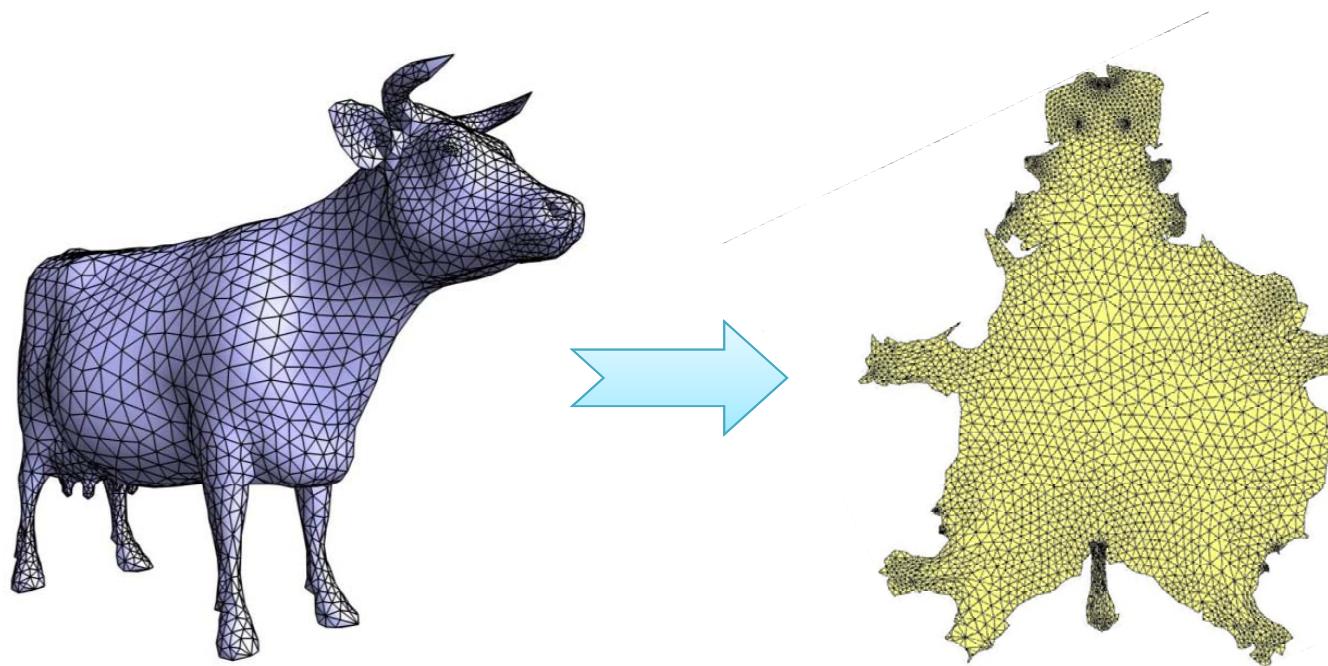
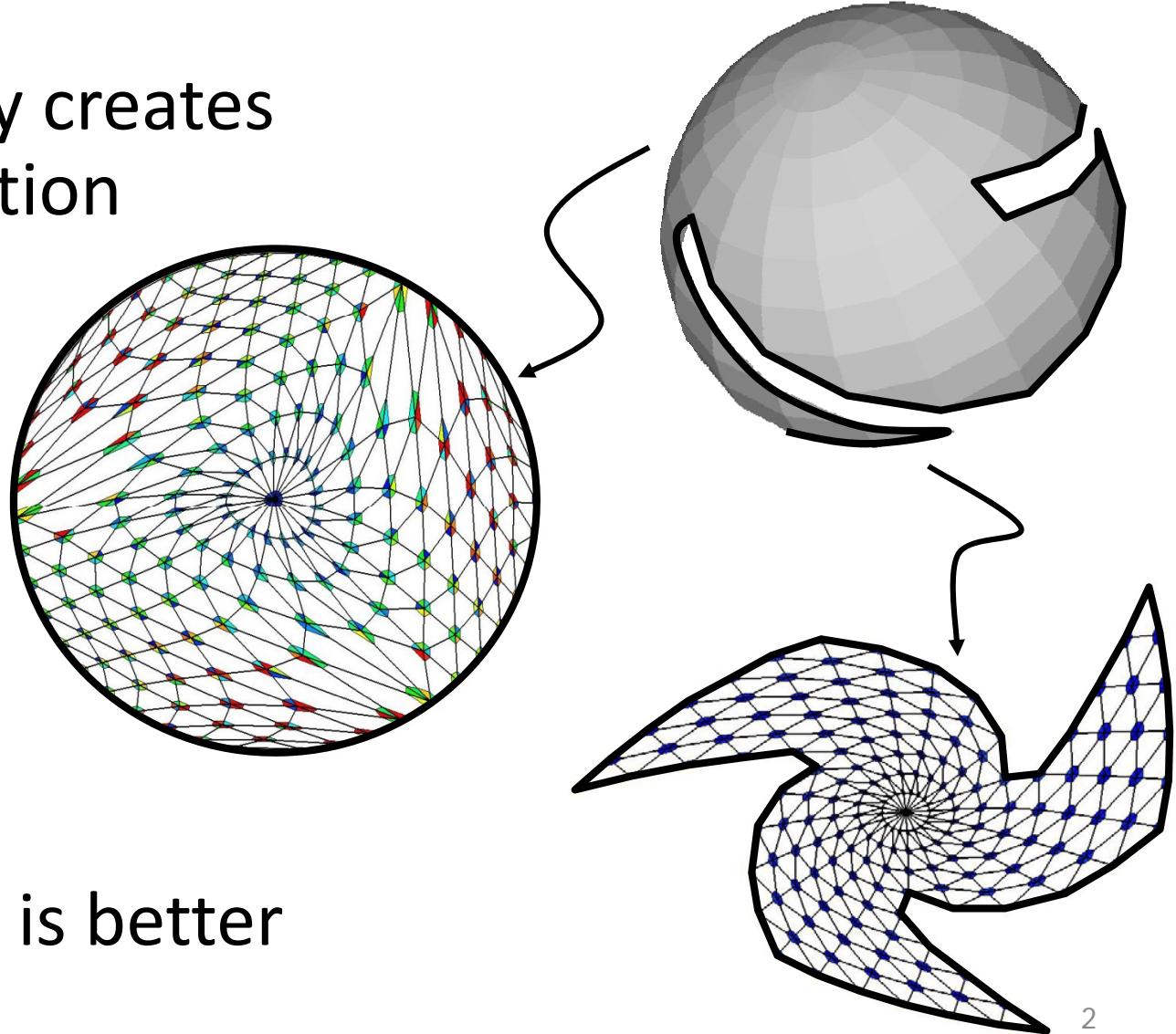


Parameterization II



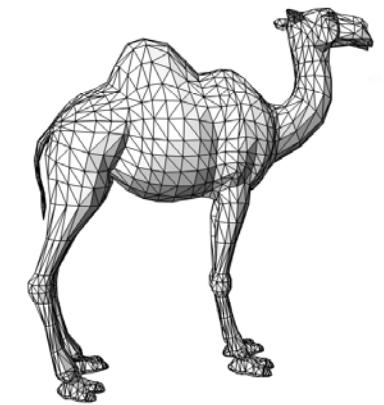
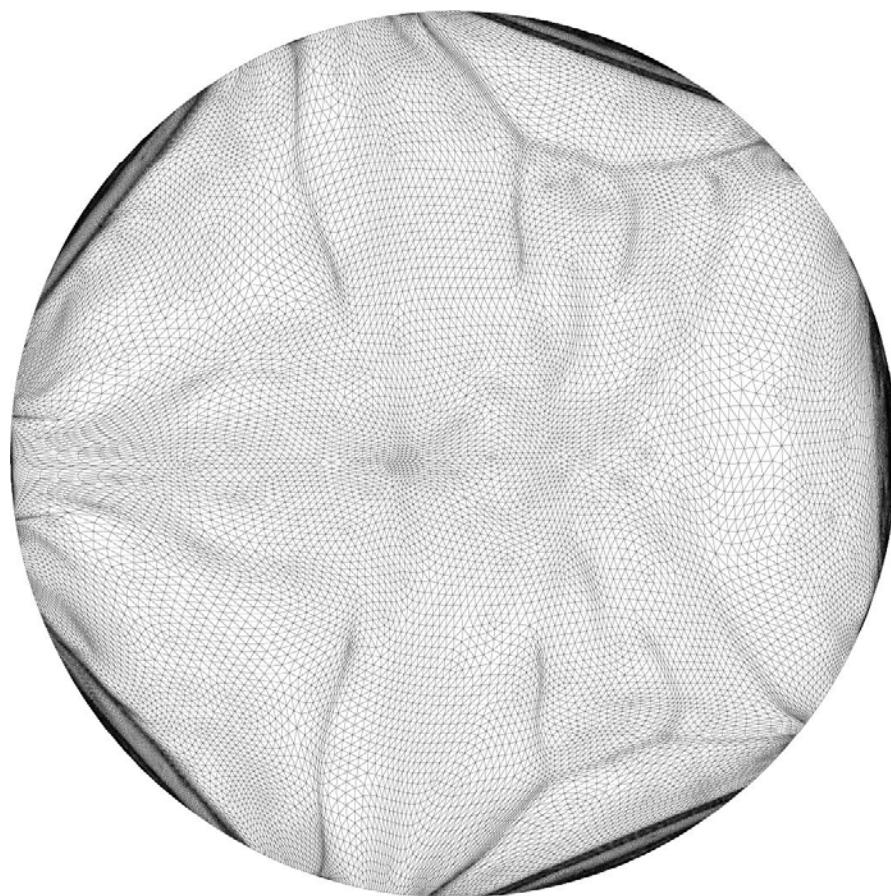
Non-Convex Boundary

- Convex boundary creates significant distortion

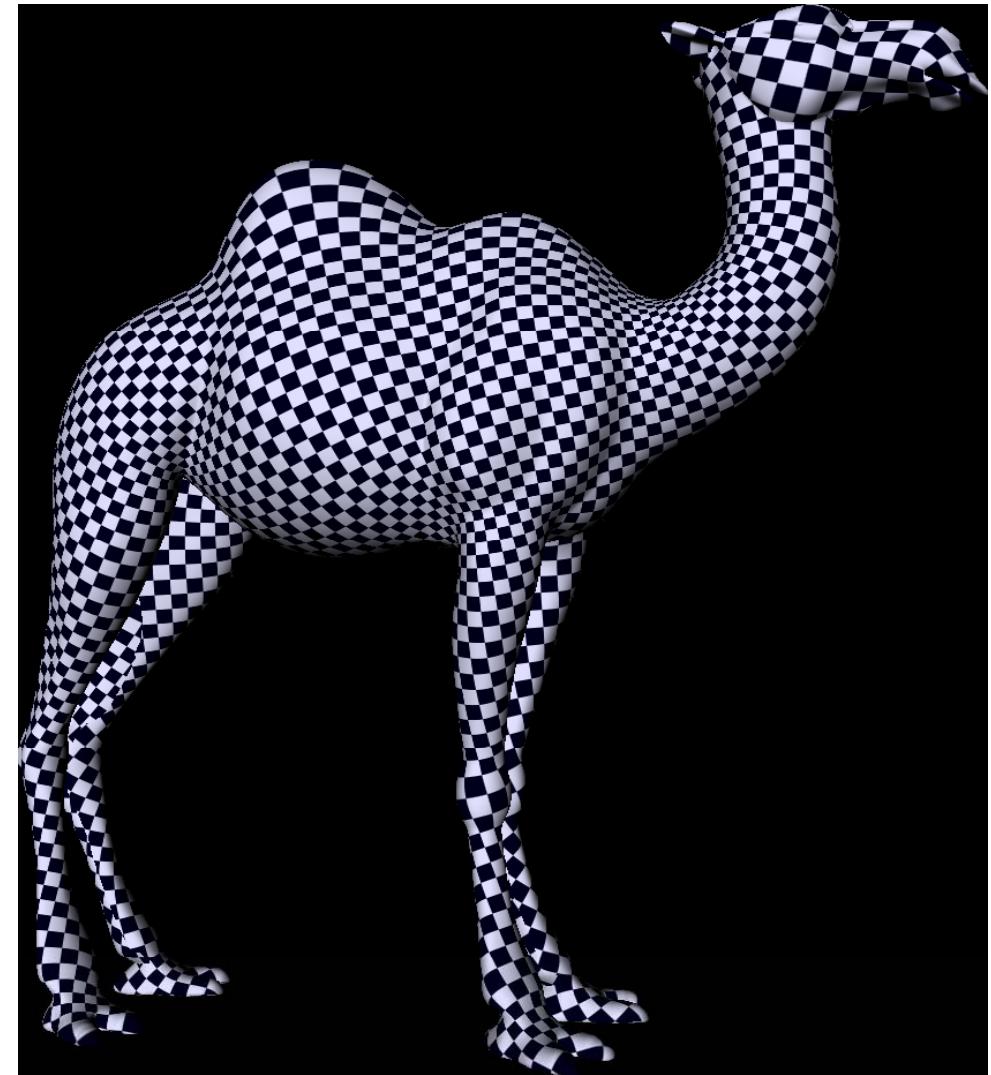
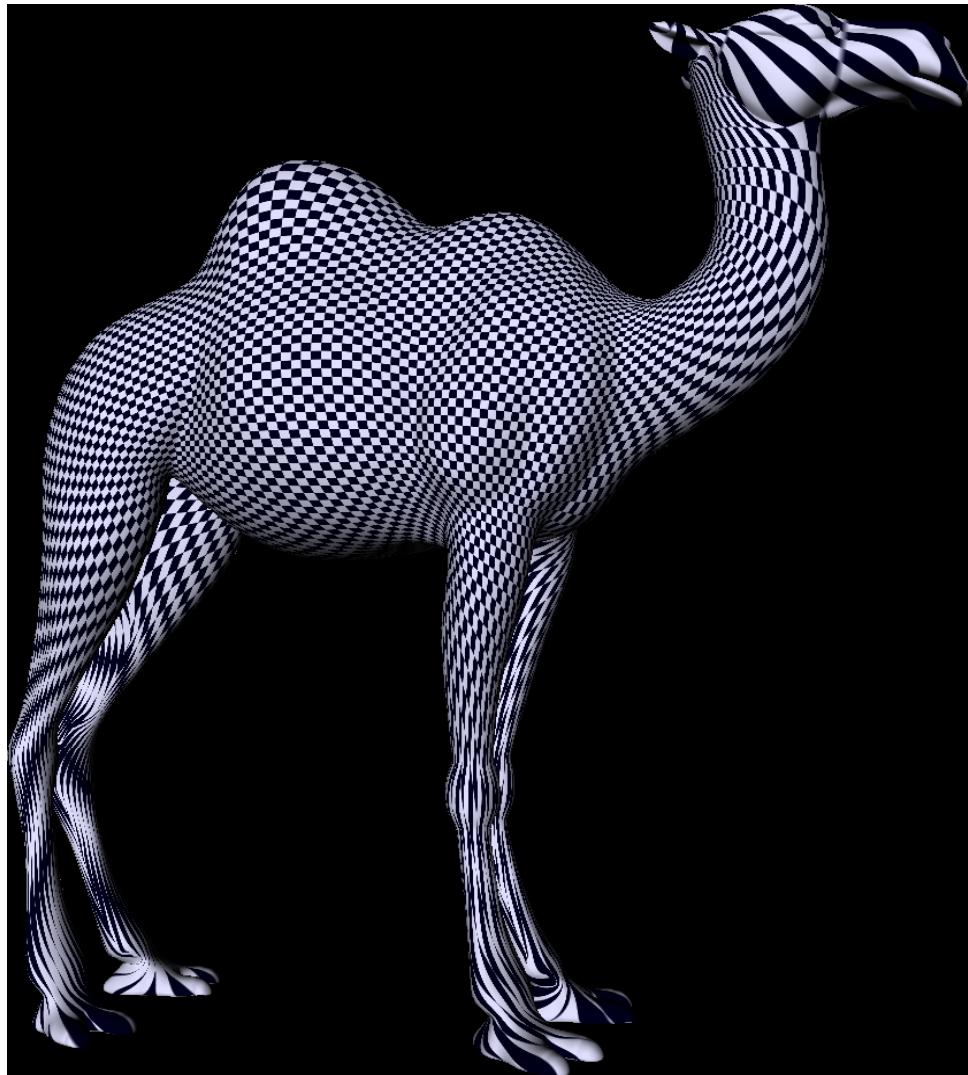


