Geometric Modeling

Shape Interpolation, Parametrization

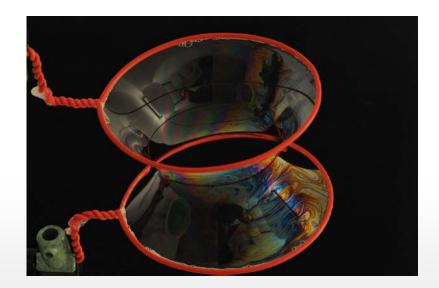


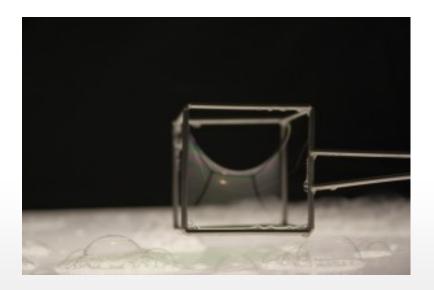
Last Lecture

Examples

Minimal Surfaces

- Surfaces with vanishing mean curvature are called minimal surfaces
- They are saddle shaped at every point
- Solution of Plateau's problem (soap films, minimal area)
 - Soap bubbles have constant mean curvature

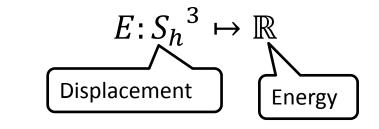


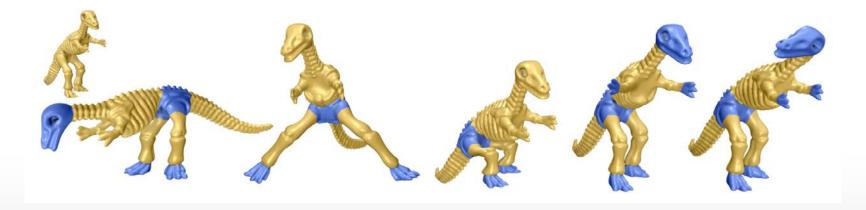


Deformation Energies

General deformation energies

 A deformation energy measures the "energy" stored in a deformation (or the "cost" of a deformation)





Constraints

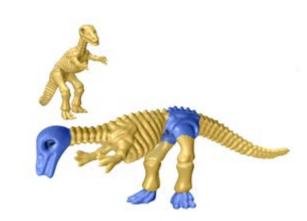
To deform the object the user sets constraints

Hard constraints:

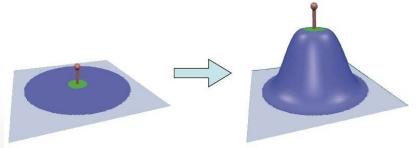
$$Au = a$$

Soft constraints:

$$E_C(u) = \frac{1}{2} ||Au - a||^2$$



- A is a rectangular matrix, a is a vector
- Use masses for irregular meshes



Quadratic Program

Soft constraints

- Minimize weighted sum of deformation energy E_L and constraints energy E_C over all displacements $u \in S_h^{-3}$
 - $\lambda \in \mathbb{R}_{>0}$

$$E(u) = E_L(u) + \lambda E_C(u)$$

- Necessary condition for a minimum u^* is $\nabla E(u^*) = 0$
- Since E is quadratic and positive definite, this is also a sufficient condition

$$\nabla E(u) = (SM^{-1}S + \lambda A^{T}A)u - \lambda A^{T}a$$

Computing the Deformation

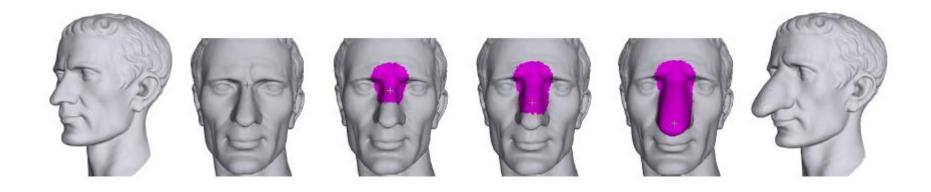
Linear system

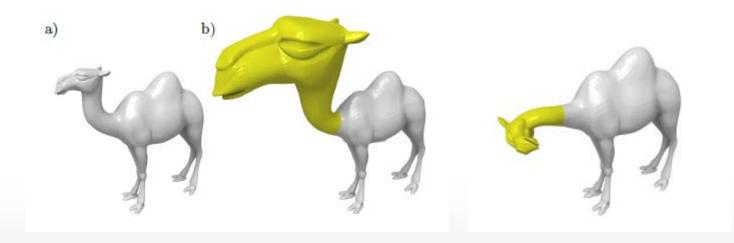
• To compute the deformation, the linear system $(SM^{-1}S + \lambda A^TA)u = \lambda A^Ta$

has to be solved

- The matrix $(SM^{-1}S + \lambda A^T A)$ is
 - sparse
 - symmetric, positive definite
- An efficient solver is a sparse Cholesky decomposition
- Since changing the positions of the handles only changes the right-hand side, the factorization can be re-used and interactive modeling is possible

Brushes





Brushes

Gradient-Based Editing

Modify the gradients of the embedding x:

$$Gx \xrightarrow{\text{editing of} \atop \text{gradients}} \widetilde{g}$$

- Find the displacement u such that the gradients of $\tilde{x} = x + u$ best match \tilde{g}
- Poisson reconstruction: Minimize

$$E_{PR}(\widetilde{x}) = \frac{1}{2} \int_{M} \|\nabla \widetilde{x} - \widetilde{g}\|^{2} dA$$

• Euler-Lagrange equation $\nabla E(\tilde{x}) = 0$ is

$$S\widetilde{x} = G^T M_V \widetilde{g}$$

Limitation: Large Deformations

This image current currently be objected	-	I	I	l	
	Approach	Pure Translation	120° bend	135° twist	70° bend
	Original model				4
-	Non-linear prism-based modeling [12]	25			
	Thin shells [10] + deformation transfer [14]	35			
	Gradient-based editing [72]	THE P. LEWIS CO.			F
	Laplacian-based editing with implicit optimization [60]				Y
	Rotation invariant coordinates [42]				F

