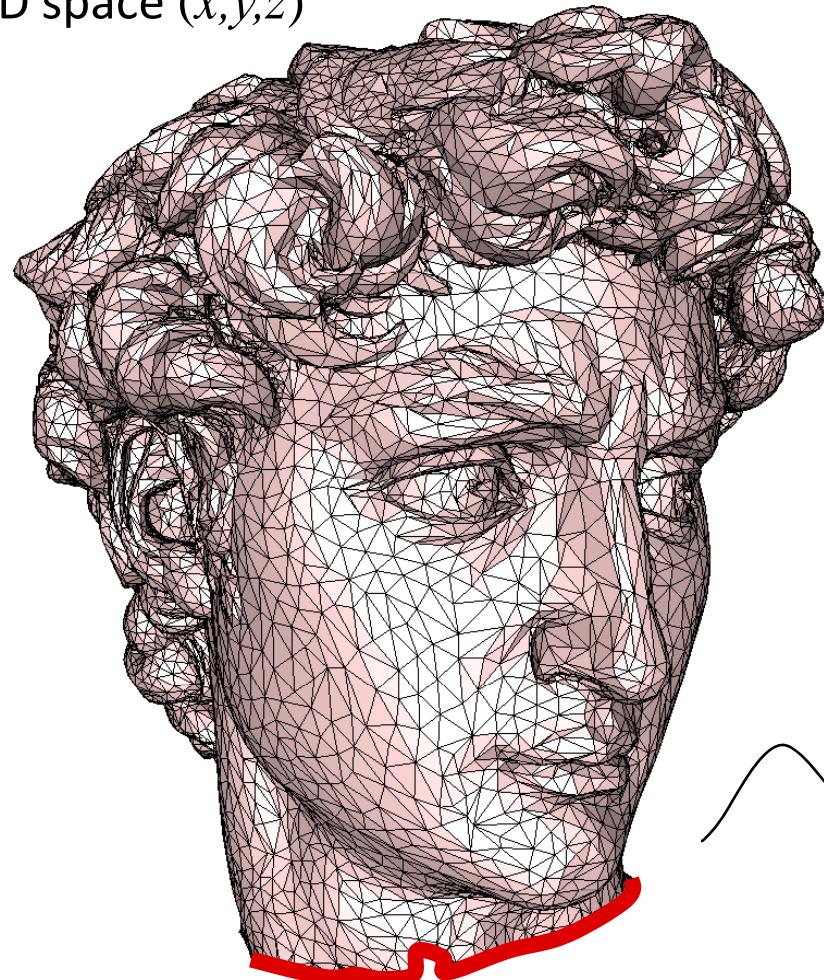
# Tri Shape Optimization



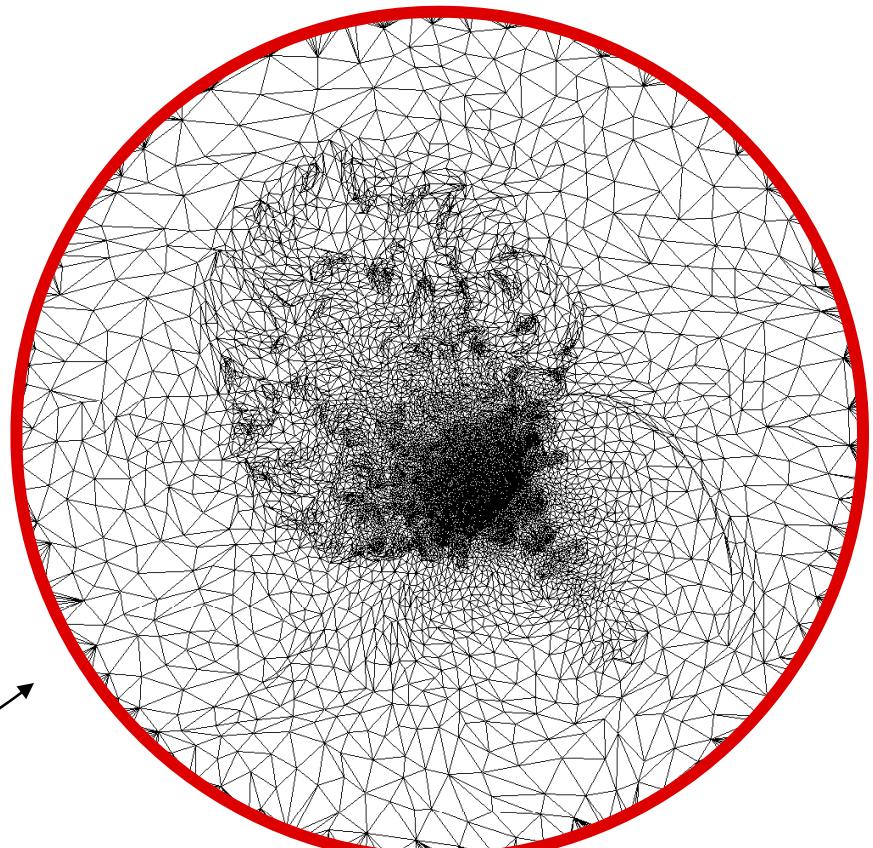
# Parameterization and Remeshing

# Surface parameterization

3D space ( $x,y,z$ )



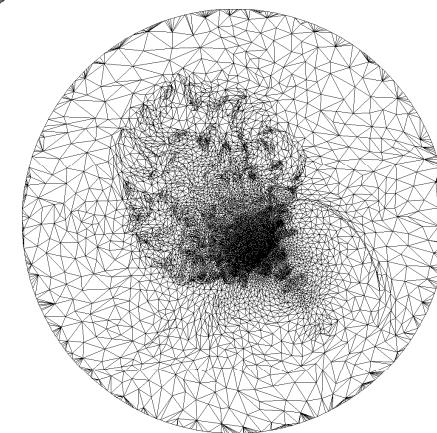
2D parameter domain ( $u,v$ )



boundary

boundary

# Texture mapping



3/12/2009

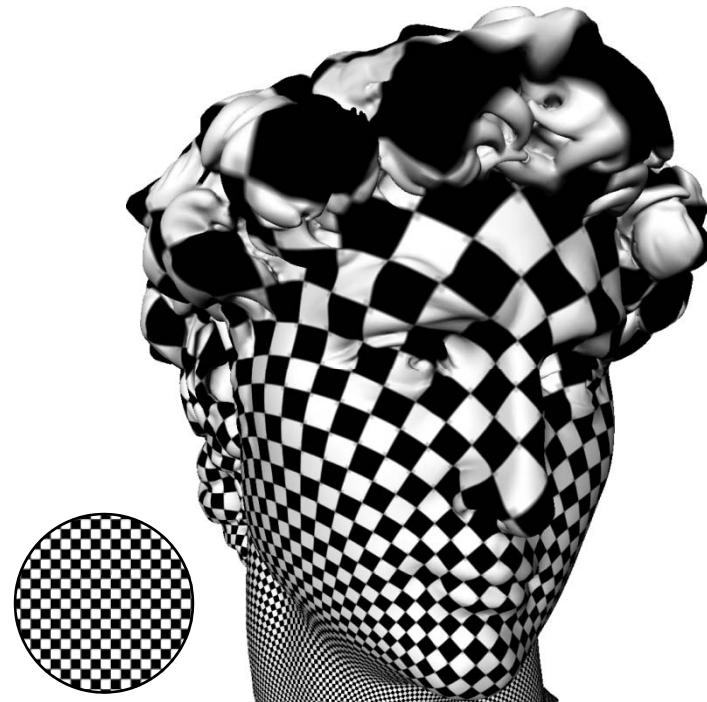
# Texture mapping



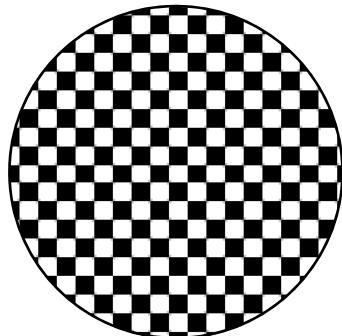
# Mesh parameterization

Requirements

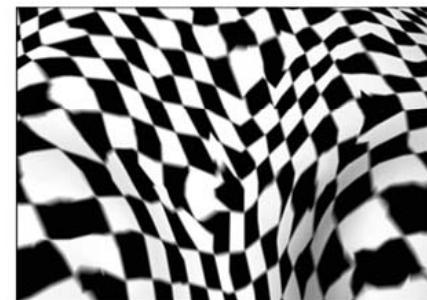
- Bijective (1-1 and onto): No triangles fold over.
- Minimal “distortion”
  - Preserve 3D angles
  - Preserve 3D distances
  - Preserve 3D areas
  - No “stretch”