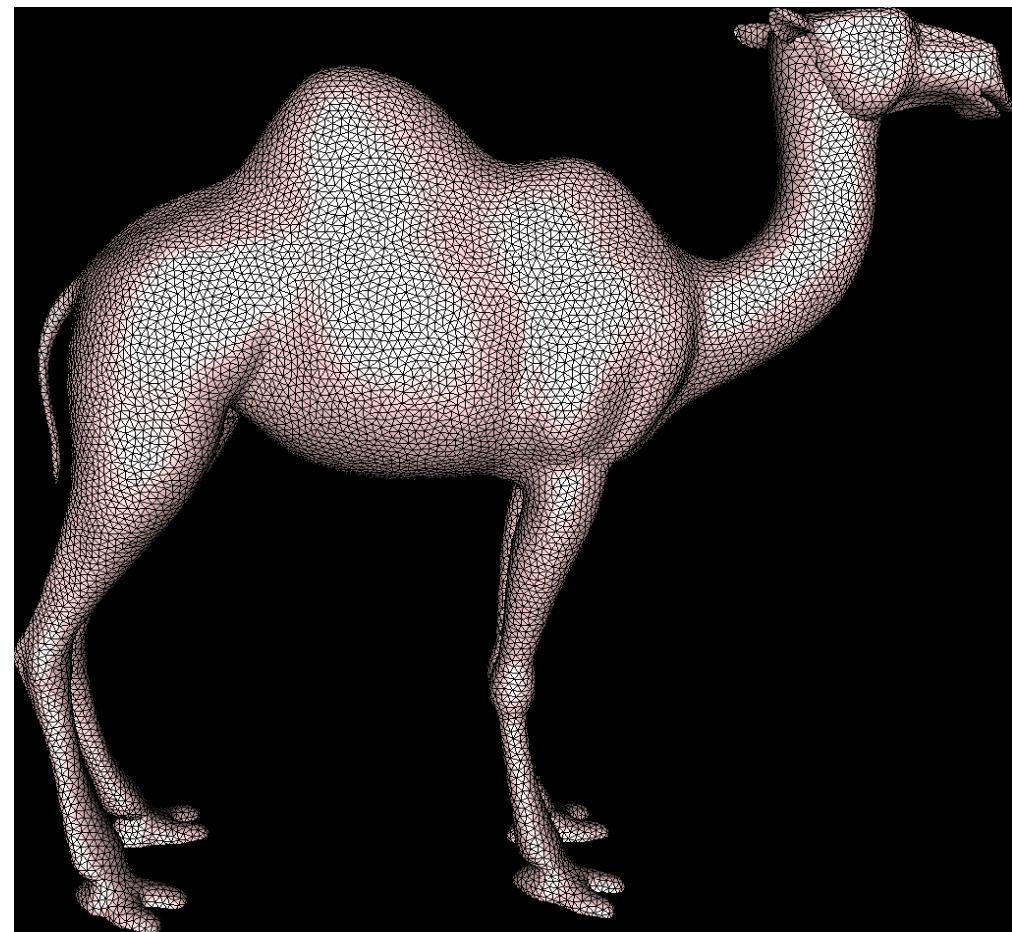
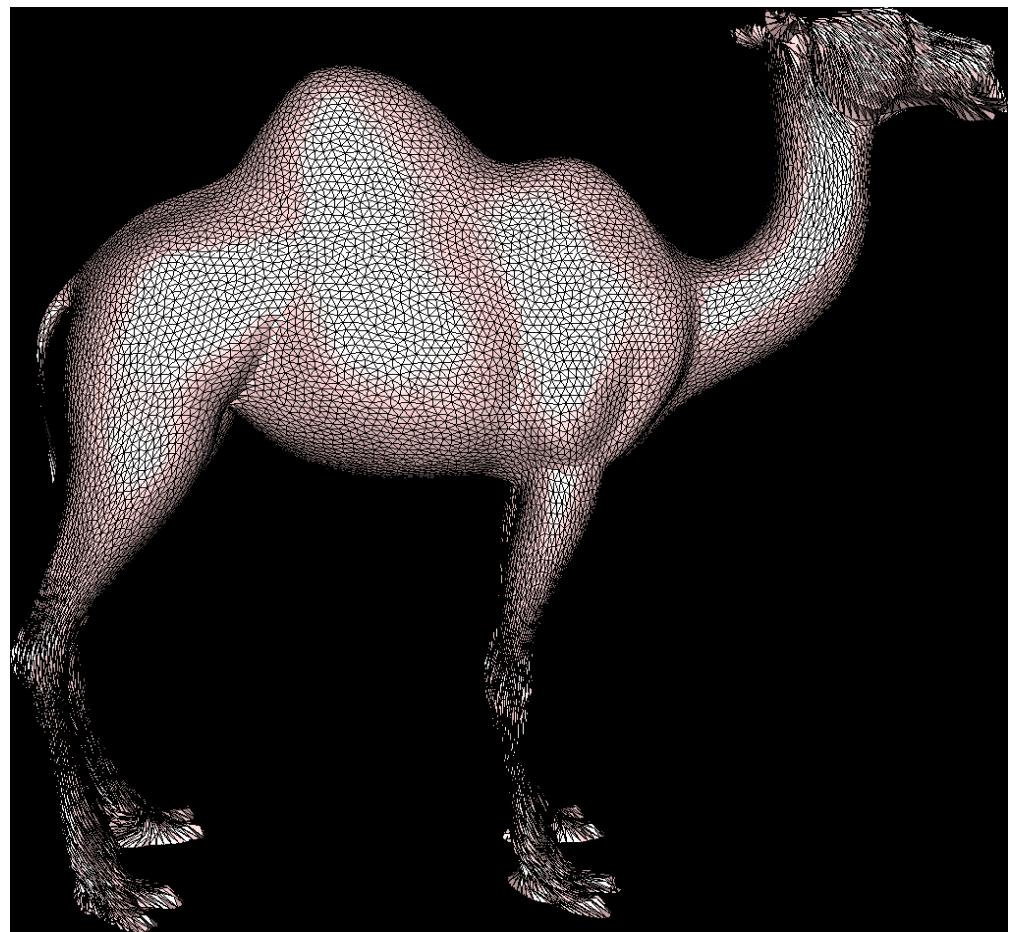
- “Free” boundary is better

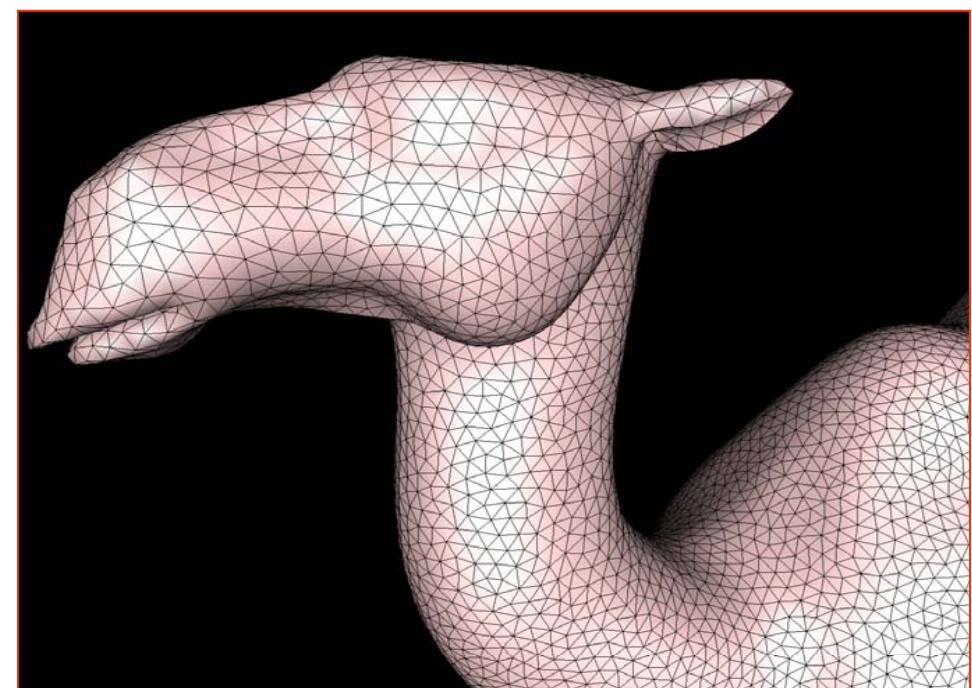
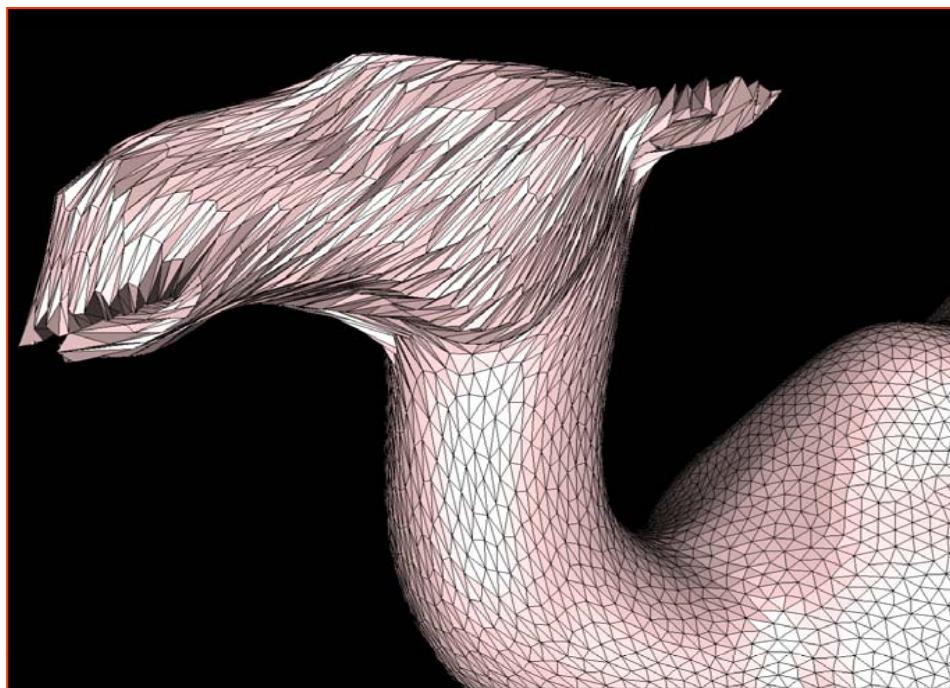
Fixed vs Free Boundary