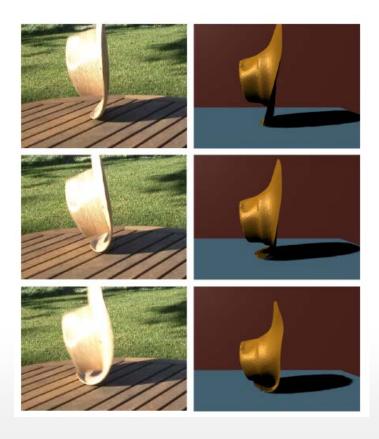
Non-Linear Energies

Discrete thin shell energy [Grinspun et al. 03]:

 $E_{TS} = E_F + \lambda \left(E_A + E_L \right)$

Flexural Energy

Membrane Energy



Discrete Shells

Discrete thin shell energy [Grinspun et al. 03]:

$$E_{TS} = E_F + \lambda \left(E_A + E_L \right)$$
Flexural Energy

Membrane Energy

• Flexural energy: length of e;

$$E_F = \frac{3}{2} \sum_{i} \frac{\|\overline{e_i}\|^2}{\overline{A}e_i} \left(\theta_{e_i} - \overline{\theta}_{e_i}\right)^2$$
 area of star e_i dihedral angle

Membrane energy:

$$E_L = \frac{1}{2} \sum_{i} \frac{1}{\|\bar{e}_i\|} (\|e_i\| - \|\bar{e}_i\|)^2$$

$$E_A = \frac{1}{2} \sum_{i} \frac{1}{\bar{A}_i} (A_i - \bar{A}_i)^2$$

Non-Linear Deformations



Further Reading

Survey on linear editing schemes

Botsch, M. and Sorkine, O. 2008. On linear variational surface deformation methods. *IEEE Transactions on Visualization and Computer Graphics* 14, 1, 213–230.

Interactive non-linear editing

JACOBSON, A., BARAN, I., KAVAN, L., POPOVIĆ, J., AND SORKINE, O. 2012. Fast automatic skinning transformations. ACM Trans. Graph. 31, 4, 77:1–77:10.

HILDEBRANDT, K., SCHULZ, C., VON TYCOWICZ, C., AND POLTHIER, K. 2011. Interactive surface modeling using modal analysis. ACM Trans. Graph. 30, 5, 119:1–119:11.

Shape Interpolation

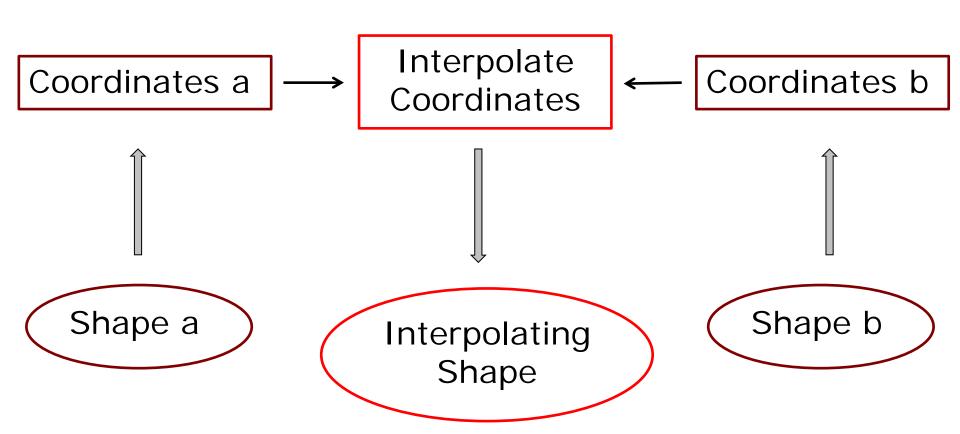
Shape Interpolation

Blending of bunny and rabbit

 Both shapes have the same mesh (hence a correspondence between the shapes is given)



General Framework

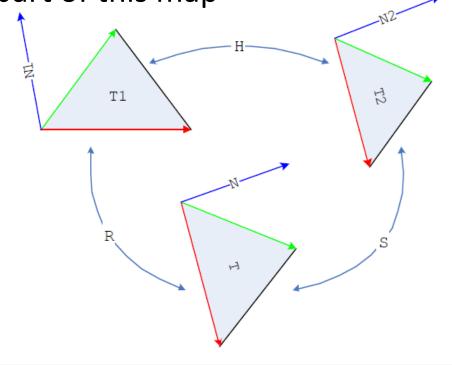


Interpolation

Interpolation between two triangles

- For every pair T_1 , T_2 of corresponding triangles on the two shapes, there is a unique affine map of \mathbb{R}^3 mapping T_1 to T_2 and the normal N_1 to the normal N_2
- Let H denote the linear part of this map
- Polar decomposition:

$$H = R S$$
 R is a rotation and S is symmetric



SVD & Polar Decomposition

Singular value decomposition (SVD)

- Let H be an arbitrary real matrix (may be rectangular)
- Then H can be written as:
 - $\bullet H = U D V^T$
 - The matrices U, V are orthogonal
 - D is a diagonal matrix (might contain zeros)

Polar decomposition from SVD

- $H = U DV^T = UV^T V DV^T = R S$
 - $R = UV^T$
 - $S = V DV^T$

Polar Decomposition

Alternative Computation (when H has full rank)

- Compute eigenvalues and –vectors of H^TH $H^TH = W^T \Lambda W$
 - *W*: orthognal matrix listing the eigenvectors
 - Λ : diagonal matrix listing the eigenvalues
- Set $S = W^T \sqrt{\Lambda} W$
 - $\sqrt{\Lambda}$ diagonal matrix listing the square roots of the eigenvalues
- Set $R = HS^{-1}$
 - Then H = RS

Why is R orthogonal?