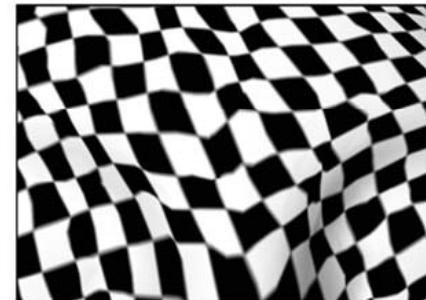
# Distortion minimization



Texture map



Kent et al '92

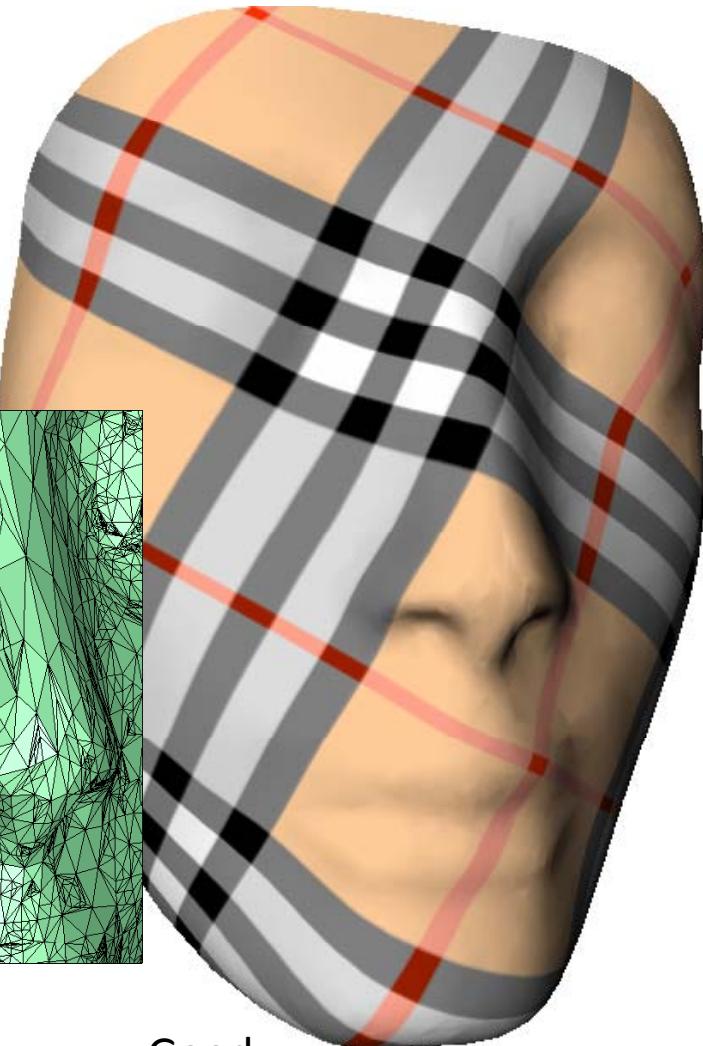
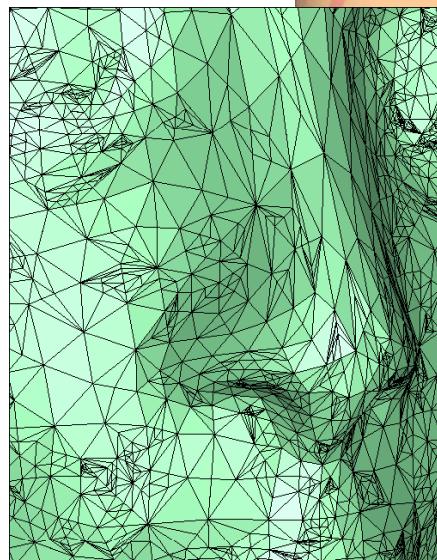
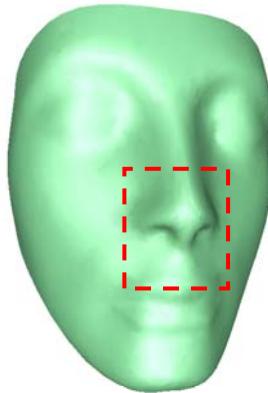


Floater 97



Sander et al '01

# Sensitivity to mesh quality

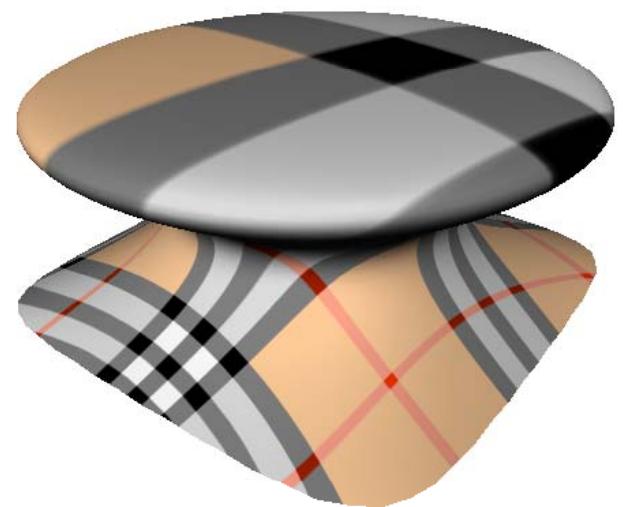
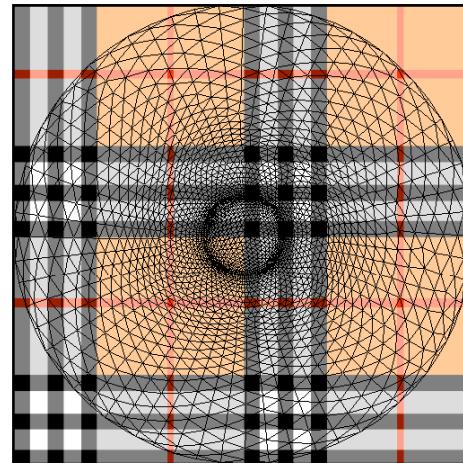
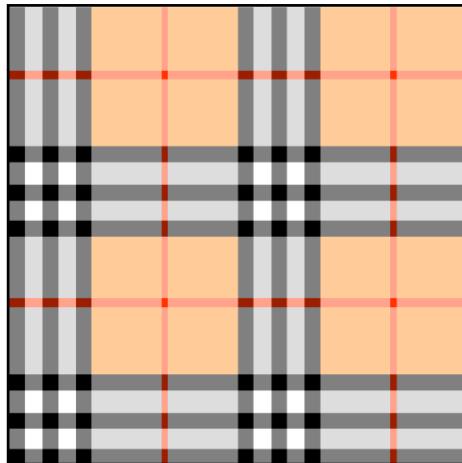


Good  
parameterization algorithm



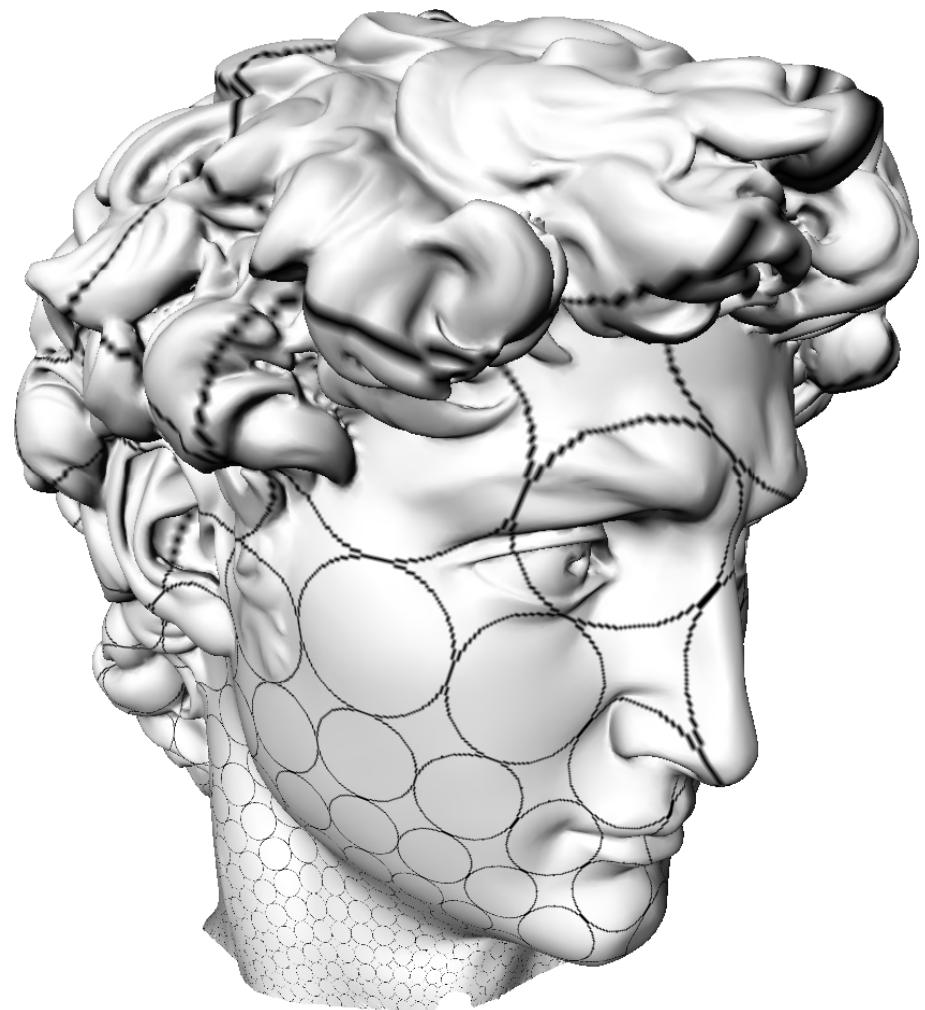
Not so good  
parameterization algorithm