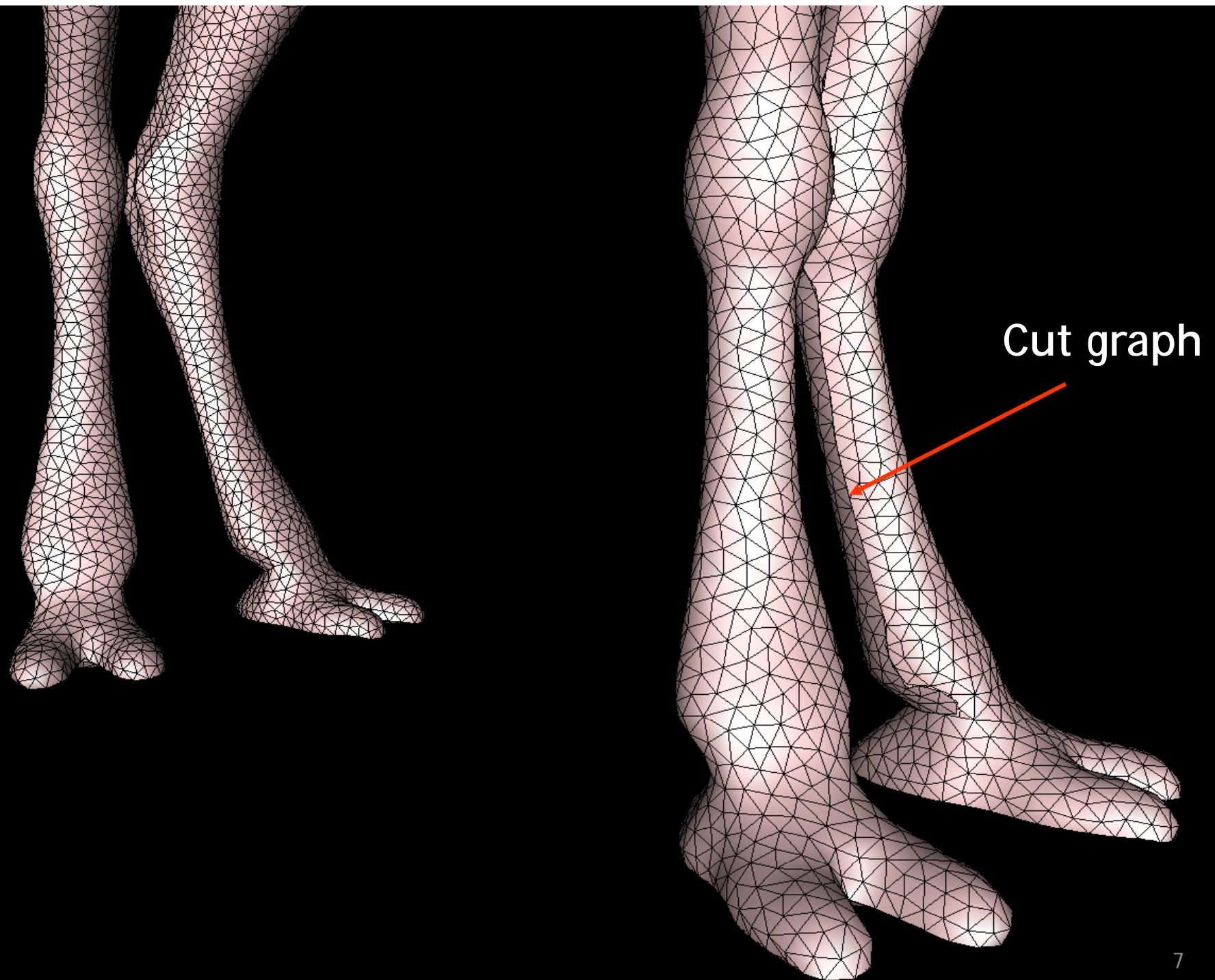


Fixed vs Free Boundary











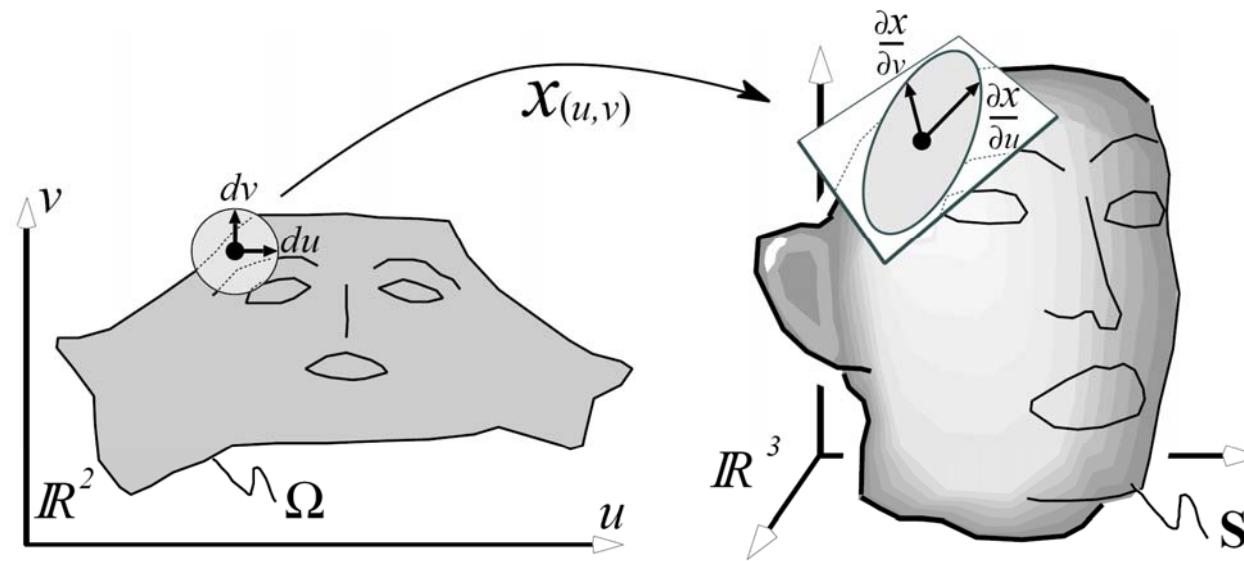
Free Boundary Methods - Tactics

- Solve for the (u,v) coordinates using the discrete first fundamental form
 - MIPS [Hormann et al., 2000]
 - Stretch optimization [Sander et al., 2001]
 - LSCM (conformal, linear) [Levy et al., 2002]
 - DCP (conformal, linear) [Desbrun et al., 2002]
- Solve for the angles of the map (conformal)
 - ABF [Sheffer et al., 2001], ABF++ [Sheffer et al., 2004]
 - LinABF (linear) [Zayer et al., 2007]

Free Boundary Methods - Tactics

- Solve for the edge lengths of the map by prescribing curvature
 - Circle patterns [Kharevych et al., 2006]
 - CPMS (linear) [Ben-Chen et al., 2008]
 - CETM [Springborn et al., 2008]
- Balance area/conformality
 - ARAP [Liu et al., 2008]
- More...

Back to the First Fundamental Form



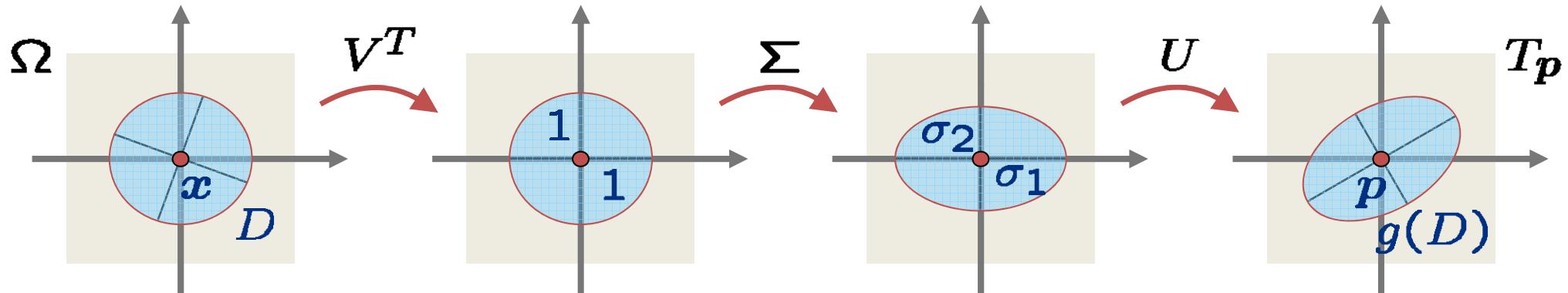
$$\mathbf{x}(u, v) = \begin{pmatrix} x(u, v) \\ y(u, v) \\ z(u, v) \end{pmatrix} \quad J = \begin{pmatrix} x_u & x_v \\ y_u & y_v \\ z_u & z_v \end{pmatrix}$$

Linear Map Surgery

- **Singular Value Decomposition (SVD) of J_f**

$$J_f = U \Sigma V^T = U \begin{pmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \\ 0 & 0 \end{pmatrix} V^T$$

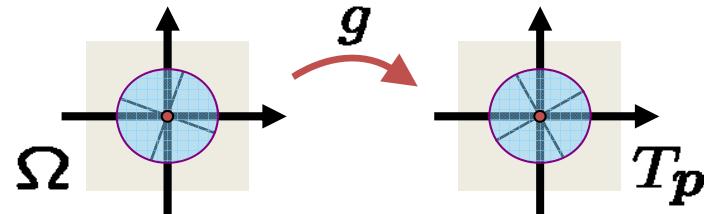
with **rotations** $U \in \mathbb{R}^{3 \times 3}$ and $V \in \mathbb{R}^{2 \times 2}$
and **scale factors** (singular values) $\sigma_1 \geq \sigma_2 > 0$



Notion of Distortion

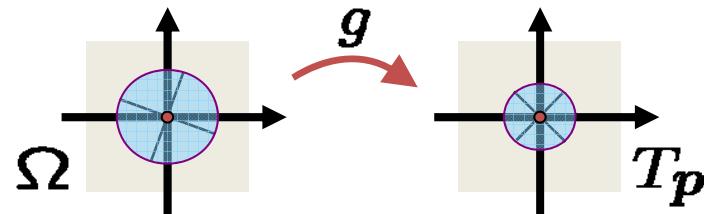
- **isometric** or **length**-preserving

$$\sigma_1 = \sigma_2 = 1$$



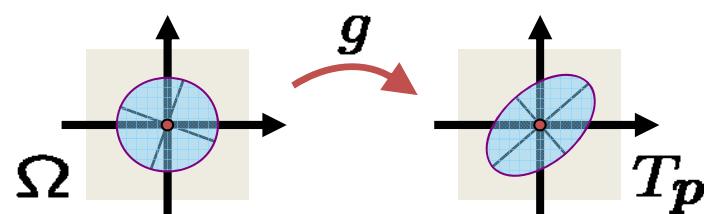
- **conformal** or **angle**-preserving

$$\sigma_1 = \sigma_2$$



- **equiareal** or **area**-preserving

$$\sigma_1 \cdot \sigma_2 = 1$$



- everything defined **pointwise** on Ω

Computing the Stretch Factors

- **first fundamental form** $\mathbf{I}_f = J_f^T J_f = \begin{pmatrix} E & F \\ F & G \end{pmatrix}$

$$E = f_u^T f_u \quad F = f_u^T f_v \quad G = f_v^T f_v$$

- **eigenvalues** of \mathbf{I}_f

$$\lambda_{1,2} = \frac{1}{2}((E+G) \pm \sqrt{4F^2 + (E-G)^2})$$

- **singular values** of J_f

$$\sigma_1 = \sqrt{\lambda_1} \text{ and } \sigma_2 = \sqrt{\lambda_2}$$

Measuring Distortion

- **local** distortion measure

$$E: (\mathbb{R}_+ \times \mathbb{R}_+) \rightarrow \mathbb{R}, \quad (\sigma_1, \sigma_2) \mapsto E(\sigma_1, \sigma_2)$$

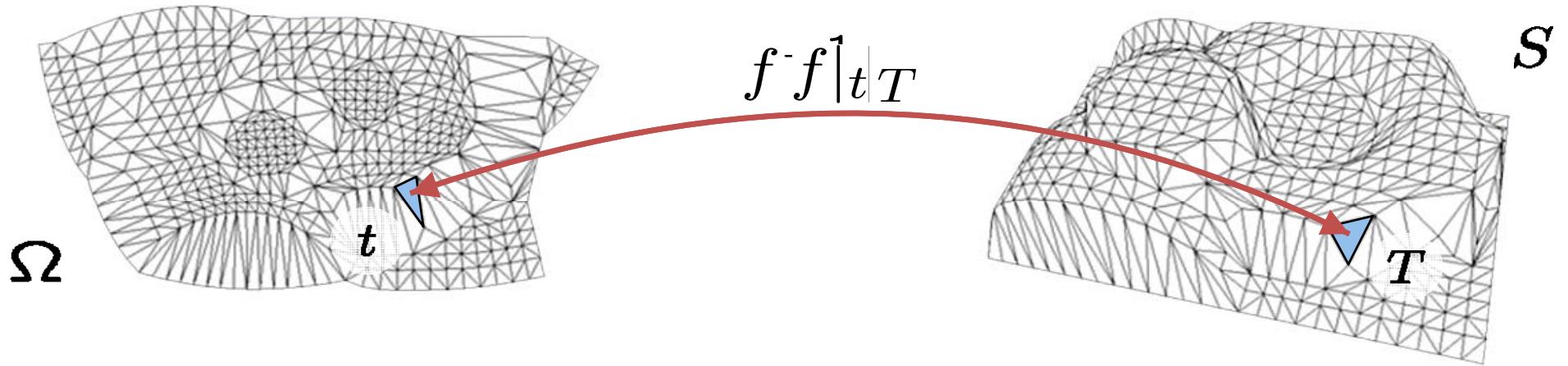
- E has **minimum** at

- $(\sigma_1, \sigma_2) = (1, 1)$ **isometric** measure
- $(\sigma_1, \sigma_2) = (x, x)$ **conformal** measure

- **overall** distortion

$$E(f) = \int_{\Omega} E(\sigma_1(u, v), \sigma_2(u, v)) du dv / \text{Area}(\Omega)$$

Piecewise Linear Parameterizations



- piecewise linear **atomic maps** $f^{-1}|_T : T \rightarrow t$
- distortion **constant** per triangle
- overall distortion $E(f^{-1}) = \sum_{T \in S} E(T)A(T) / \sum_{T \in S} A(T)$

Distortion Based Methods

- Define energy functional F as a function of $J_p, I_p, \sigma_1, \sigma_2$
- Expand their expression in F in function of the unknown u_i, v_i
- Design an algorithm to find the u_i, v_i 's that minimizes F