•
$$R^T R = (HS^{-1})^T HS^{-1} = \dots = Id$$

Reminder: Rotation Axis and Angle

Rotation axis and angle of a rotation matrix

- The eigenvector with eigenvalue 1 points in direction of the rotation axis
- The angle θ of a rotation matrix R satisfies $\operatorname{trace}(R) = 1 + 2\operatorname{cos}(\theta)$

Hence
$$\theta = \cos^{-1}\left(\frac{\operatorname{trace}(R)-1}{2}\right)$$

Interpolate Triangles

Interpolate Rotations

- To interpolate the rotation R, we fix the rotation axis and linearly interpolate the angle
- To interpolate *H*:

$$H(t) = R(t)((1-t)Id + tS)$$
[Interpolate the rotation over [0,1]]

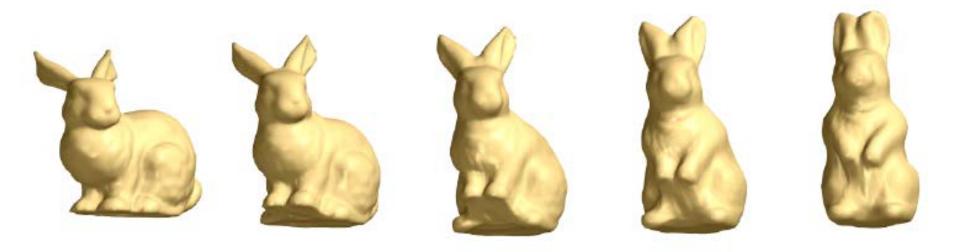
Interpolate the translation by barycentric interpolation

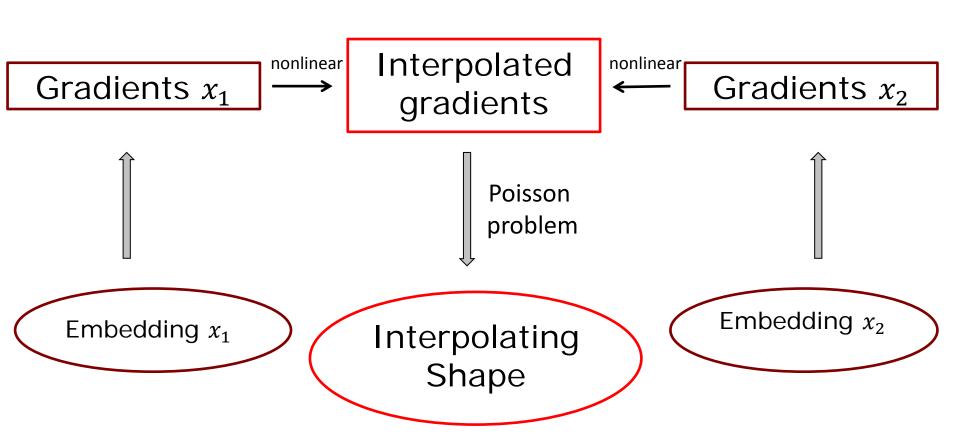
Interpolating the surfaces

- Using H(t), we obtain an interpolation of all pairs triangles of the two surfaces, but the interpolating triangles are not connected (triangle soup)
- Problem: Find are connected surface that best matches the unconnected triangle soup
- Idea: Use Poisson reconstruction
 - Compute gradients of the interpoating triangles and construct a surface that best matches the gradients

Representing the shape as functions

- One triangle mesh M and two embeddings x_1 and x_2
 - M can be one of the shapes





[Xu, Zhang, Wang, Bao 2005]

Poisson reconstruction

- For any t, we can reconstruct a shape x(t) that best matches the gradients g(t)
- Minimize

$$E_{PR}(\mathbf{x}) = \frac{1}{2} \int_{M} \|\nabla x - g\|^{2} dA$$

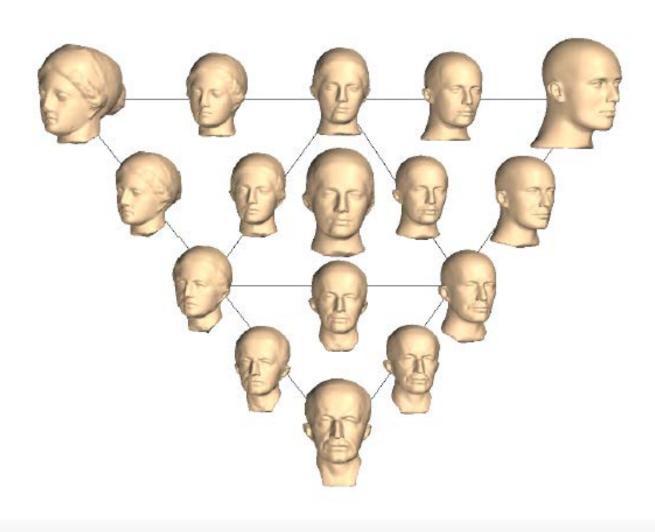
• Euler-Lagrange equation $\nabla E(x) = 0$ is $Sx = G^T M_V g$