# Area distortion vs. angle distortion

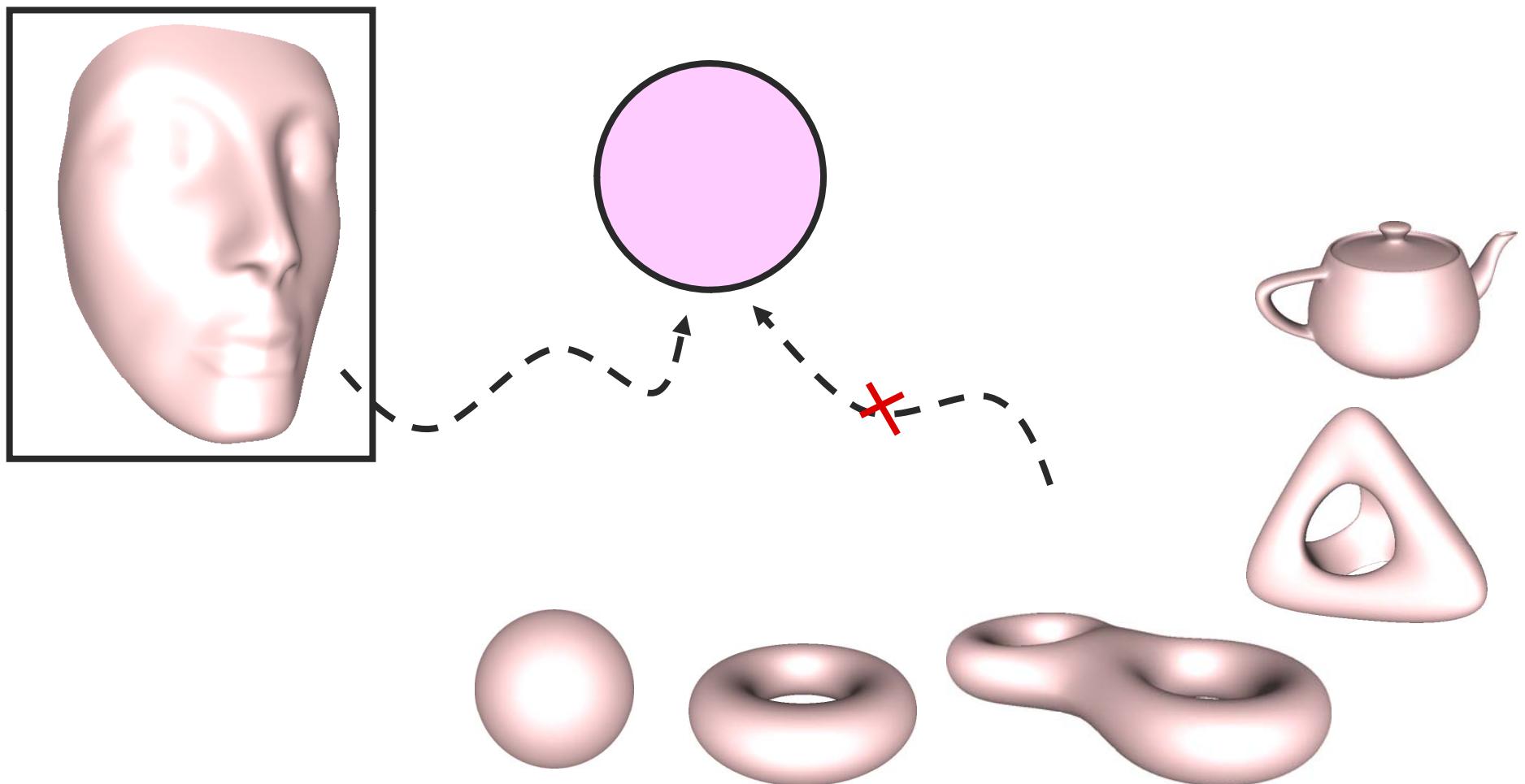


# Conformal parameterization

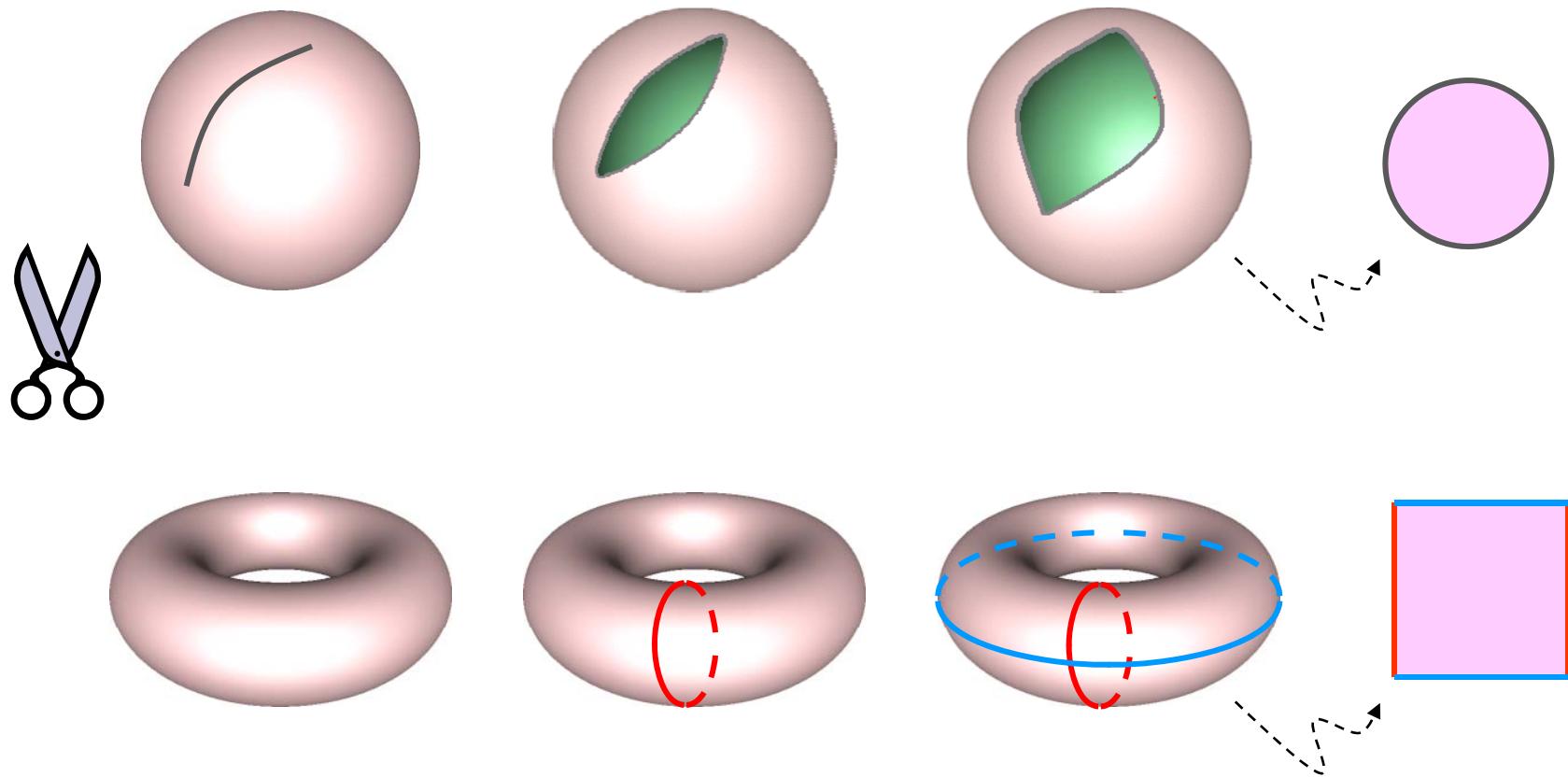
angle preservation; circles are mapped to circles



# Non-disk domains



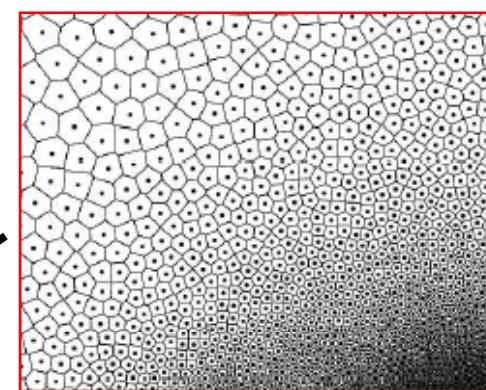
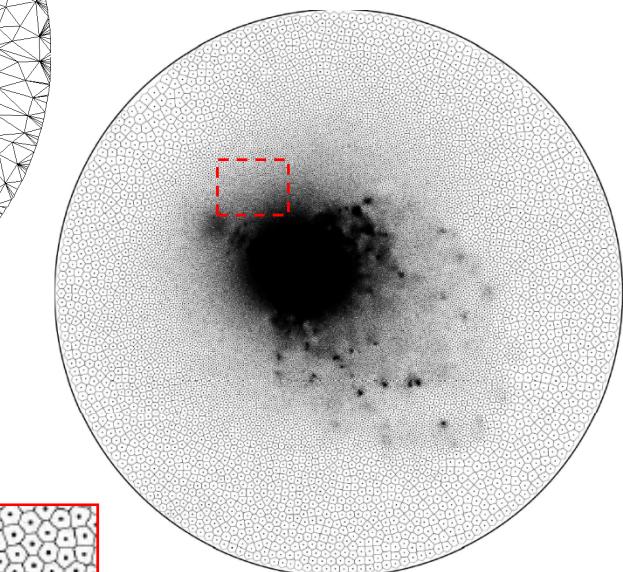
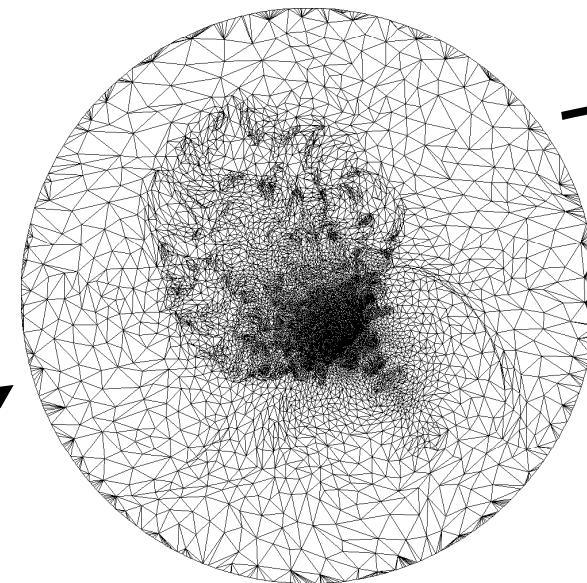
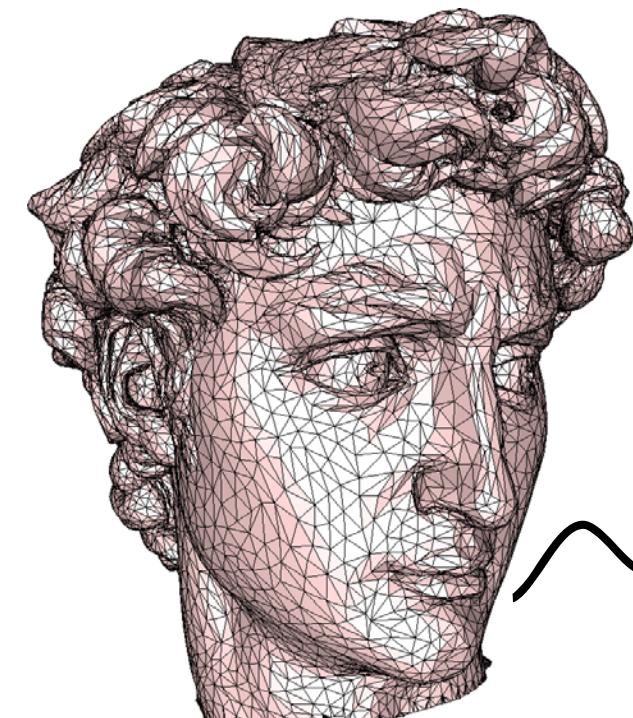
# Cutting