Linear Methods

- the terms $\sigma_1^2 + \sigma_2^2$ and $\sigma_1\sigma_2$ are **quadratic** in the parameter points u_i
- **Dirichlet** energy

$$E_D = \frac{1}{2}(\sigma_1^2 + \sigma_2^2) \geq \sigma_1\sigma_2$$

[Pinkall & Polthier 1993]
[Eck et al. 1995]

- **Conformal** energy

$$E_C = (\sigma_1 - \sigma_2)^2 / 2$$

[Lévy et al. 2002]
[Desbrun et al. 2002]

- minimization yields **linear** problem

Linear Methods

- both result in **barycentric mappings** with **discrete harmonic** weights for **interior** vertices
- **Dirichlet maps** require to **fix all** boundary vertices
- **Conformal maps** only two
 - result depends on this choice
 - best choice → [Mullen et al. 2008]
- both maps **not** necessarily **bijective**

Non-linear Methods

- **MIPS** energy [Hormann & Greiner 2000]

$$E_M = \kappa_F(J_f) = \|J_f\|_F \|J_f^{-1}\|_F = \frac{\sigma_1}{\sigma_2} + \frac{\sigma_2}{\sigma_1}$$

- **Area-preserving MIPS** [Degener et al. 2003]

$$E_\theta = \left(\frac{\sigma_1}{\sigma_2} + \frac{\sigma_2}{\sigma_1} \right) \left(\sigma_1 \sigma_2 + \frac{1}{\sigma_1 \sigma_2} \right)^\theta$$

Non-linear Methods

- **Green-Lagrange** deformation tensor [Maillot et al. 1993]

$$E_G = \|\mathbf{I}_f - \mathbf{Id}\|_F^2 = (\sigma_1^2 - 1)^2 + (\sigma_2^2 - 1)^2$$

- **Stretch** energies (L^2 , L^∞ , and symmetric stretch)

$$E_2 = \frac{1}{\sqrt{2}} \|J_f\|_F = \sqrt{(\sigma_1^2 + \sigma_2^2)/2} = \sqrt{E_D}$$

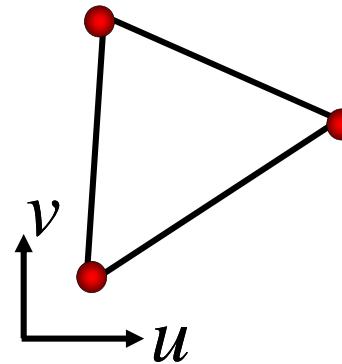
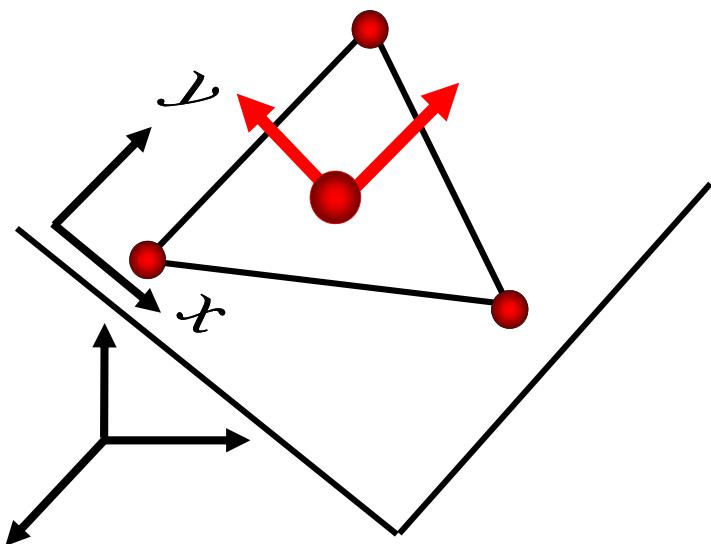
$$E_\infty = \|J_f\|_2 = \sigma_1$$

[Sander et al. 2001]

$$E_S = \max(\sigma_1, \frac{1}{\sigma_2})$$

[Sorkine et al. 2002]

Conformal Parameterization - LSCM



Cauchy Riemann equations:

No Piecewise Linear solution in general

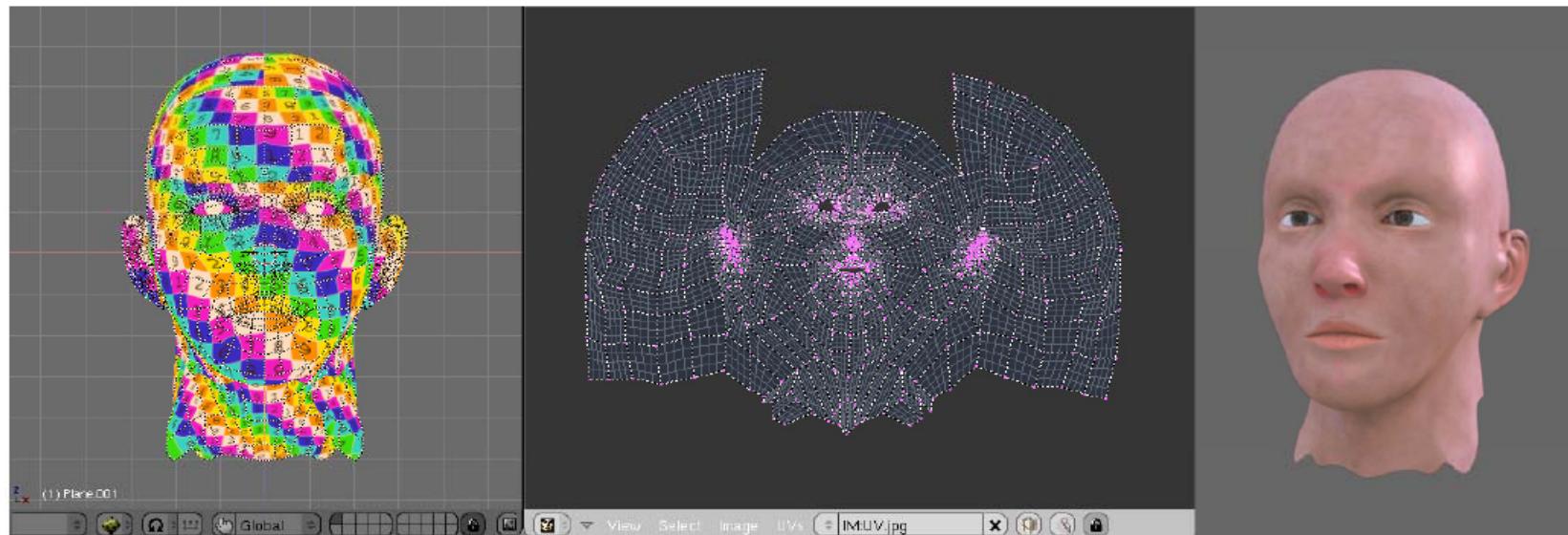
$$\begin{cases} \frac{\partial v}{\partial x} = -\frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial y} = \frac{\partial u}{\partial x} \end{cases}$$

Conformal Parameterization – LSCM

[Levy et al., 2002]

Minimize $\sum_T \left\| \begin{pmatrix} \frac{\partial v}{\partial x} \\ \frac{\partial v}{\partial y} \end{pmatrix} - \begin{pmatrix} -\frac{\partial u}{\partial y} \\ \frac{\partial u}{\partial x} \end{pmatrix} \right\|^2$

Fix two vertices to
determine rot,transl,scaling

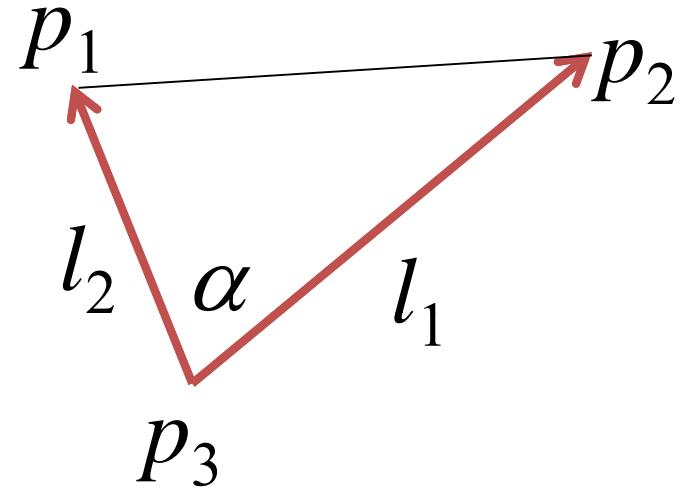


Conformal Parameterization - LSCM

Equivalent formulation, uses only **edge lengths**
ratios and angles

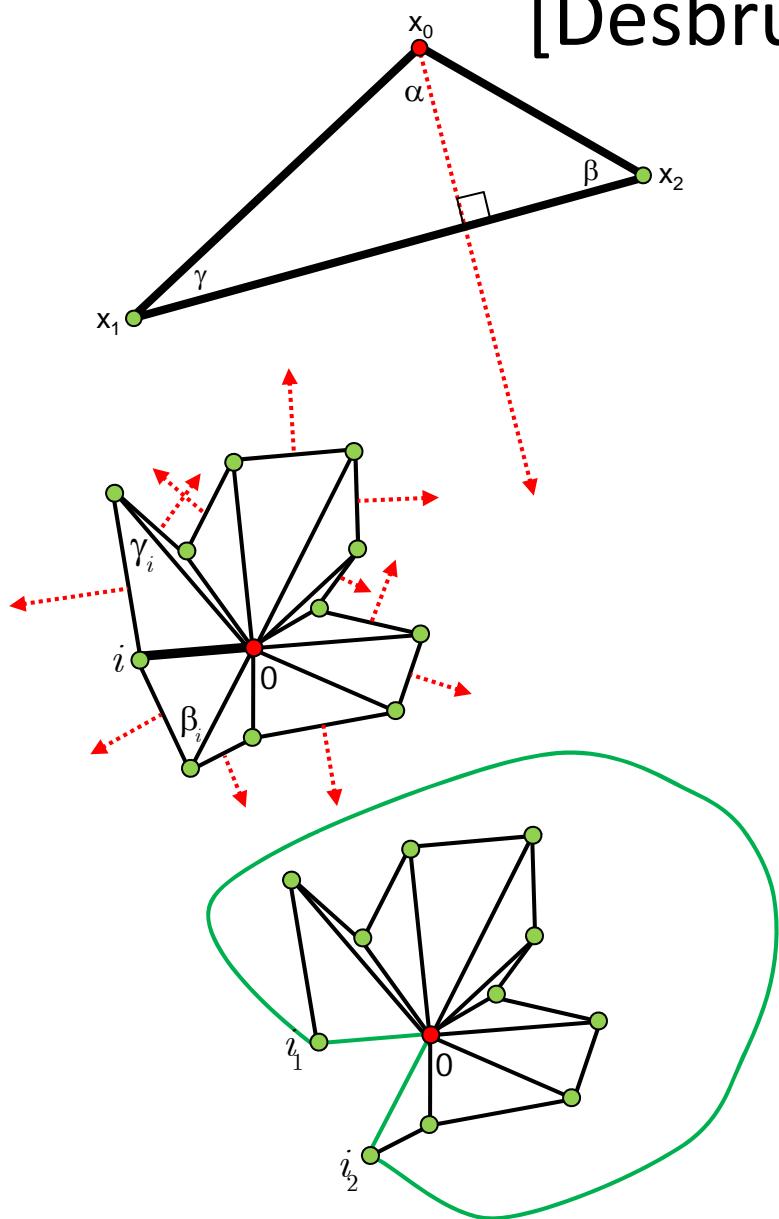
For each triangle:

$$(p_1 - p_3) = R^\alpha (p_2 - p_3) \frac{l_2}{l_1}$$



Conformal Parameterization – DCP

[Desbrun et al., 2002]



For one triangle:

$$\cot \gamma (x_2 - x_0) + \cot \beta (x_1 - x_0) = R^{90} (x_1 - x_2)$$

For complete one-ring of triangles
(interior vertex):

$$\sum_{i=1}^n (\cot \gamma_i + \cot \beta_i) (x_i - x_0) = 0$$

For incomplete one-ring of triangles
(boundary vertex):

$$\sum_{i=1}^n (\cot \gamma_i + \cot \beta_i) (x_i - x_0) = R^{90} (x_{i_1} - x_{i_2})$$