• This means that the x(t) satisfy $Sx(t) = G^T M_V g(t)$

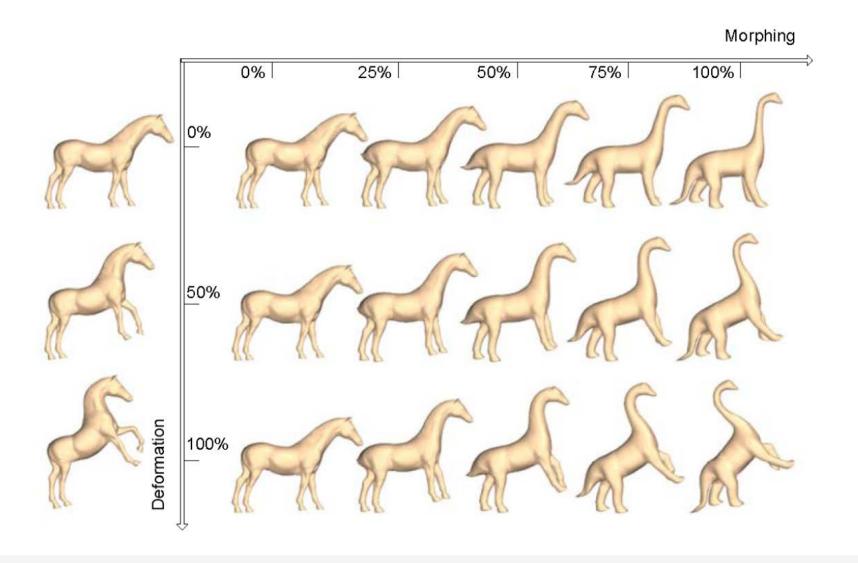
Results



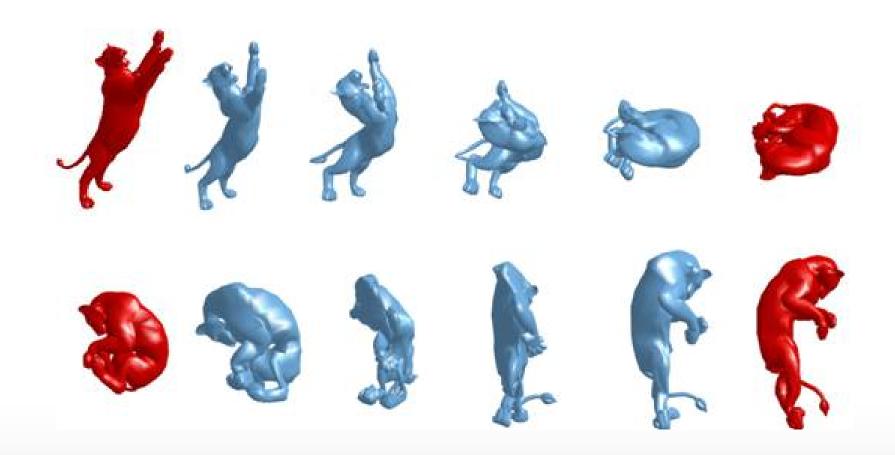
Results



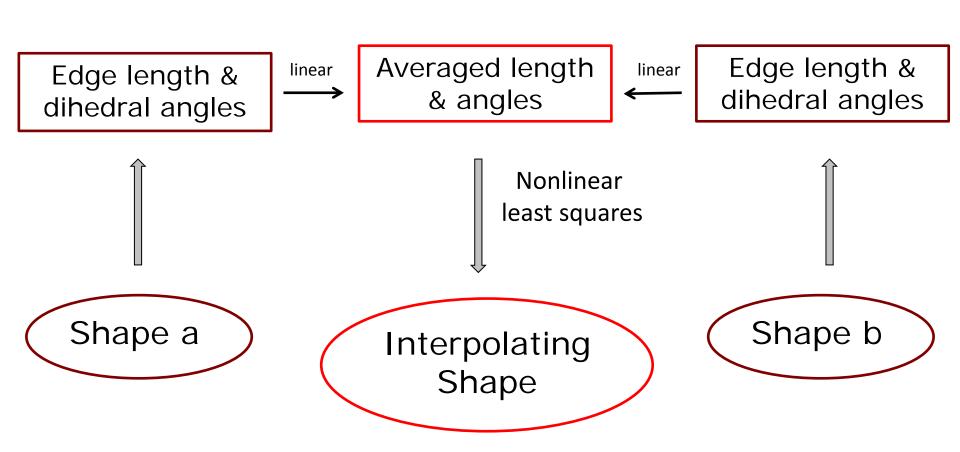
Deformation Transfer



Limitations

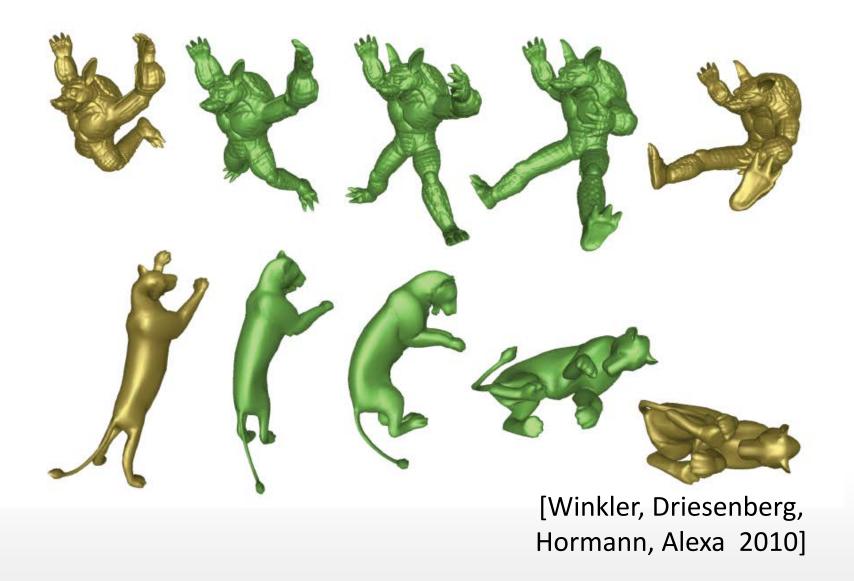


Dihedral Angles and Edge Length

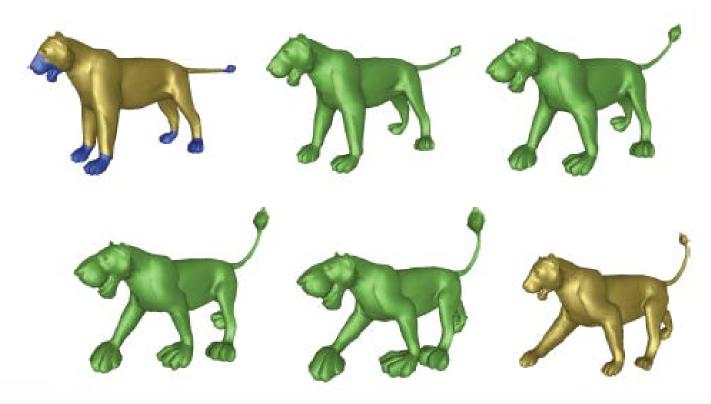


[Winkler, Driesenberg, Hormann, Alexa 2010]