# Why parameterization?

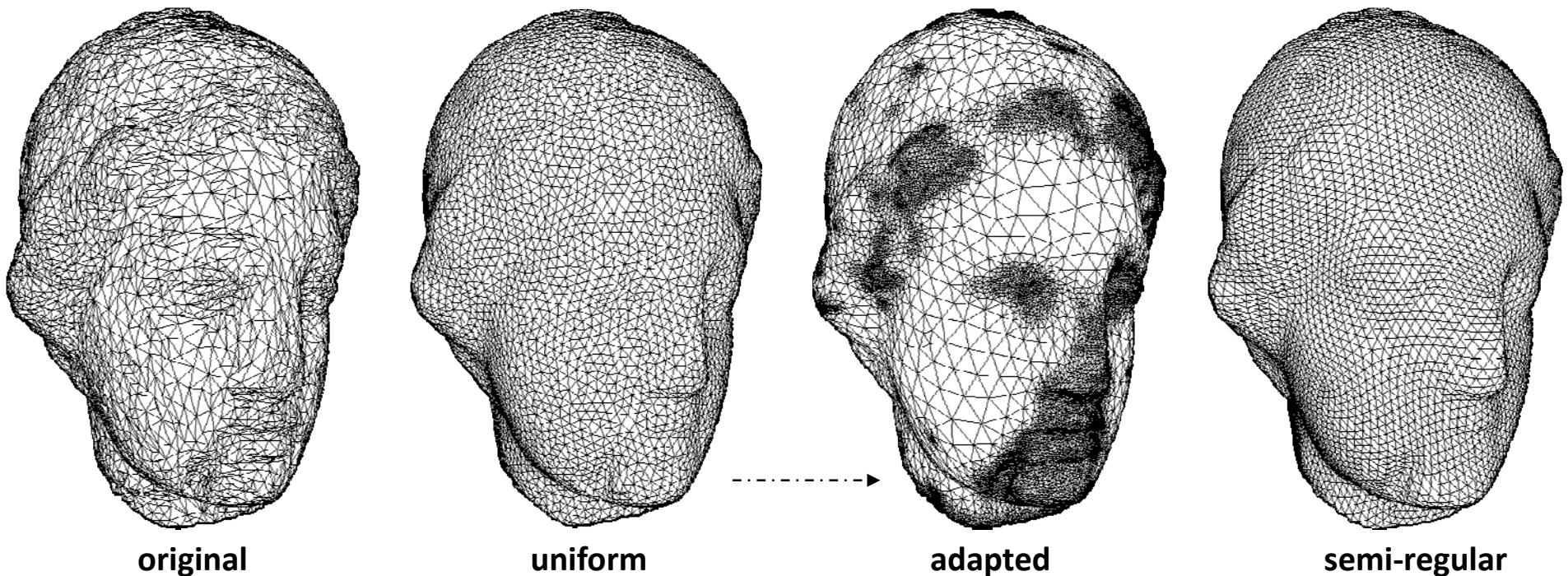
- Allows us to do many things in 2D and then map those actions onto the 3D surface
- It is often easier to operate in the 2D domain
- Mesh parameterization allows to use some notions from continuous surface theory

# Remeshing

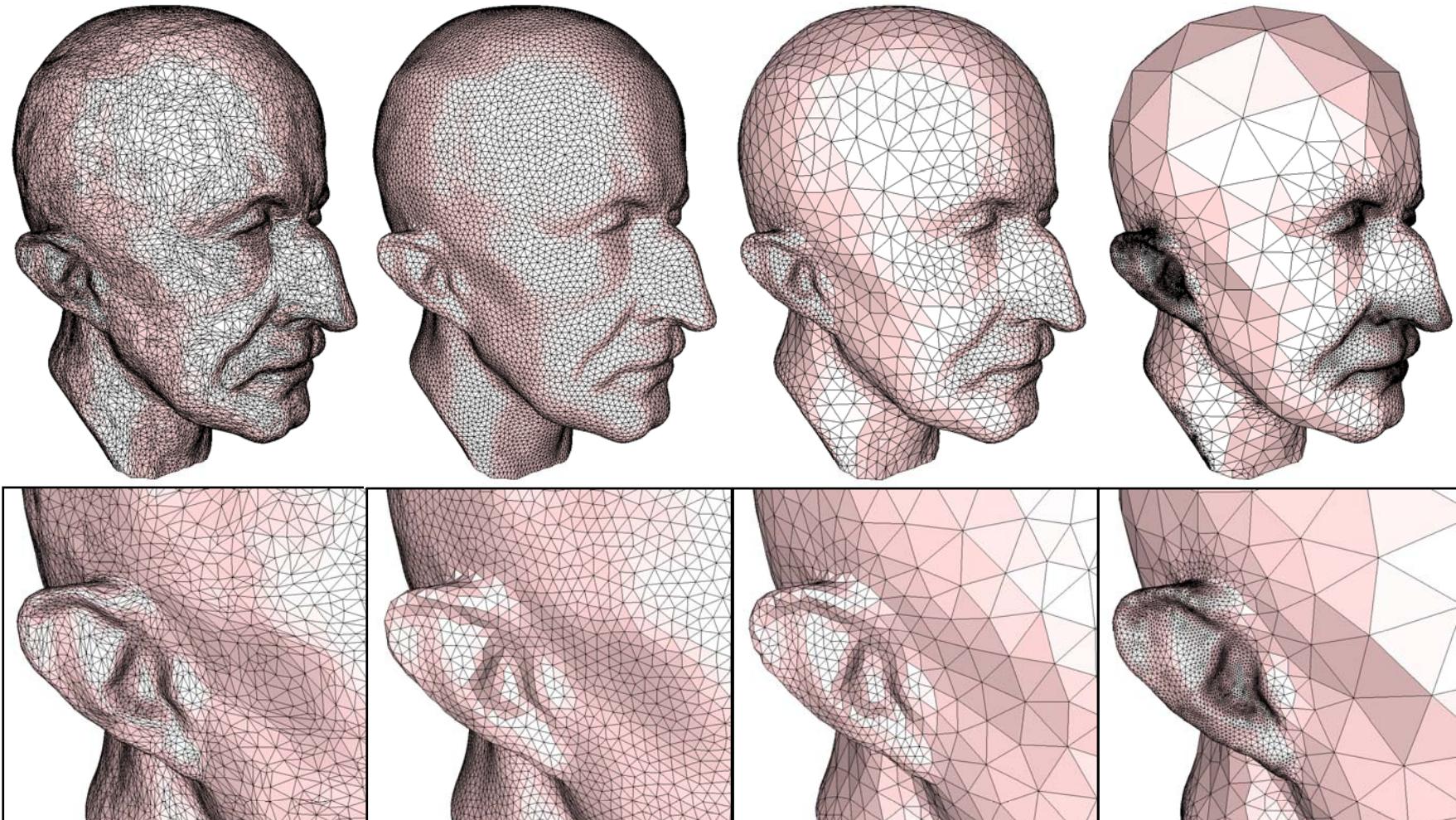


# Remeshing

- Particular remeshing type according to application

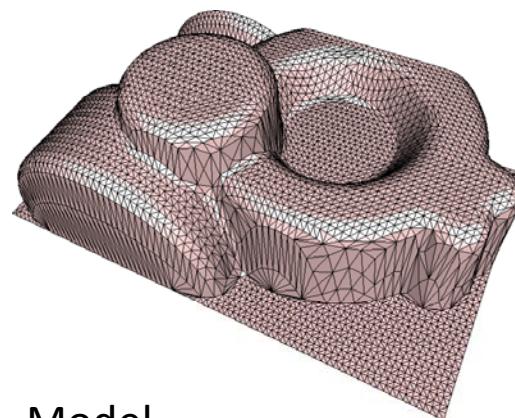


# Remeshing examples

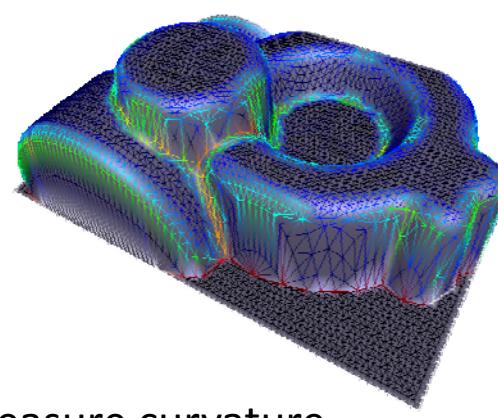


# Interactive geometry remeshing

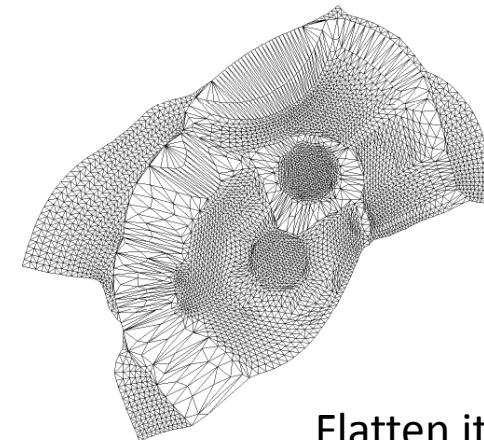
[Alliez et al., SIGGRAPH 2002]