Isotropic Parameterizations

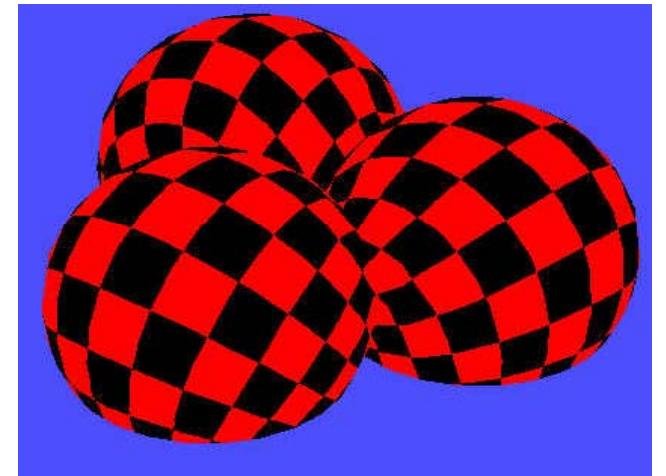
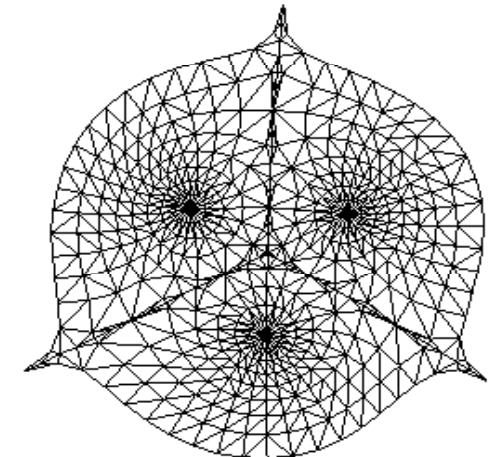
Conformal = Harmonic

$$\underbrace{\int_S \det(\mathbf{J}) ds}_{\text{area of the surface}} = \underbrace{\frac{1}{2} \int_S \|f_u\|^2 + \|f_v\|^2 ds}_{\text{Dirichlet's energy}} - \underbrace{\frac{1}{2} \int_S \|f_v - \text{rot}_{90}(f_u X)\|^2}_{\text{conformal energy}}$$

$$\underbrace{\frac{\partial X}{\partial u} \frac{\partial Y}{\partial v} - \frac{\partial X}{\partial v} \frac{\partial Y}{\partial u}}_{\det(\mathbf{J})} = \underbrace{\frac{1}{2} \left(\left\| \begin{pmatrix} \frac{\partial X}{\partial u} \\ \frac{\partial Y}{\partial u} \end{pmatrix} \right\|^2 + \left\| \begin{pmatrix} \frac{\partial X}{\partial v} \\ \frac{\partial Y}{\partial v} \end{pmatrix} \right\|^2 \right)}_{\|f_u\|^2 + \|f_v\|^2} - \underbrace{\frac{1}{2} \left(\left\| \begin{pmatrix} \frac{\partial X}{\partial v} + \frac{\partial Y}{\partial u} \\ \frac{\partial Y}{\partial v} - \frac{\partial X}{\partial u} \end{pmatrix} \right\|^2 \right)}_{\|f_v - \text{rot}_{90}(f_u)\|^2}$$

Angle Based Flattening (ABF)

- Fact: Triangular 2D mesh is defined by its angles (up to similarity)
- Define problem in angle space
- Angle based formulation:
 - Distortion as function of angles
 - Validity - set of angle constraints



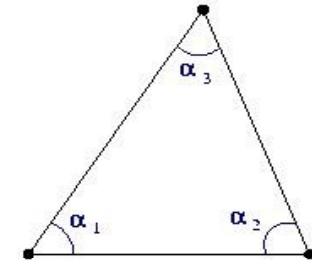
Constrained Minimization

- Notation: β_i are (given) 3D angles. α_i are (unknown) 2D angles.
- Objective: minimize (relative) deviation of angles:

$$D(\alpha_i) = \sum_{i=1}^{3T} (\alpha_i - \beta_i)^2$$

Constraints

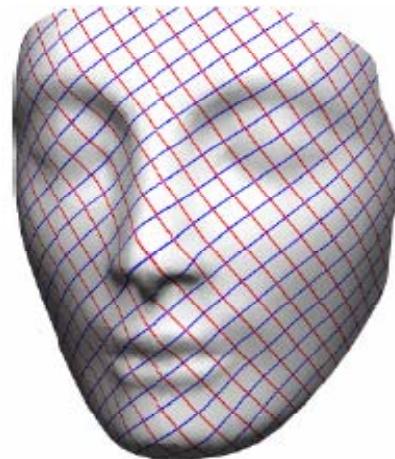
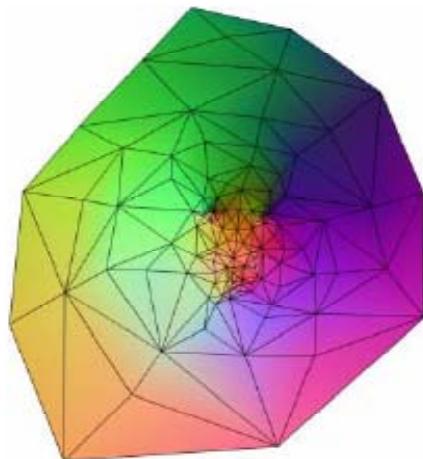
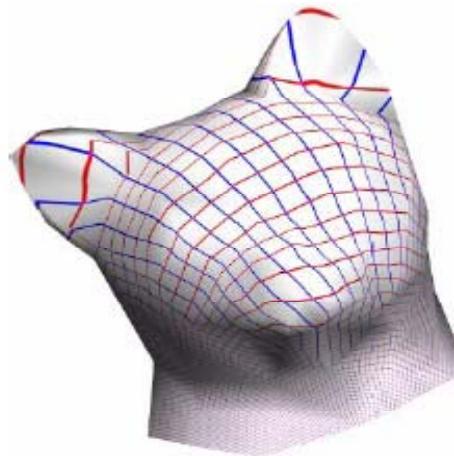
- All angles are positive (linear inequalities).
- Sum of angles in each triangle is π (linear equalities).
- Sum of angles around each vertex is 2π (linear equalities).
- All one-rings close properly (non-linear equalities).



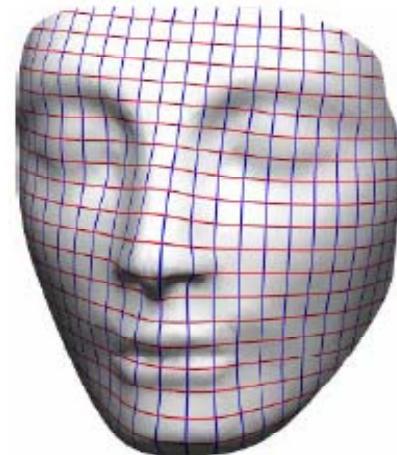
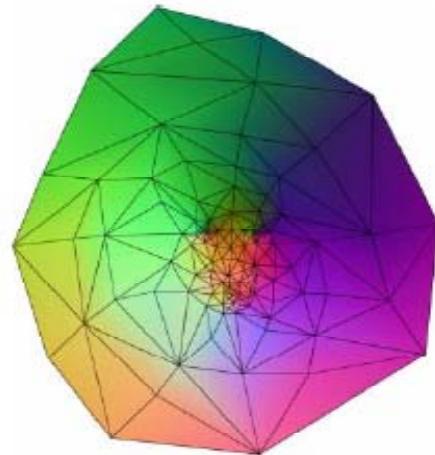
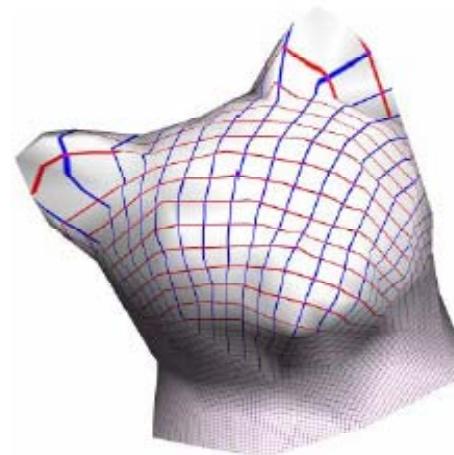
Solving

- Use non-linear solver (Lagrange multipliers, Newton method) to solve for α_i
- Use LSCM to embed in plane based on α_i

Examples

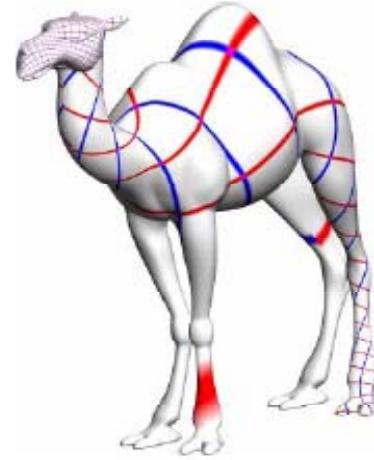
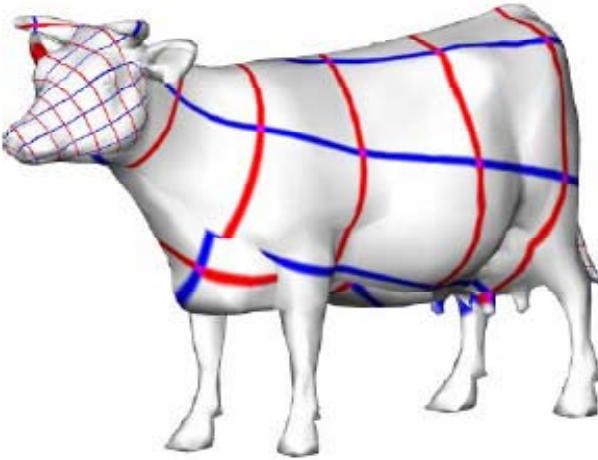


LSCM

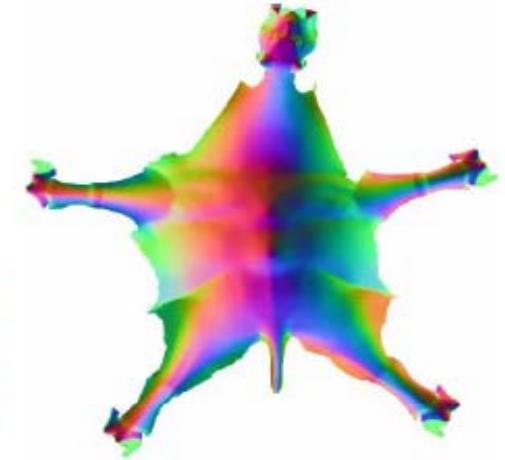
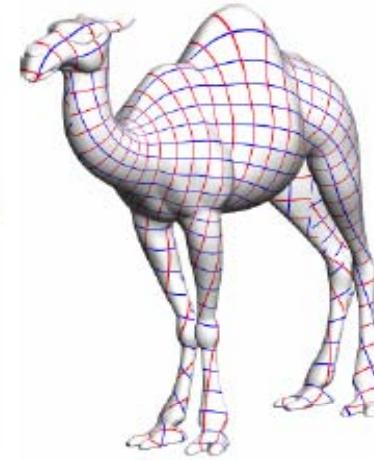
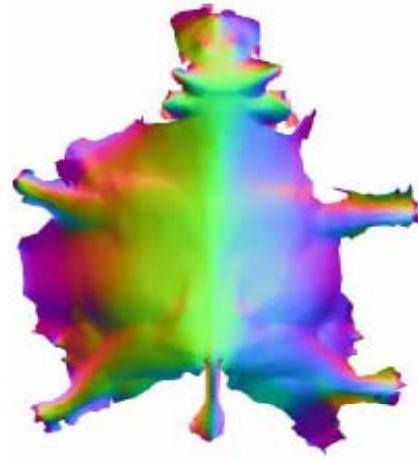
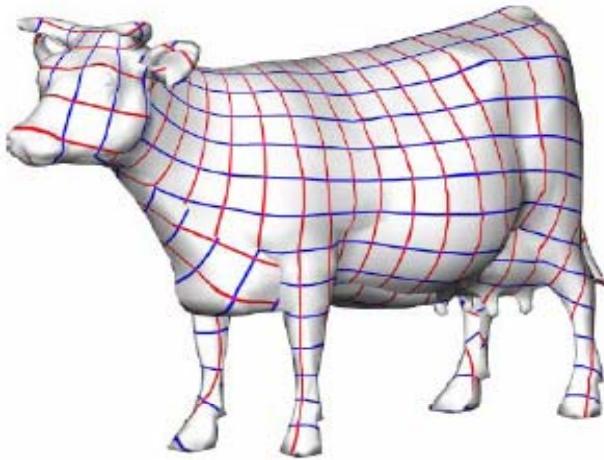


ABF

Examples

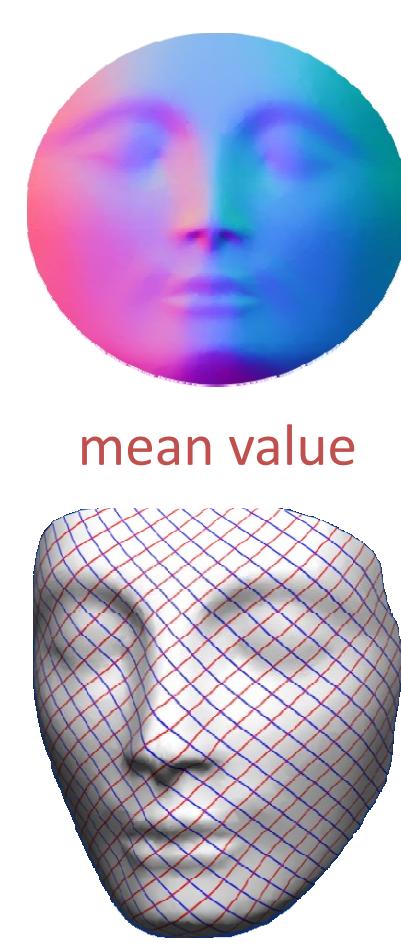
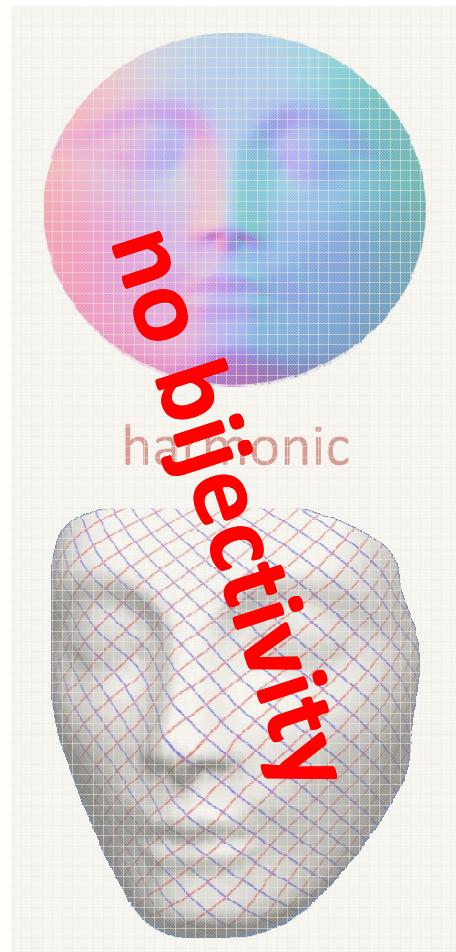
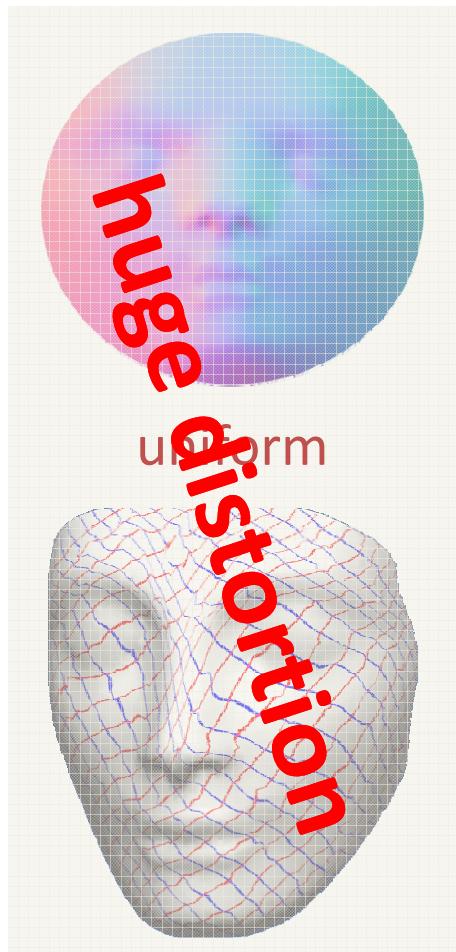