Results



Results



[Winkler, Driesenberg, Hormann, Alexa 2010]

Extrapolation



[Winkler, Driesenberg, Hormann, Alexa 2010]

Real-Time Nonlinear Shape Interpolation

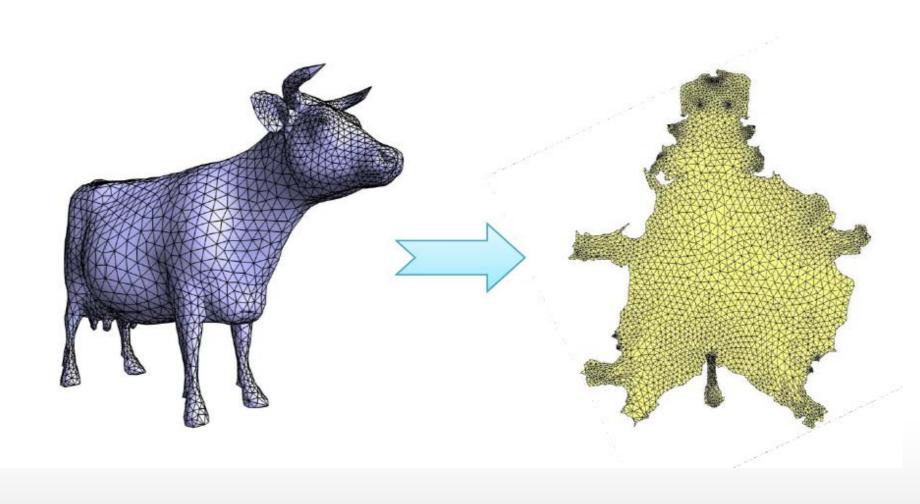
Real-time Nonlinear Shape Interpolation

(no audio)

Real-Time Nonlinear Shape Interpolation
Christoph von Tycowicz, Christian Schulz, Hans-Peter Seidel, Klaus Hildebrandt
ACM Transactions on Graphics 34(3) 2015

Parametrization

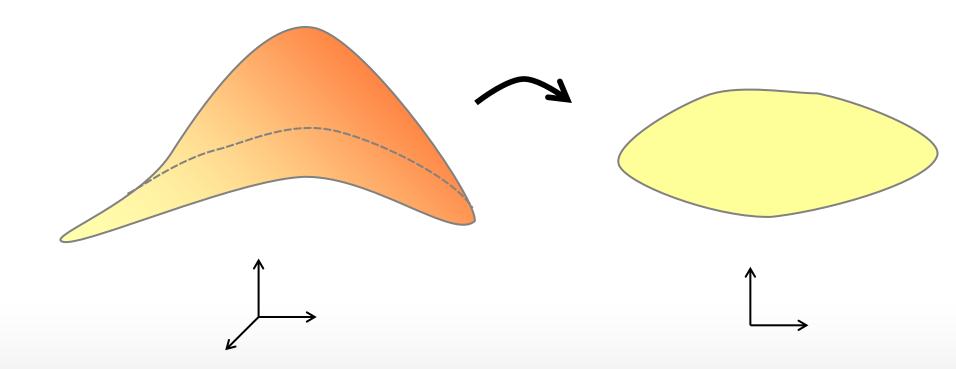
Surface Parameterization



Surface Parameterization

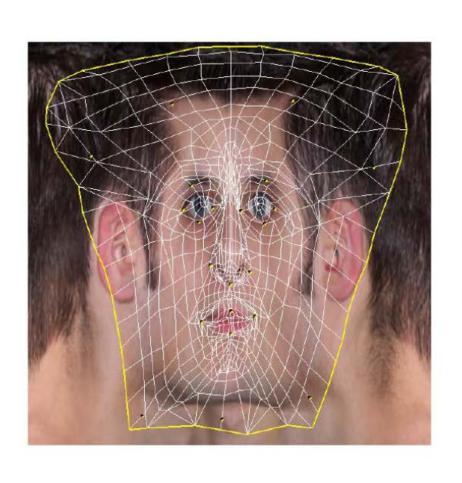
Problem

Given a surface (mesh), construct a (good, useful)
 bijective map from the surface to the plane