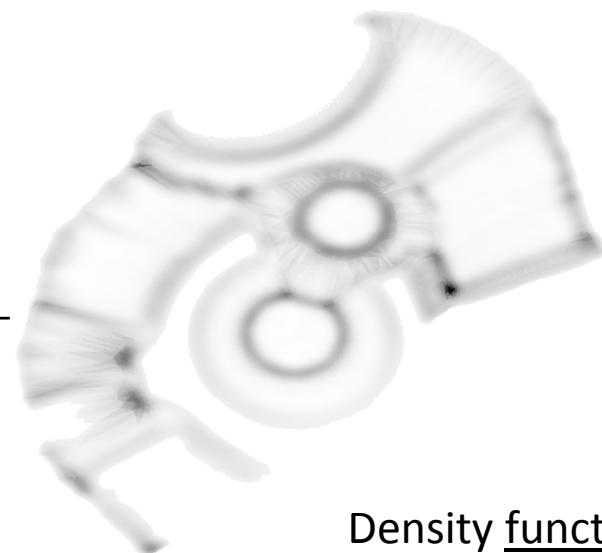
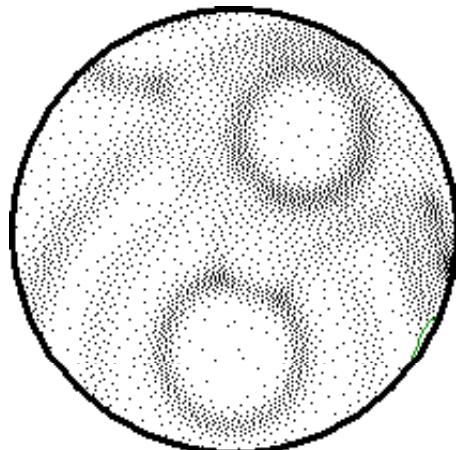
Model



Measure curvature



Flatten it  
**conformally**

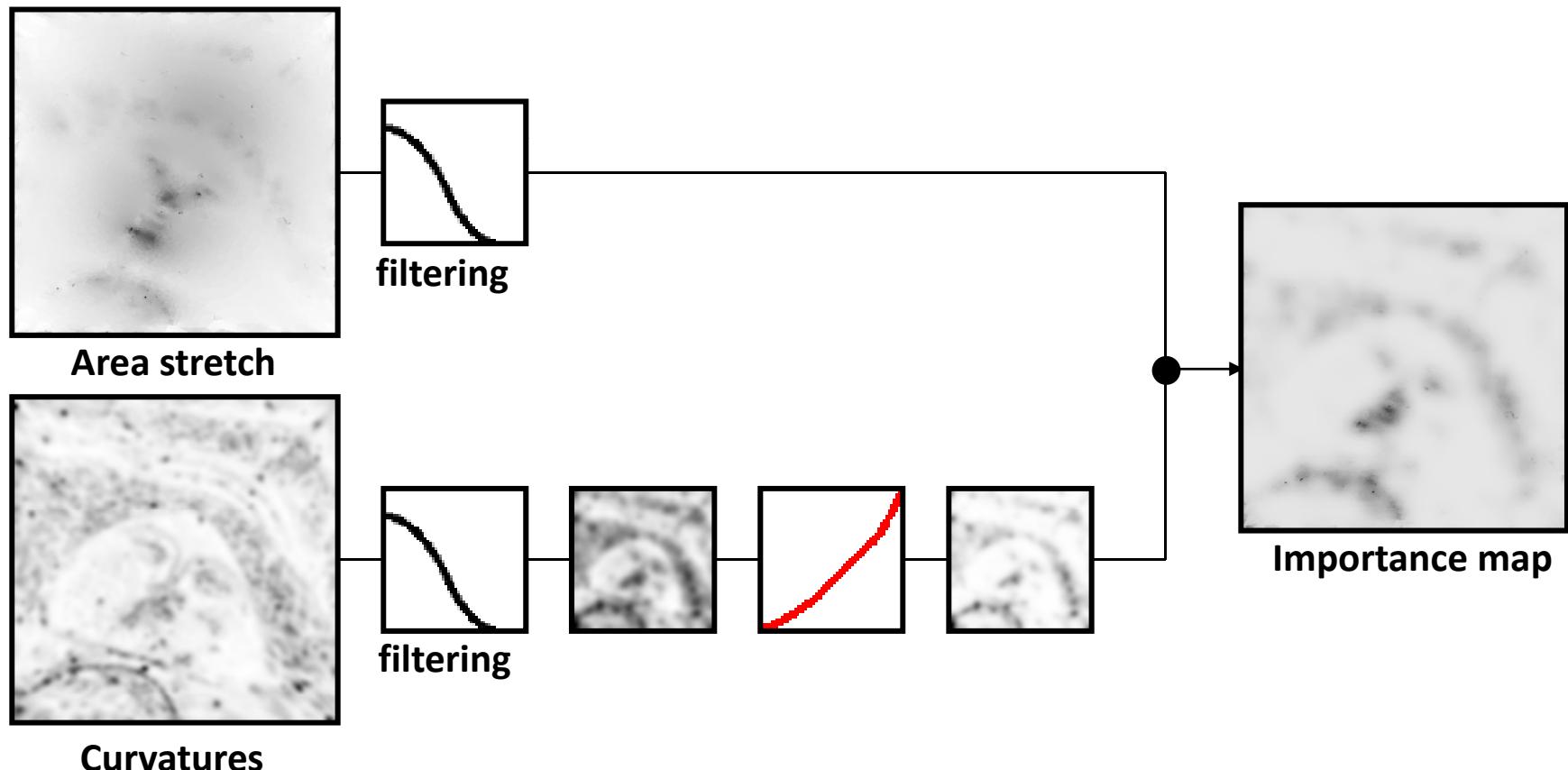


Density function in parameter space

# Interactive geometry remeshing

[Alliez et al., SIGGRAPH 2002]

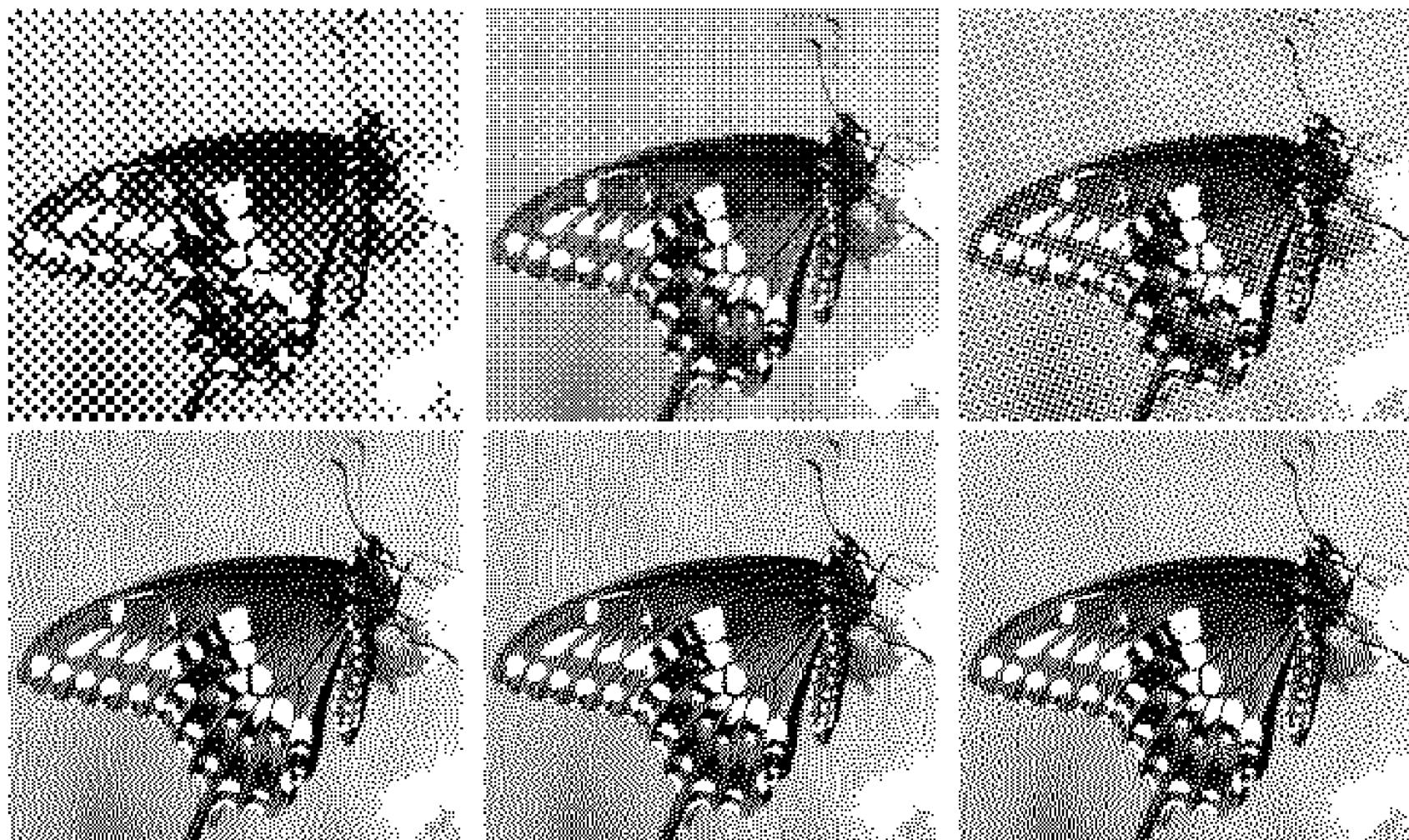
- Importance map created according to application needs