LSCM

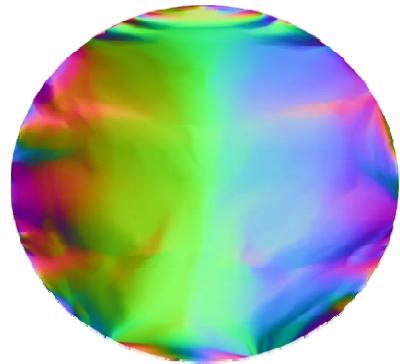


ABF

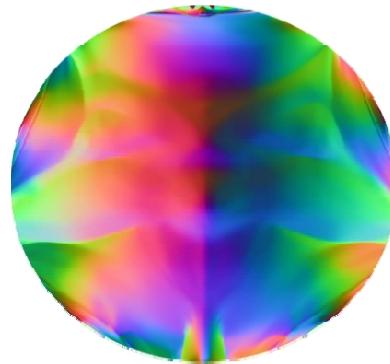
Linear Methods



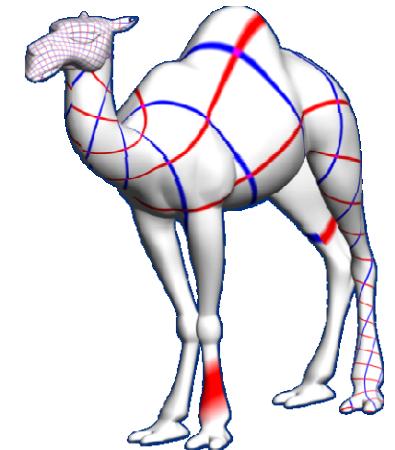
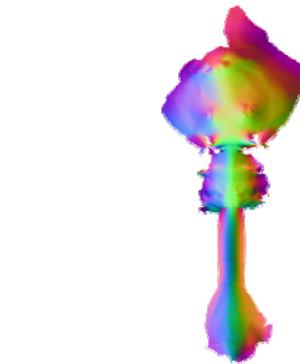
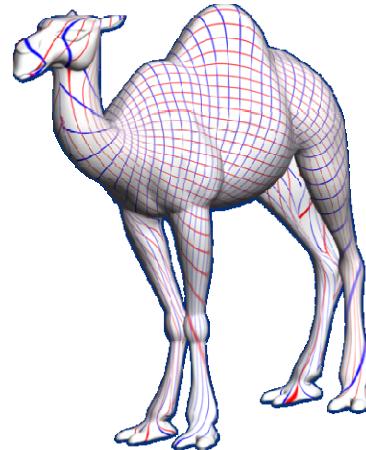
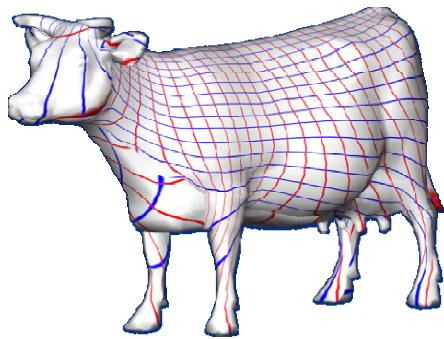
Linear Methods



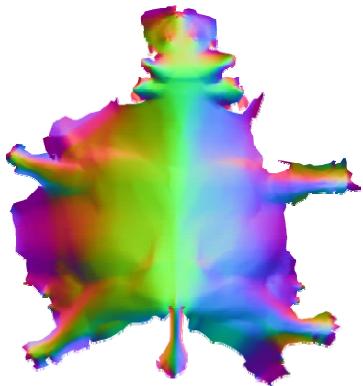
mean value



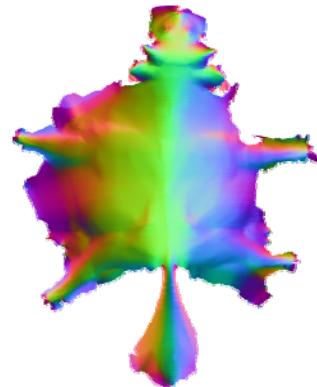
conformal



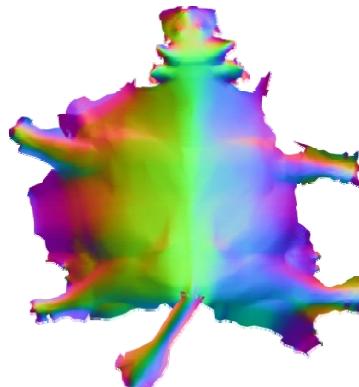
Non-Linear Methods



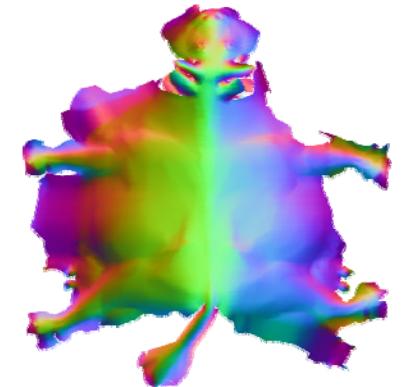
ABF++



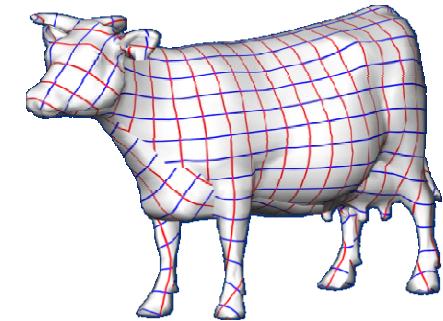
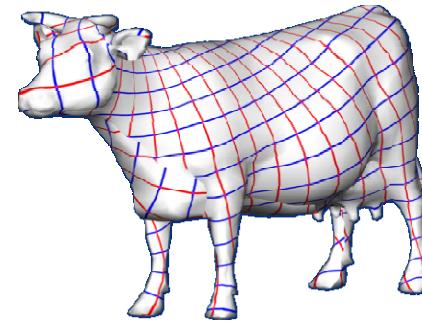
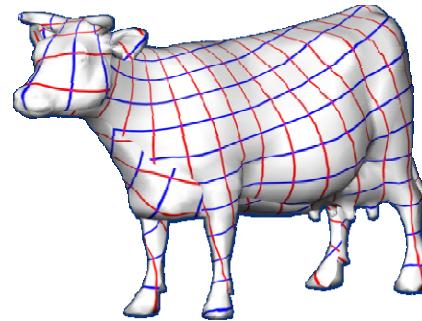
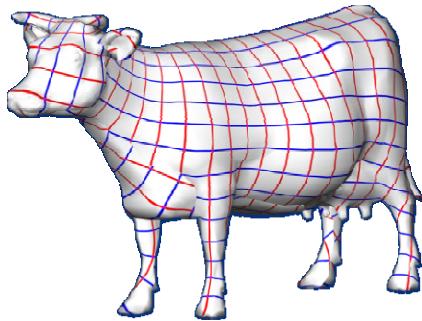
circle patterns



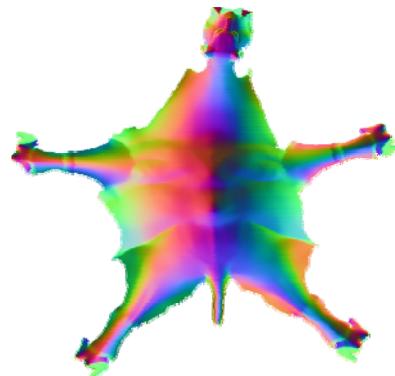
MIPS



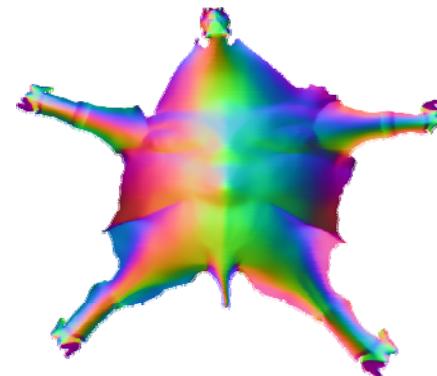
stretch



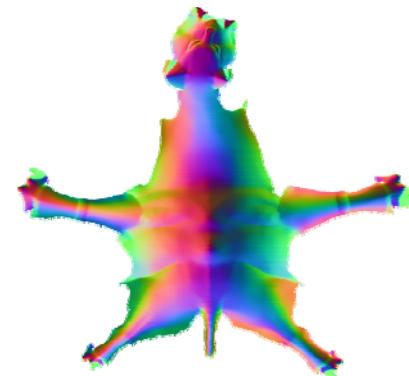
Non-Linear Methods



ABF++



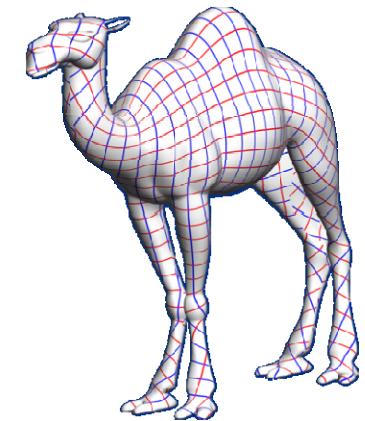
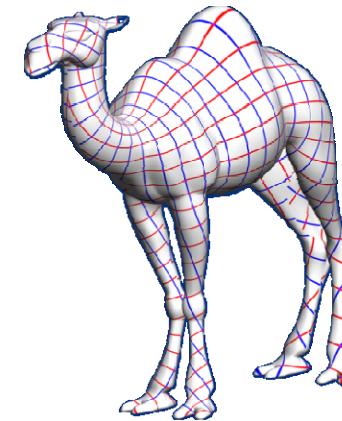
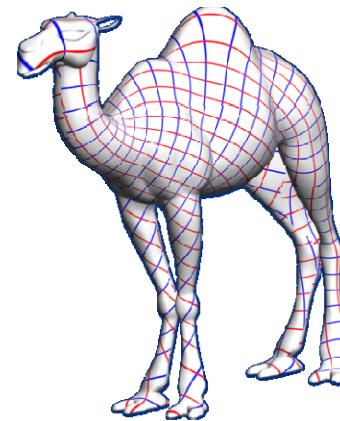
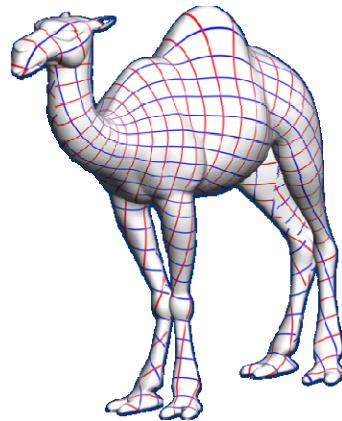
circle patterns



MIPS



stretch



Curvature Prescription

LSCM [2002]

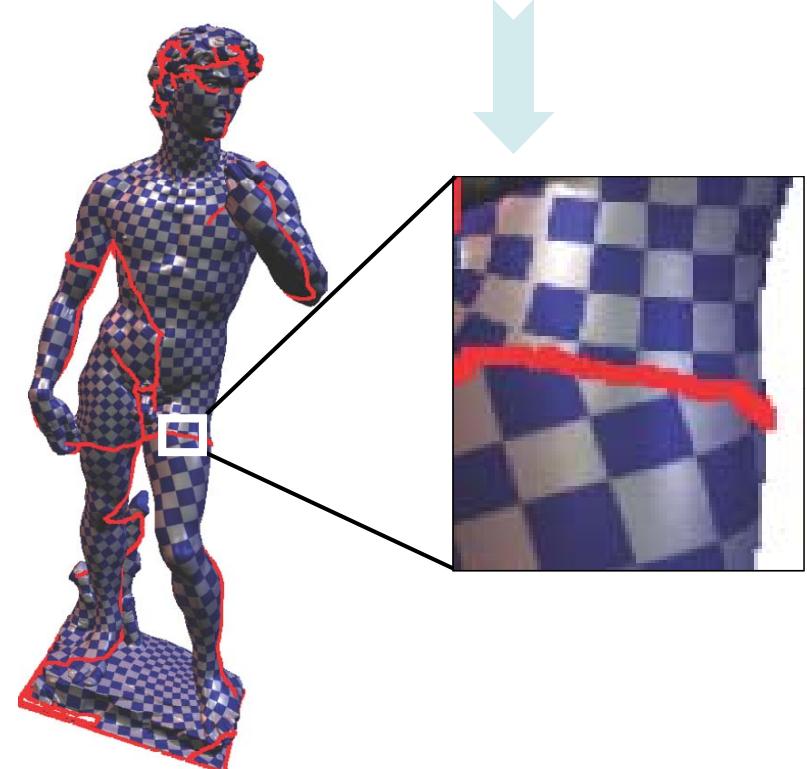
ABF++ [2005]

LinABF [2007]

(and many more...)