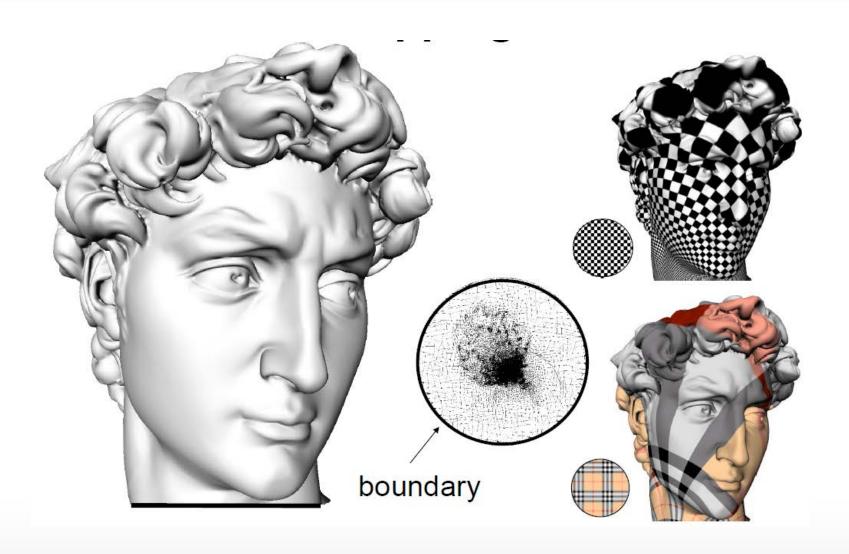
Motivation

Texture mapping





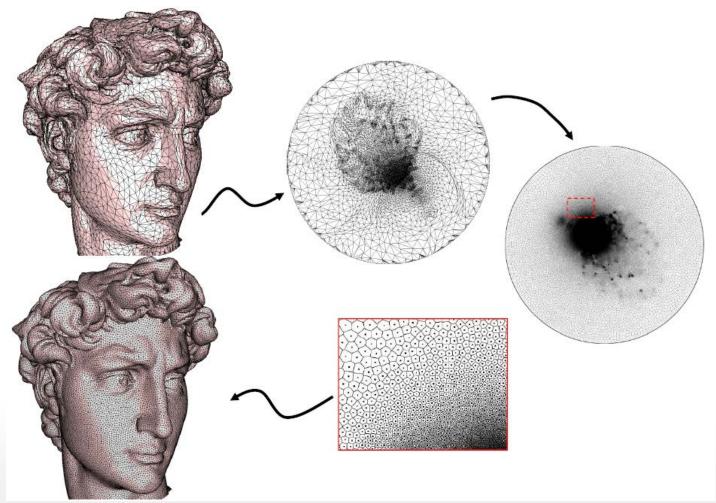
Motivation



Motivation

Operations in 2D are often simpler

• Example: Remeshing ("good" parametrization is essential)



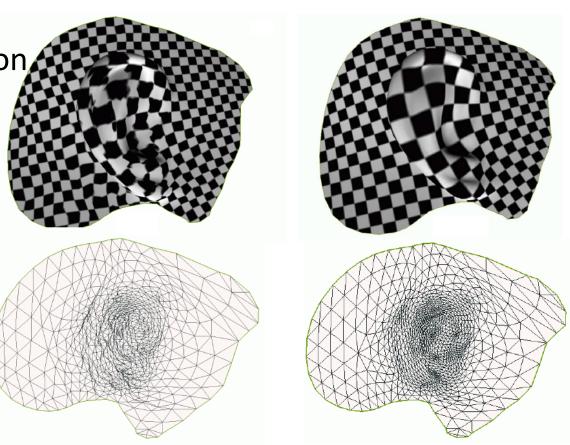
Parametrization

Desirable properties

Low distorsion

Bijective mapping

Efficient computation



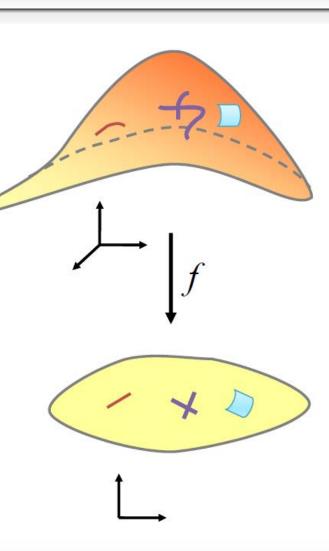
Parametrization

Setting

• Consider a map $f: M \mapsto \Omega \subset \mathbb{R}^2$

Properties of the map

• The map is *isometric* if the length of every curve c in M is the same as the length of the image f(c) in Ω



Isometries

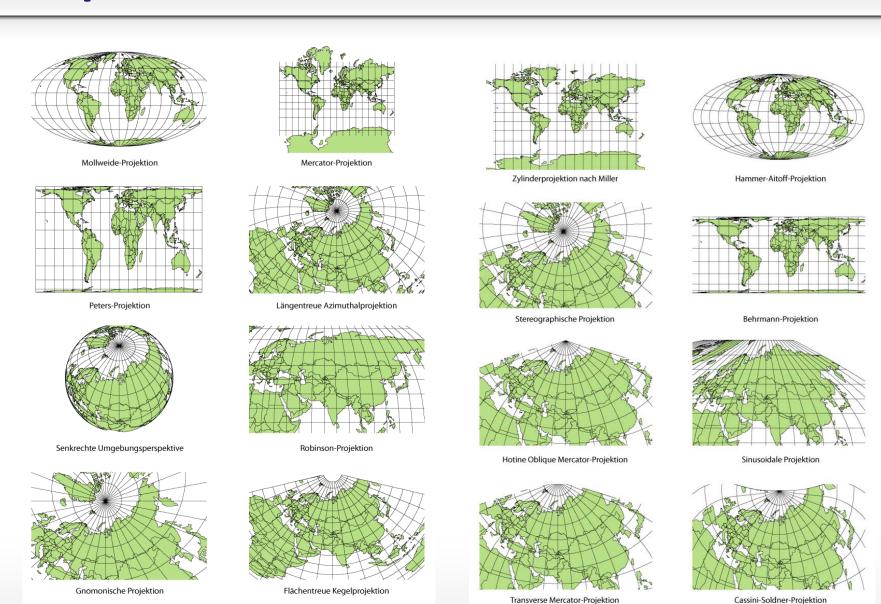
Is there always an isometry?