# Interactive geometry remeshing

[Alliez et al., SIGGRAPH 2002]

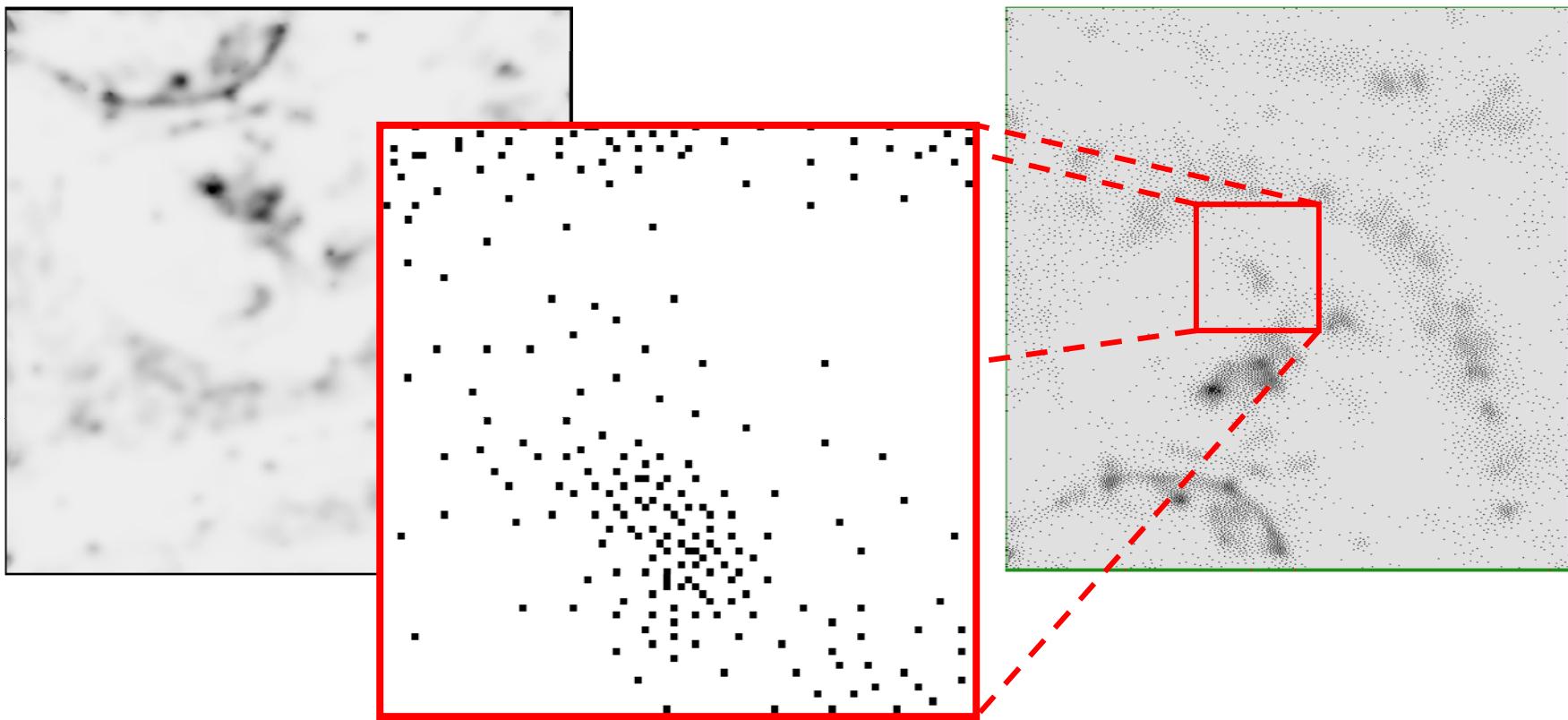
- Importance map is sampled by points – as in halftoning



# Interactive geometry remeshing

[Alliez et al., SIGGRAPH 2002]

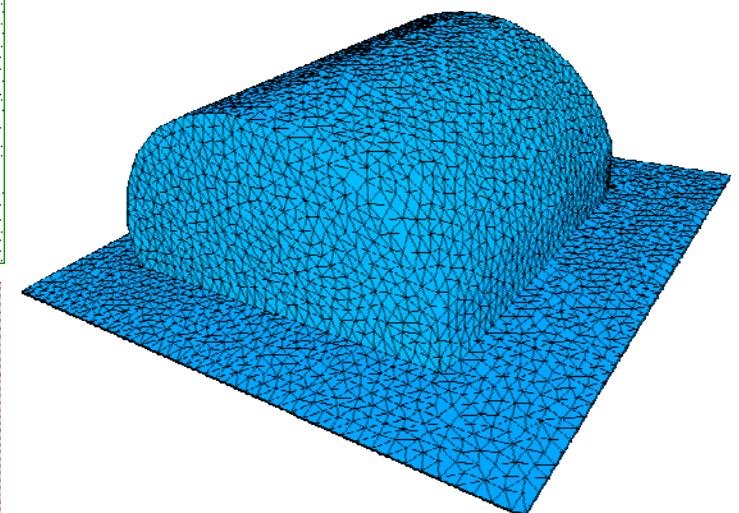
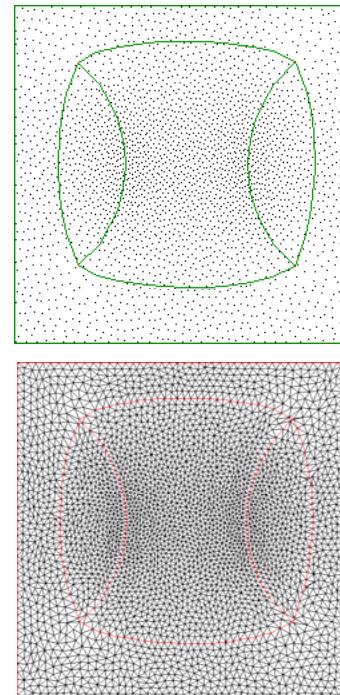
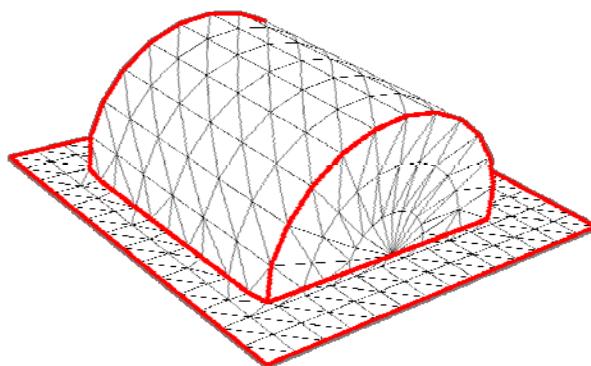
- Importance map is sampled by points – as in halftoning (error diffusion process)



# Interactive geometry remeshing

[Alliez et al., SIGGRAPH 2002]

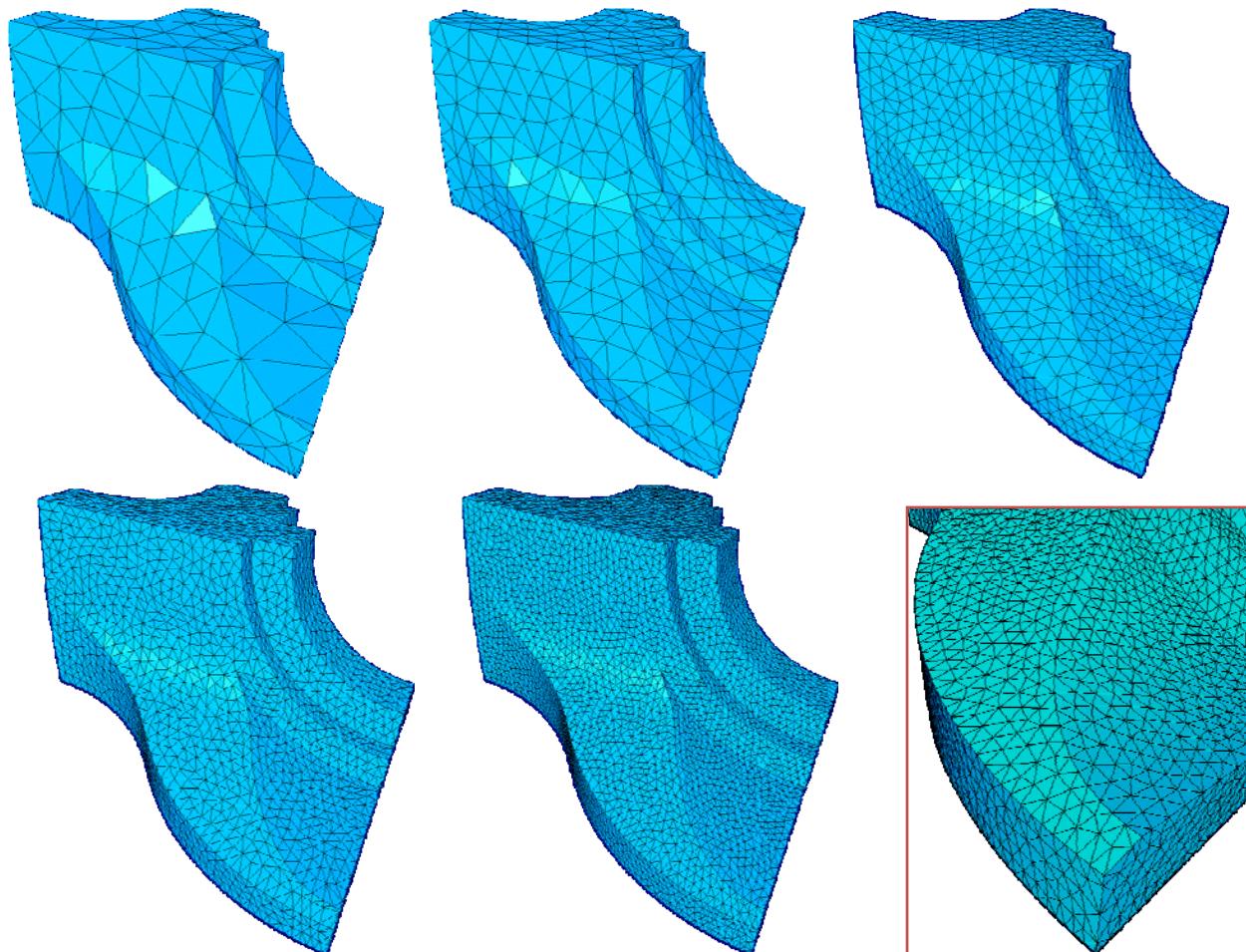
- Sampled points are triangulated using Delaunay
- Using the parameterization, the 2D points are lifted back into 3D