1. Cut mesh to disk
2. **Compute new angles/lengths**
3. Embed in plane
→ Discontinuities in scale



Circle Patterns [2006]

Ricci Flow [2007]

CPMS, CETM [2008]

Curvature Prescription

LSCM [2002]

ABF++ [2005]

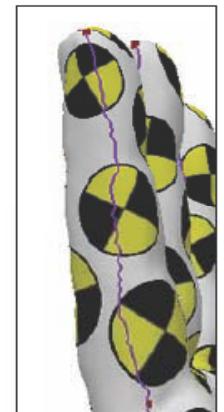
LinABF [2007]

(and many more...)

Circle Patterns [2006]

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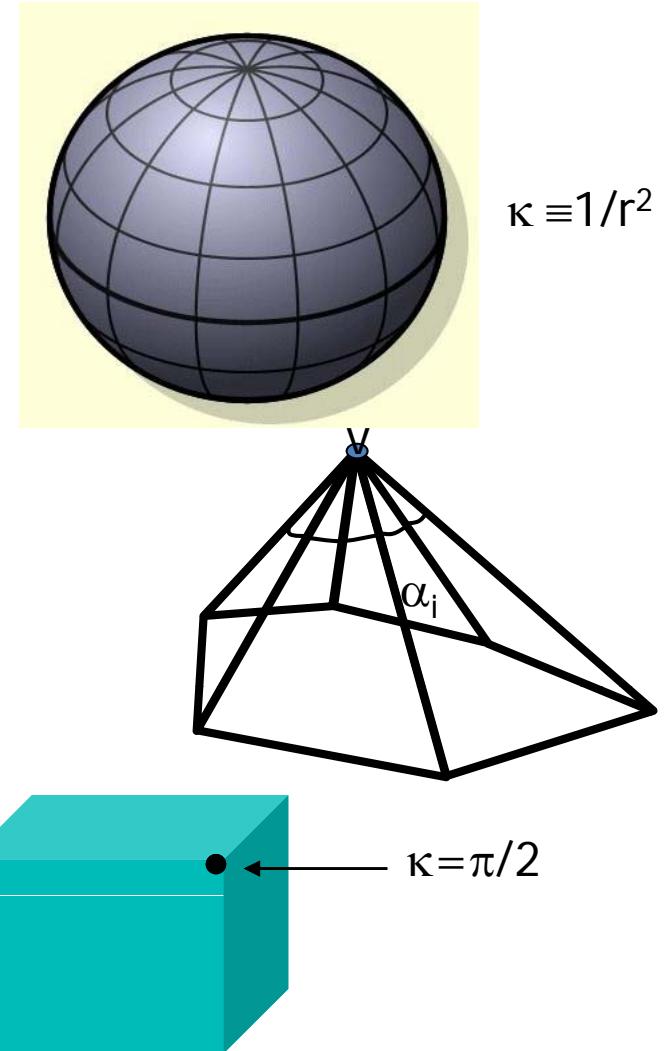


1. **Compute new angles/lengths**
2. Cut Mesh to disk
3. Embed in plane
→ No discontinuities in scale

Gaussian Curvature

Continuous definition
Scale dependent

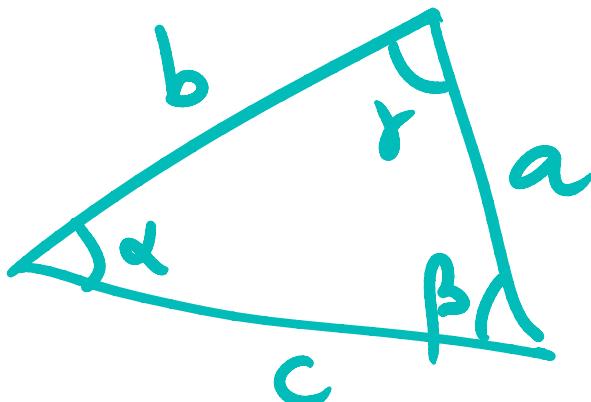
Gauss-Bonnet theorem: $\int \kappa = 2\pi\chi$



Discrete definition: $\kappa(v) = 2\pi - \sum \alpha_i$
Scale independent !

Gauss-Bonnet theorem: $\sum \kappa = 2\pi\chi$

Discrete Metric



Triangle edge lengths (a, b, c)
are the discrete metric

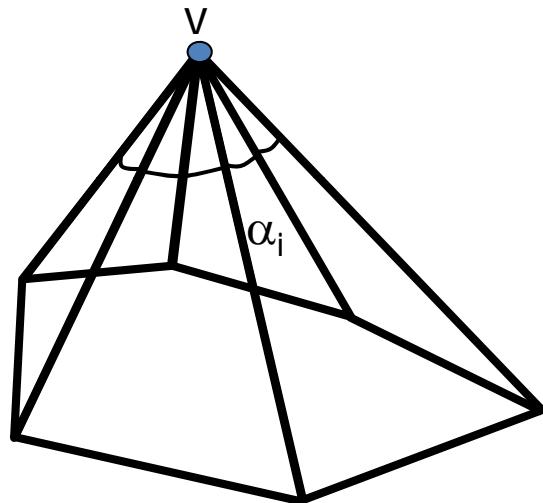


The metric defines the angles by the cosine law

$$\cos \alpha = \frac{b^2 + c^2 - a^2}{2bc}$$



The angles define the Gaussian curvature



$$k = 2\pi - \sum \alpha_i$$

CPMS [2008]

Original edge lengths (metric)

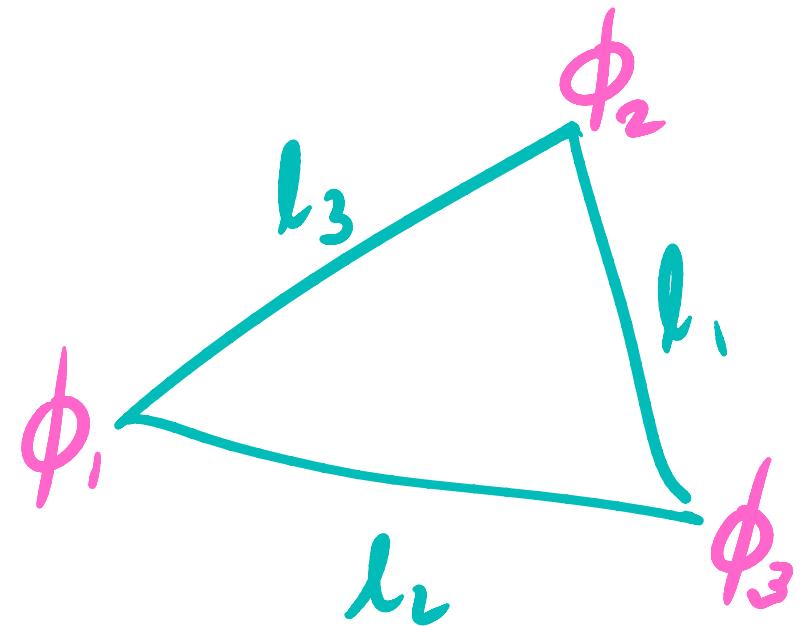
Conformal factor (per vertex)

Extend ϕ to edge

$$\phi(t) = \phi_2 t + \phi_3 (1 - t)$$

Integrate e^ϕ on edge

Scale to get new edge lengths



$$\tilde{l}_1 = l_1 \int e^{\phi(t)}$$

What Happens to the Curvature?

- Continuous conformal map

$$e^{2\phi} \tilde{\kappa} = \kappa - \nabla^2 \phi$$

Curvature after map Curvature before map Conformal factor

```
graph LR; A[Curvature after map] --> B[e^{2\phi} \tilde{\kappa}]; C[Curvature before map] --> D[\kappa]; E[Conformal factor] --> F[\nabla^2 \phi];
```

What Happens to the Curvature?

- Continuous conformal map

$$e^{2\phi} \tilde{\kappa} = \kappa - \nabla^2 \phi$$

- Discrete conformal map (for small changes)

$$\tilde{\kappa} \approx \kappa - \nabla^2 \phi$$

- Where did the exponent go?
 - Discrete curvature doesn't scale!

The Flattening Algorithm for a Topological Disk

- Set target curvature to 0 on interior vertices

- Set ϕ to 0 on the boundary

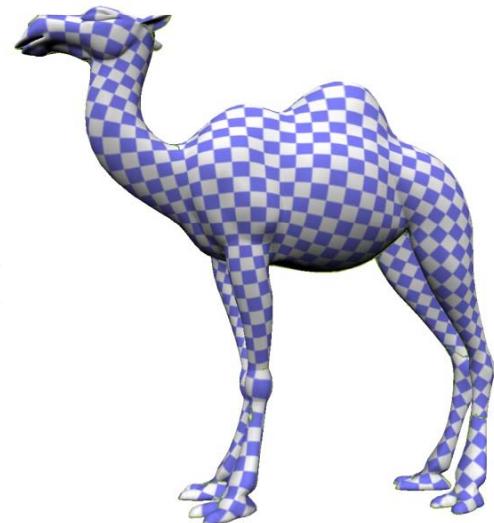
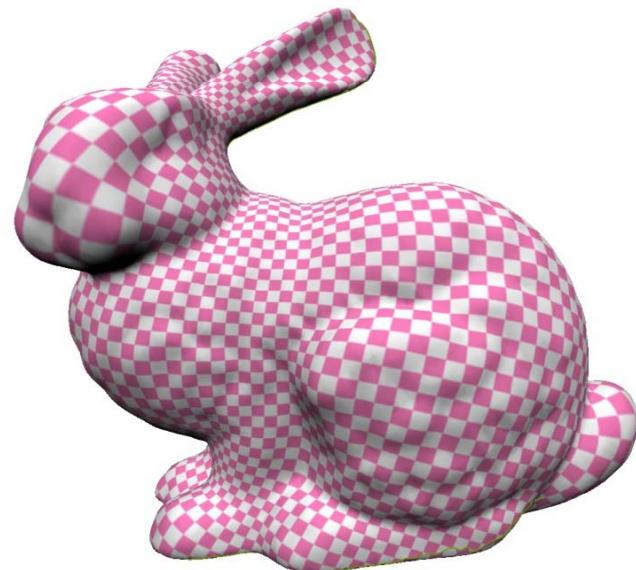
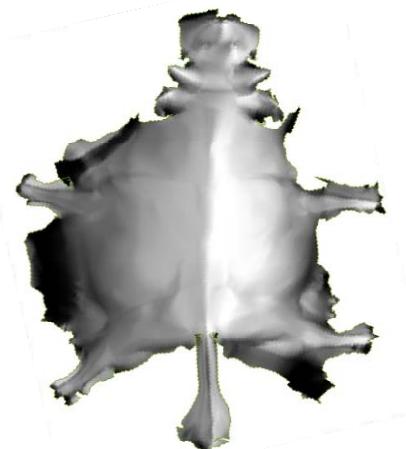
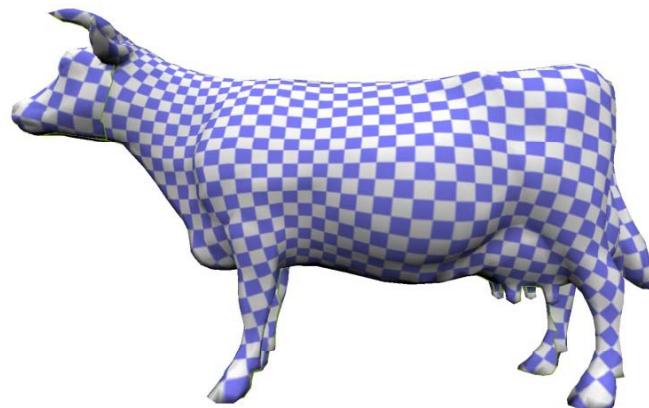
- Find ϕ on interior vertices by solving

$$\nabla^2 \phi = \tilde{\kappa} - \kappa$$

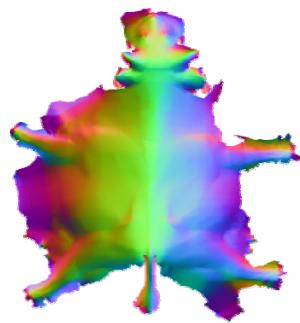
- Scale edge lengths to get conformal metric