Maps of the Earth



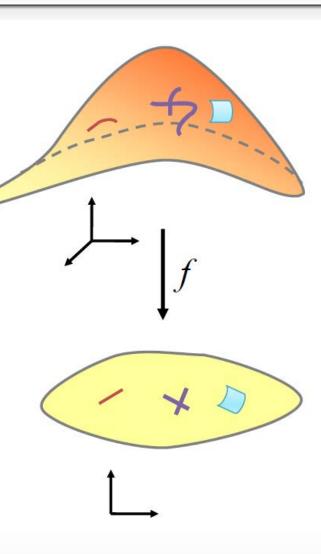
Parametrization

Setting

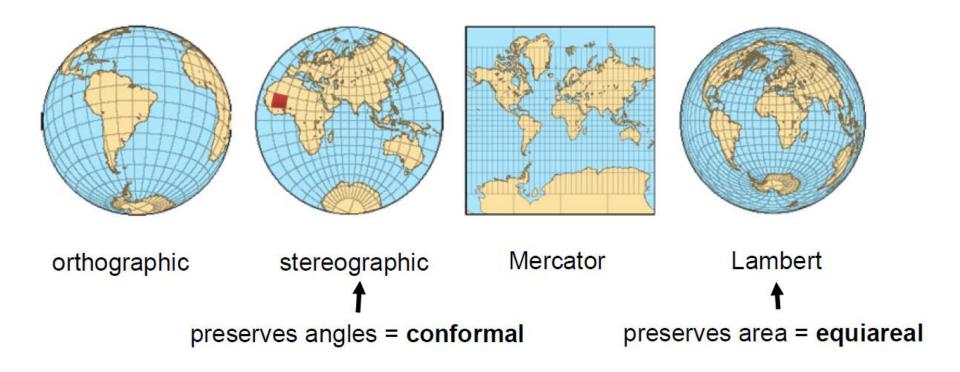
• Consider a map $f: M \mapsto \Omega \subset \mathbb{R}^2$

Properties of the map

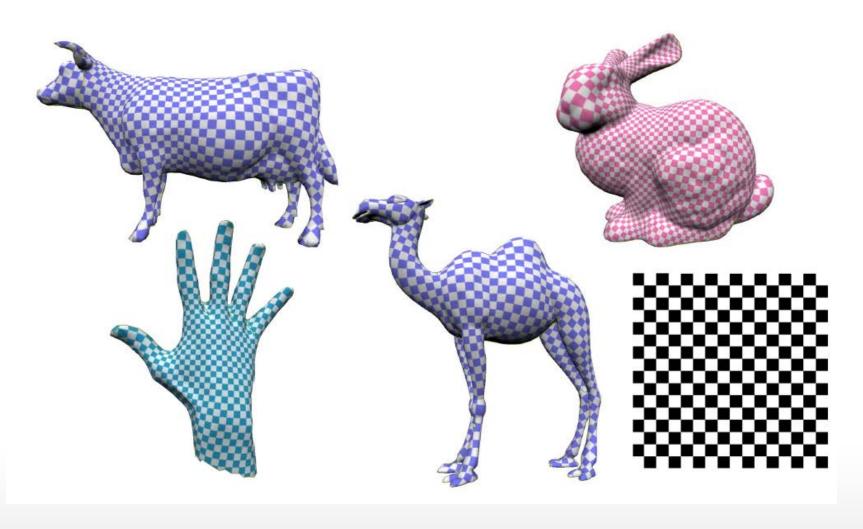
- The map is *isometric* if the length of every curve c in M is the same as the length of the image f(c) in Ω
- The map is conformal if the angles between any pair of curves are preserved
- The map is equiareal if the area of every domain in M is the same as the area of its image



Maps of the Earth



Conformal maps



Characterization of the Properties

Notation

• $f(p) = \binom{u(p)}{v(p)}$, N(p) is the surface normal at $p \in M$

Properties

- f is an isometry if
 - $\langle \nabla u(p), \nabla v(p) \rangle = 0$ and $\|\nabla u(p)\| = \|\nabla v(p)\| = 1$ holds for all p
 - This means the gradients $\nabla u(p)$, $\nabla v(p)$ are orthonormal at all p
 - There is an orthogonal (e.g. rotation) that maps the standard basis in \mathbb{R}^2 to $\nabla u(p)$, $\nabla v(p)$

Characterization of the Properties

Properties

- *f* is conformal if
 - $\nabla u(p) + N(p) \times \nabla v(p) = 0$ holds for all p
 - Then, $\langle \nabla u(p), \nabla v(p) \rangle = 0$ and $\|\nabla u(p)\| = \|\nabla v(p)\|$
 - There is a map that combines a rotation and scaling and maps the standard basis in \mathbb{R}^2 to $\nabla u(p)$, $\nabla v(p)$
 - Remark: f is anti-conformal if $\nabla u(p) N(p) \times \nabla v(p) = 0$ holds for all p

Characterization of the Properties

Properties

- *f* is conformal if
 - $\nabla u(p) + N(p) \times \nabla v(p) = 0$ holds for all p
 - Then, $\langle \nabla u(p), \nabla v(p) \rangle = 0$ and $\|\nabla u(p)\| = \|\nabla v(p)\|$
 - There is a map that combines a rotation and scaling and maps the standard basis in \mathbb{R}^2 to $\nabla u(p), \nabla v(p)$
 - Remark: f is anti-conformal if $\nabla u(p) N(p) \times \nabla v(p) = 0$ holds for all p
- *f* is equiareal if
 - $\|\nabla u(p) \times \nabla v(p)\| = 1$
 - This means that the parallelogram formed by the gradients has unit area.