# Interactive geometry remeshing

[Alliez et al., SIGGRAPH 2002]

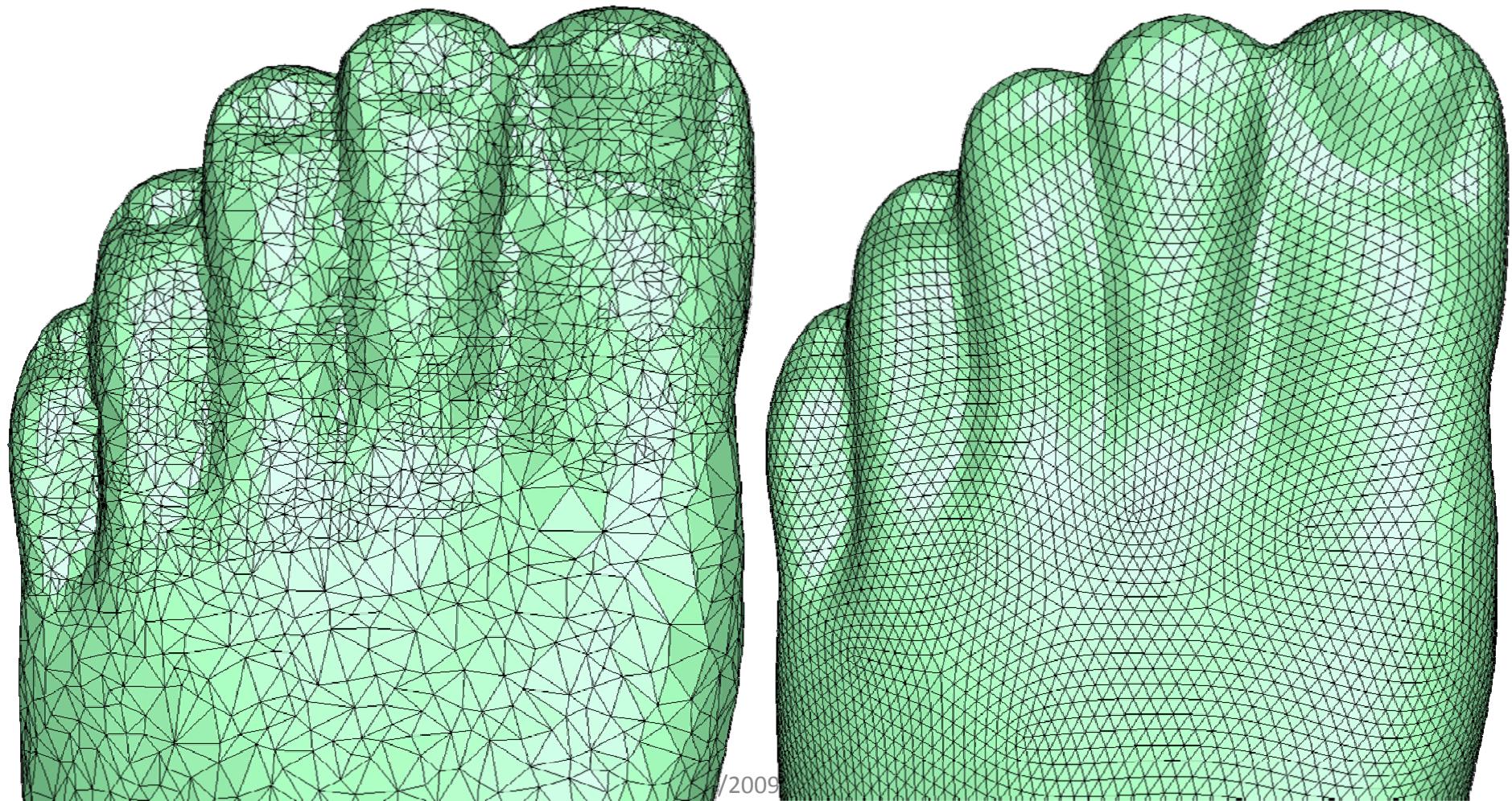
- More results



# Interactive geometry remeshing

[Alliez et al., SIGGRAPH 2002]

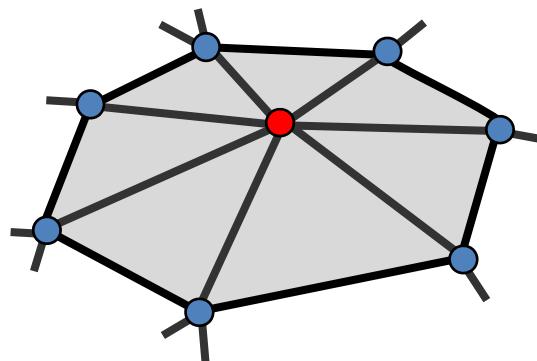
- More results



# Computing parameterizations

# General idea

- Want to flatten the mesh → no curvature → Laplace operator gives zero.



$\mathbf{v} = (u, v)$  domain

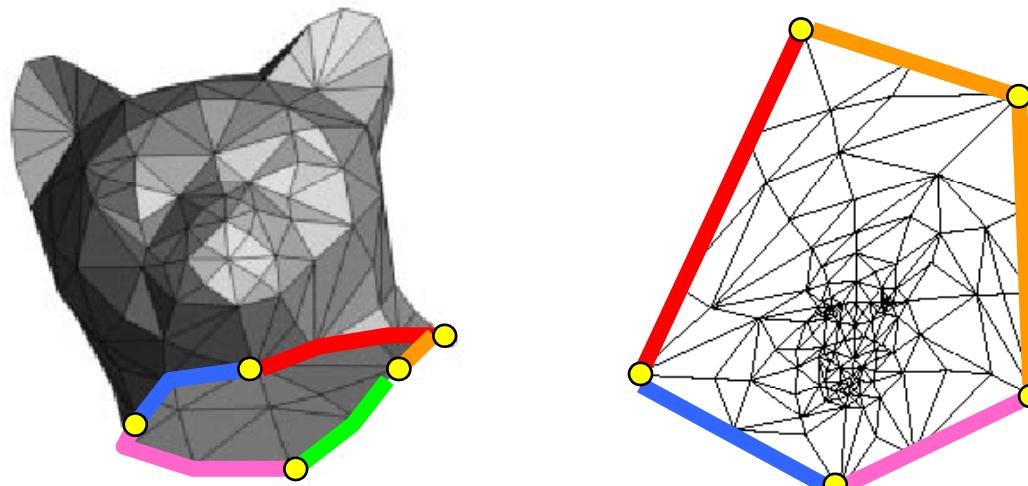
$$L\mathbf{v} = 0$$

need boundary constraints  
to prevent trivial solution;

which Laplacian operator  
(which weights?)