- Embed new edge lengths using LSCM

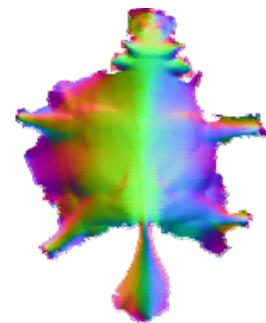
Some Results



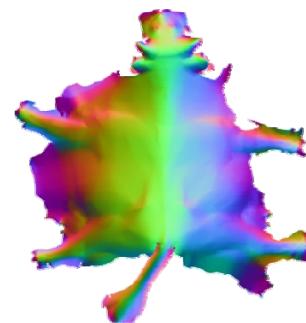
Comparison with Non-Linear Methods



ABF++



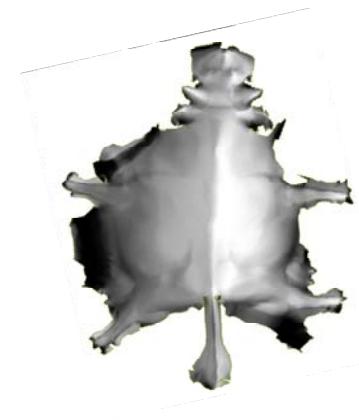
circle patterns



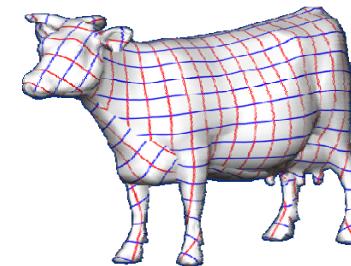
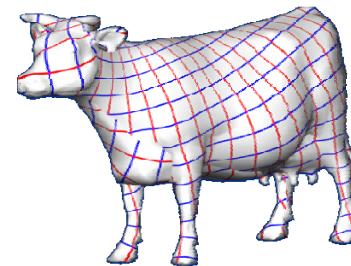
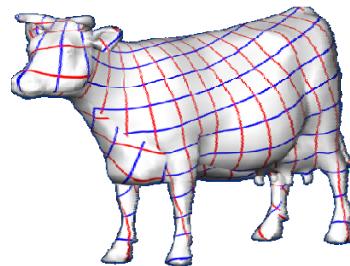
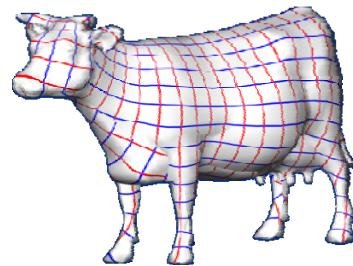
MIPS



stretch

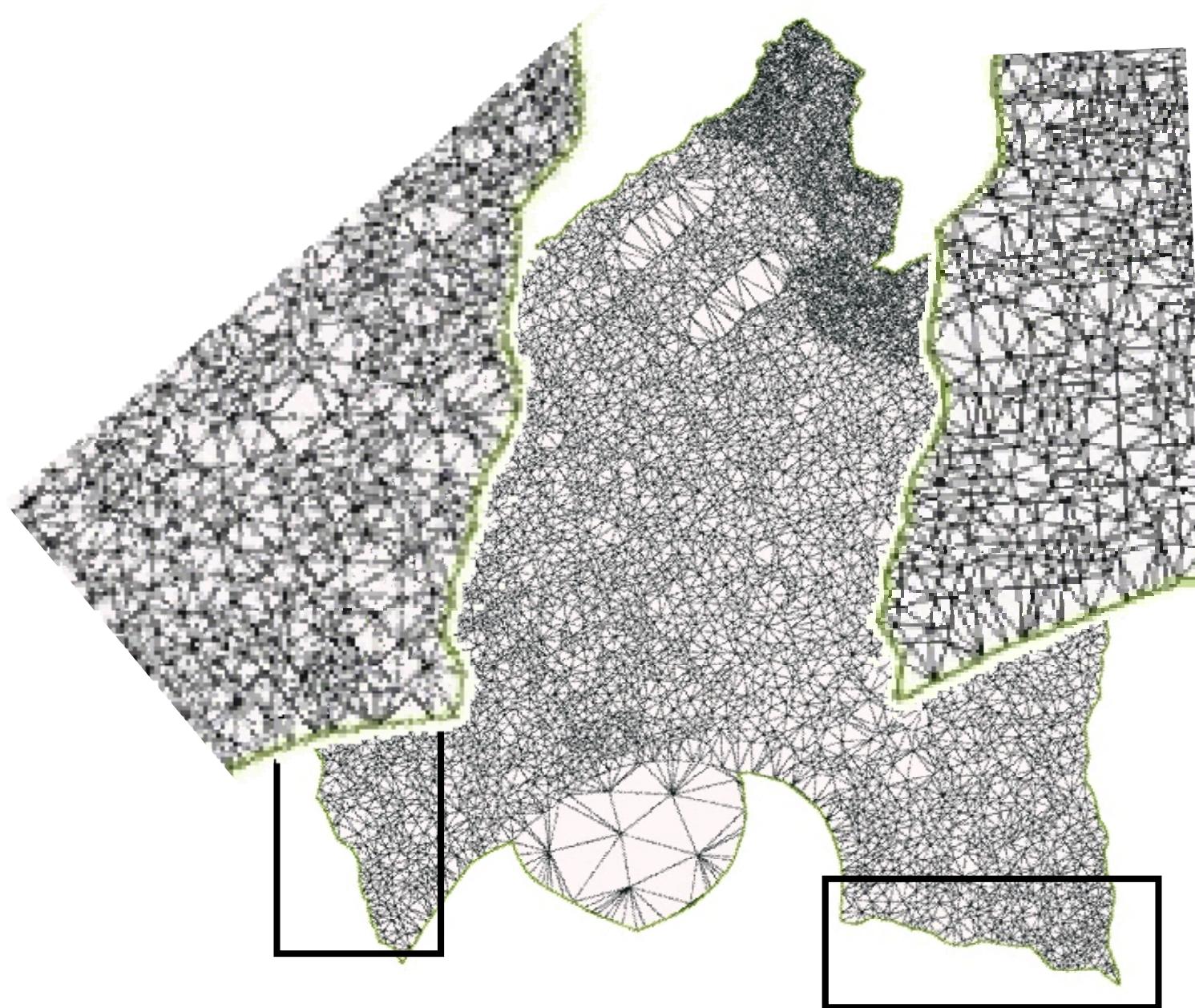


CPMS



Reducing Area Distortion

- Gather Gaussian curvature into “cone points”
 - Target curvature not 0 everywhere
- First compute new edge lengths, then cut through cone points
- No discontinuities in scale



Conformal Equivalence of Triangle Meshes [2008]

Definition [Luo 2004]

Two discrete metrics $\ell, \tilde{\ell}$ on M are (*discretely*) *conformally equivalent* if

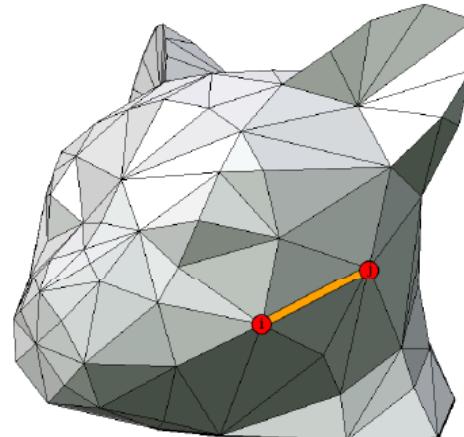
$$\tilde{\ell}_{ij} = e^{\frac{1}{2}(u_i + u_j)} \ell_{ij}$$

for some function $u : V \rightarrow \mathbb{R}$

- use $\lambda_{ij} = 2 \log \ell_{ij}$

so $\ell_{ij} = e^{\lambda_{ij}/2}$

and $\tilde{\lambda}_{ij} = \lambda_{ij} + u_i + u_j$



Conformal Equivalence of Triangle Meshes [2008]

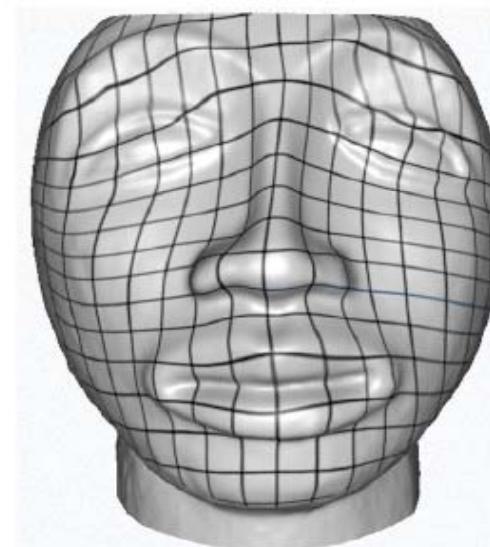
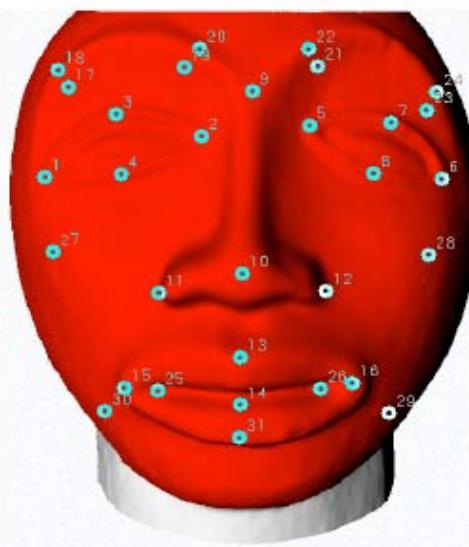
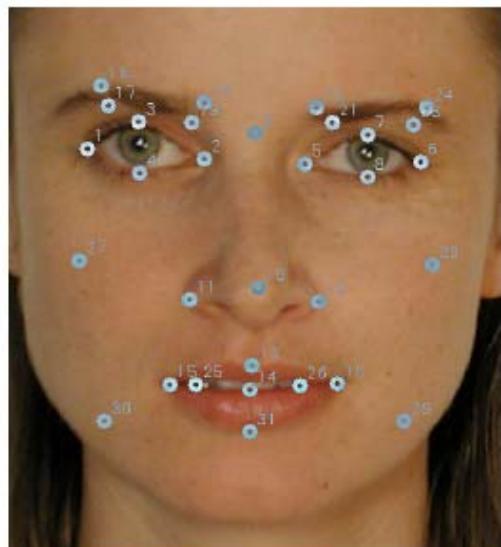
- Given input mesh and target curvatures, find u s.t.
 - New mesh is conformally equivalent to source mesh
 - Target curvatures are achieved
- Minimize convex energy \rightarrow global minimum
- First optimization step equivalent to CPMS

Conformal Equivalence of Triangle Meshes [2008]

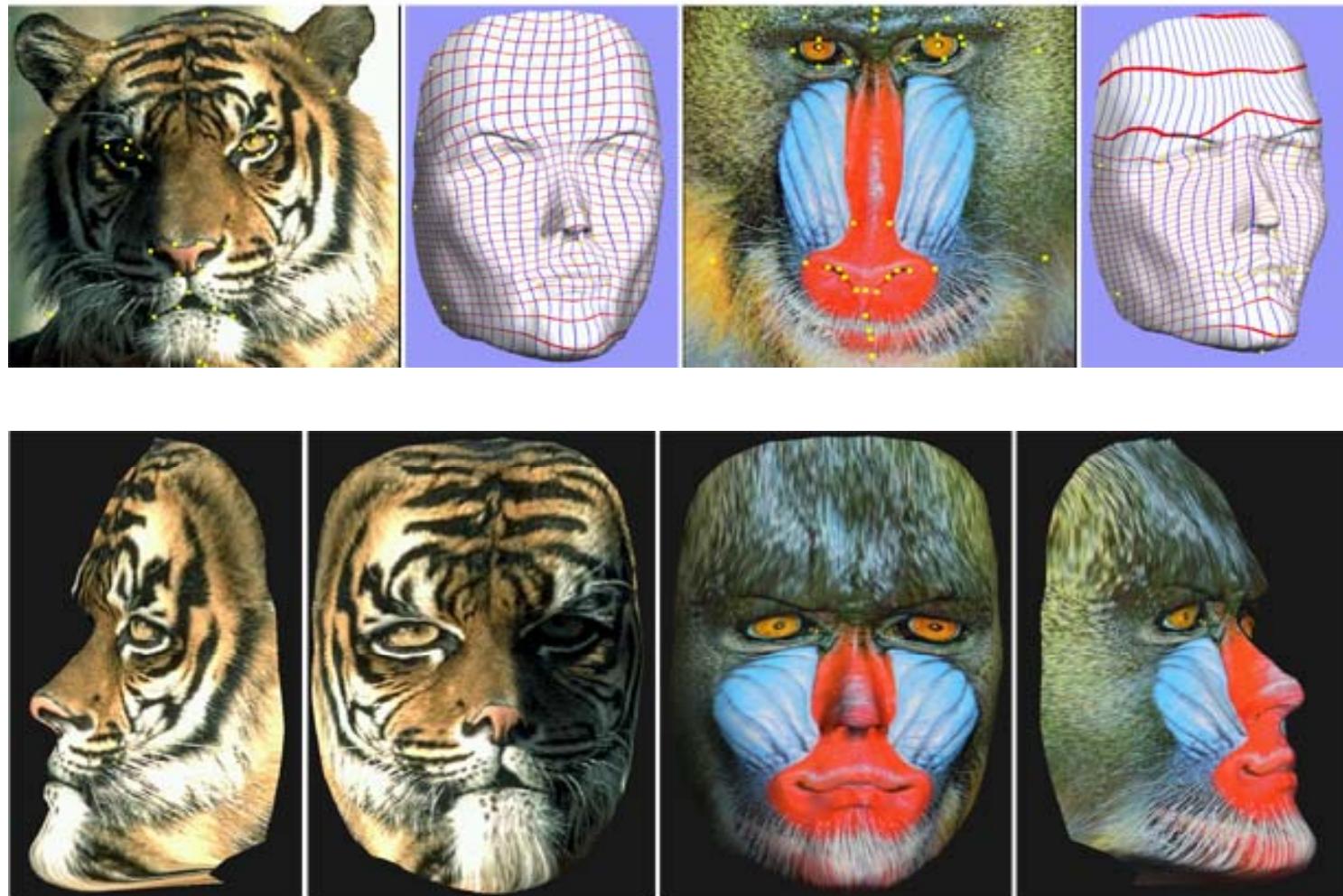


Constrained Parameterizations

- Construct well-behaved mapping while satisfying constraints



Constrained Parameterizations



Constrained Parameterizations

- Objective function

$$C(f) = \sum_{j \in M} \underbrace{(f(\mathbf{u}_j) - \mathbf{v}_j)^2}_{\text{fitting constraints}} + \epsilon \underbrace{\int_S \gamma(\mathbf{u}) d\mathbf{u}}_{\text{distortion}}$$

points in domain constrained positions

set of constraints

Parameterization - Conclusions

- Many MANY methods out there
 - Fixed / free boundary
 - Bijective / non bijective
 - Conformal / area preserving
 - Linear / non linear
 - First cut then embed / cone points
- Best method depends on mesh
 - Close to disk → linear methods can work well
 - Require large distortion → prefer non-linear methods