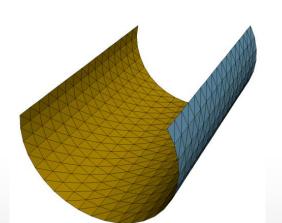
Isometries of domains in the plane

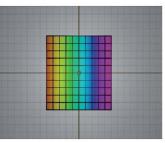
Rotations, reflections, translation (and concatenations of these)

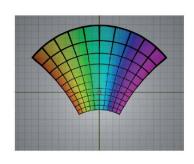
Conformal maps of domains in the plane

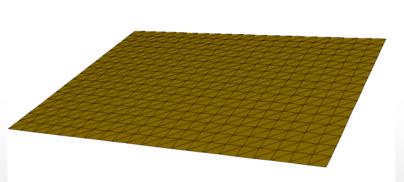
Möbius transformations

Example: Isometry









Isometry = Conformal + Equiareal

Theorem

• The map f is an isometry if and only if it is conformal (or anti-conformal) and equiareal

Remarks

- Isometries are ideal, but exist only for special surfaces.
- There are always conformal and equiareal maps, however, they are not linear polynomials.
- In practice we construct approximations of conformal/ equiareal maps or mixtures

Approximating Conformal Maps

Least Squares Conformal Maps (LSCM)

- Idea: Penalize deviation from conformality in a least squares sense while fixing the maps on the boundary
- Minimize

$$E_{LSCM}(\mathbf{u}, \mathbf{v}) = \frac{1}{2} \int_{M} \|\nabla u(p) + N(p) \times \nabla v(p)\|^{2} dA$$

over all pairs of maps $u, v: M \mapsto \mathbb{R}$ agree with prescribed functions on the boundary of M

• Discretization use linear polynomials as before

Dirichlet Energy and LSCM

Use Dirichlet energy instead of LSCM

We have

$$E_{LSCM}(\mathbf{u}, \mathbf{v}) = E_D(\mathbf{u}, \mathbf{v}) + \text{Area}(\mathbf{u}, \mathbf{v})$$

• For fixed boundary, Area(u, v) does not change when u and v are varied. Then, E_{LSCM} and E_D have the same minimizer

Discrete Problem

Computation

- Energy: (Remember: $f = \begin{pmatrix} u \\ v \end{pmatrix}$) $E_D(f) = f^T S f$
- Constraints:

$$Af = a$$

• Use Lagrange multipliers λ :

$$\begin{bmatrix} S & A^T \\ A & 0 \end{bmatrix} \begin{bmatrix} f \\ \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ a \end{bmatrix}$$

- Boundary conditions can be modified interactively
- Other types of boundary conditions are possible

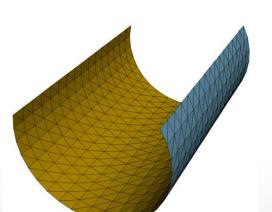
Isometries of domains in the plane

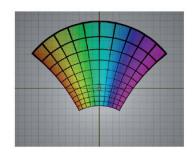
Affine transformations (rotation + translation)

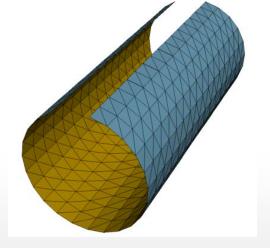
Conformal maps of domains in the plane

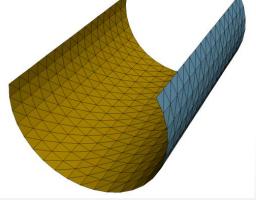
Möbius transformations

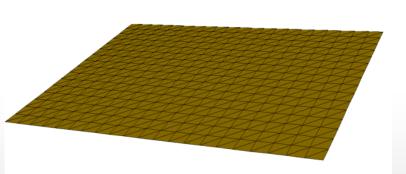
Example: Isometry



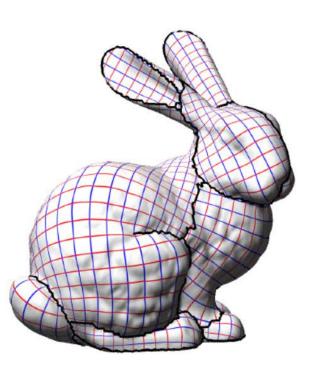


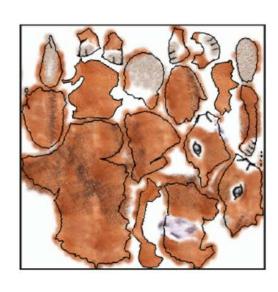






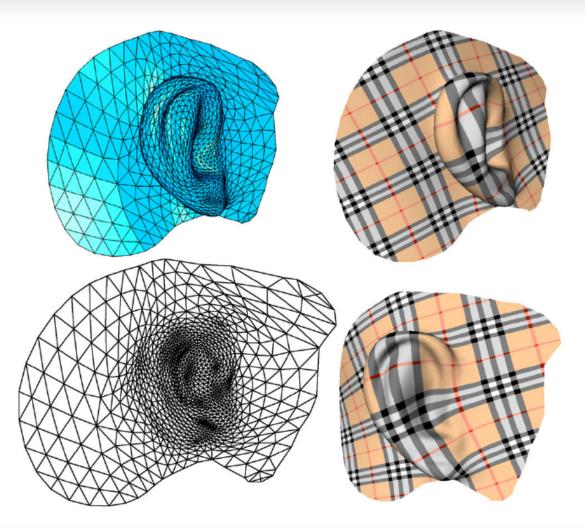
Texture mapping







Lévy, Petitjean, Ray, and Maillot: Least squares conformal maps for automatic texture atlas generation, SIGGRAPH 2002



Intrinsic parameterizations of surface meshes M Desbrun, M Meyer, P Alliez Computer Graphics Forum 21 (3), 209-218

Limitations

Boundary conditions

- Quality of results depends strongly on choice of boundary conditions. How to find good boundary conditions?
- Some approaches:
 - Neumann boundary conditions [Desbrun et al. 2002]
 - Spectral conformal maps [Mullen et al. 2008]
- Nonlinear methods
 - Circle pattern [Kharevych et al. 2005]
 - Angle-based flattening [Sheffer et al. 2005]
 - Ricci Flow [Springborn et al. 2008]

Global Conformal Maps

Seamless parametrizations of surfaces with genus>0

Global conformal maps [Gu and Yau 2003]