# Convex mapping (Tutte, Floater)

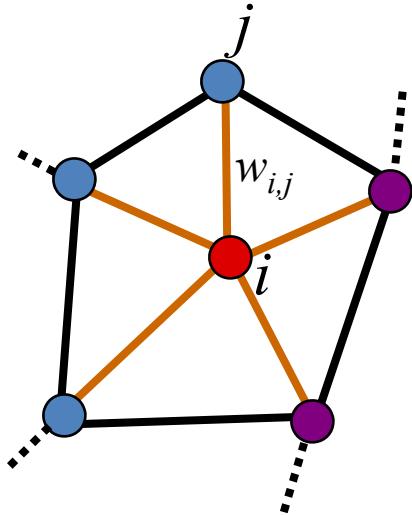
- Boundary vertices are fixed
- Convex weights



$\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  – inner vertices;     $\mathbf{v}_n, \mathbf{v}_{n+1}, \dots, \mathbf{v}_N$  – boundary vertices

# Convex mapping (Tutte, Floater)

- Boundary vertices are fixed
- Convex weights



$$L(\mathbf{v}_i) = \sum_{\mathbf{v}_j \in N_1(\mathbf{v}_i)} w_{ij} (\mathbf{v}_j - \mathbf{v}_i) = 0$$
$$w_{ij} > 0$$
$$\sum_{\mathbf{v}_j \in N_1(\mathbf{v}_i)} w_{ij} = 1$$

$\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  – inner vertices;  $\mathbf{v}_n, \mathbf{v}_{n+1}, \dots, \mathbf{v}_N$  – boundary vertices

# Convex mapping (Tutte, Floater)

- Solve the linear system

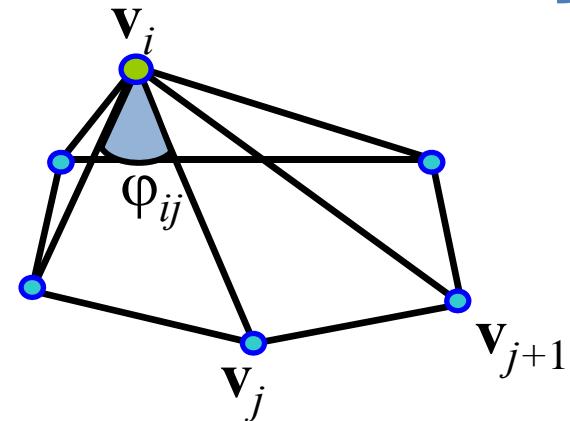
$$L\mathbf{v} = 0$$

- Remember to substitute the fixed boundary vertices
- Property: if boundary is **convex** and weights are **convex**, the flattening is **guaranteed** to be **legal** (no fold-overs)

# Convex weights

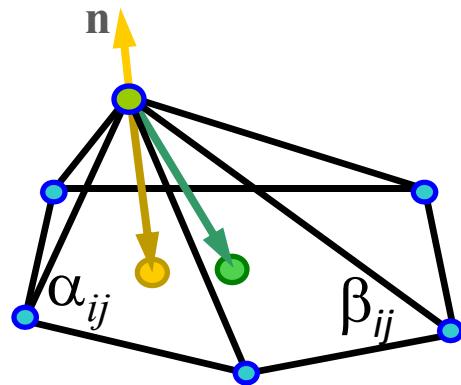
- Tutte (1966): uniform weights  $w_{ij} = \frac{1}{d_i}$ 
    - Only depend on connectivity
    - High distortion
  - Floater (1997): shape-preserving weights
  - Floater (2003): mean-value weights
- weights taken from 3D mesh

$$w_{ij} = \tan \frac{\varphi_{i,j}}{2} + \tan \frac{\varphi_{i,j+1}}{2}$$



# Conformal parameterization

- Preserve angles as much as possible
- “Project” each vertex along its normal direction

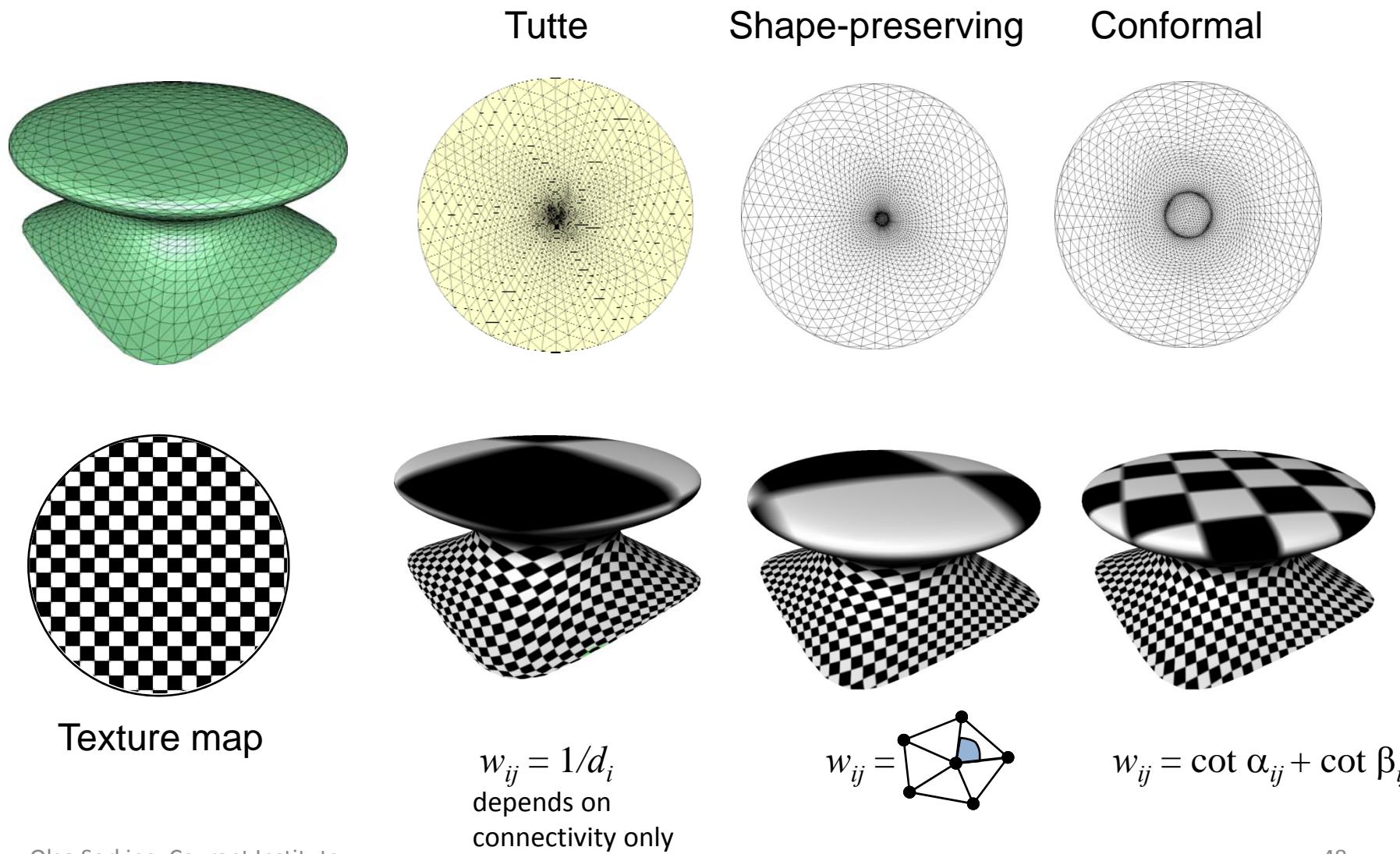


$$L_{cot} \mathbf{v} = 0$$

$$w_{ij} = \cot \alpha_{ij} + \cot \beta_{ij}$$

$\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  – inner vertices;     $\mathbf{v}_n, \mathbf{v}_{n+1}, \dots, \mathbf{v}_N$  – fixed boundary vertices

# Comparison



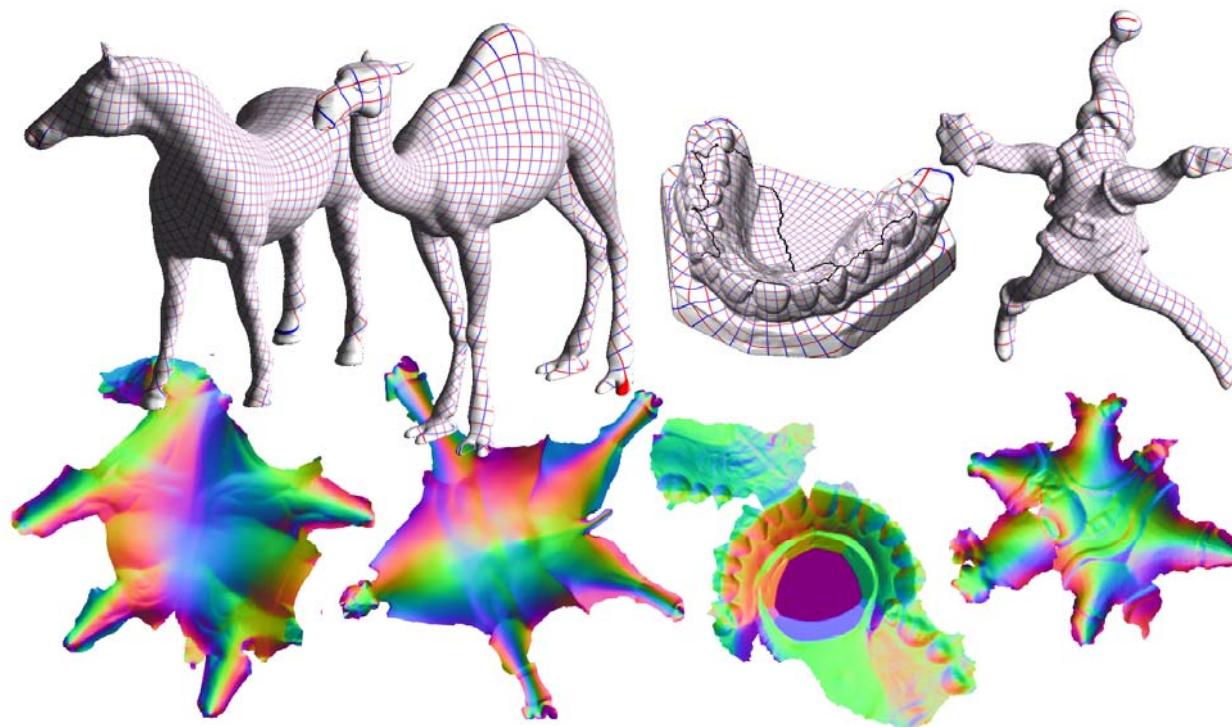
# Discussion

- The results of **harmonic mapping** are **better** than those of **convex** mapping (local area and angles preservation).
- But: the mapping is **not always legal** (the weights can be negative for badly-shaped triangles...)



# Discussion

- Both mappings have the problem of **fixed boundary** – it constrains the minimization and causes **distortion**.
- More advanced methods do not require boundary conditions (see references on the website).



ABF++ method,  
Sheffer et al.