

# Surface Parameterization

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# Outline

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- Motivation
- Objectives and Discrete Mappings
- Angle Preservation
  - Discrete Harmonic Maps
  - Discrete Conformal Maps
  - Angle Based Flattening
- Reducing Area Distortion
- Alternative Domains

# Surface Parameterization

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Mercator-Projektion



Mollweide-Projektion

[[www.wikipedia.de](http://www.wikipedia.de)]

# Surface Parameterization



Mollweide-Projektion



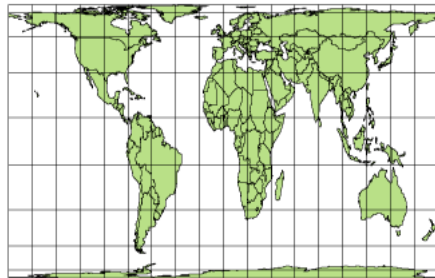
Mercator-Projektion



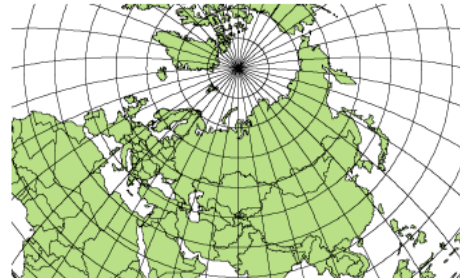
Zylinderprojektion nach Miller



Hammer-Aitoff-Projektion



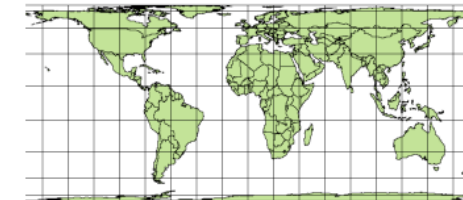
Peters-Projektion



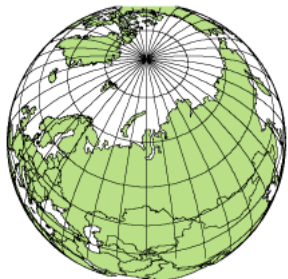
Längentreue Azimuthalprojektion



Stereographische Projektion



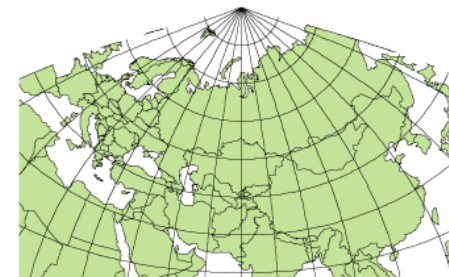
Behrmann-Projektion



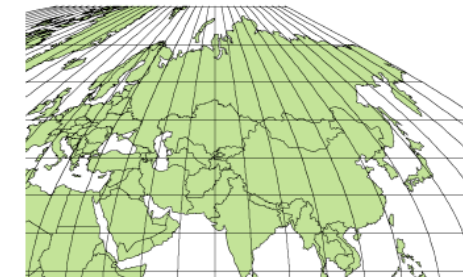
Senkrechte Umgebungsperspektive



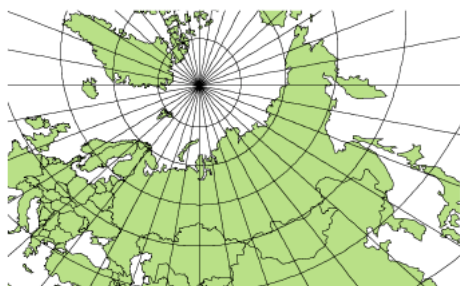
Robinson-Projektion



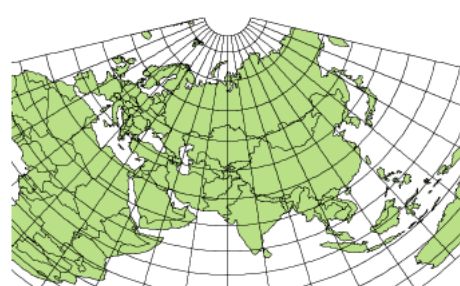
Hotine Oblique Mercator-Projektion



Sinusoidale Projektion



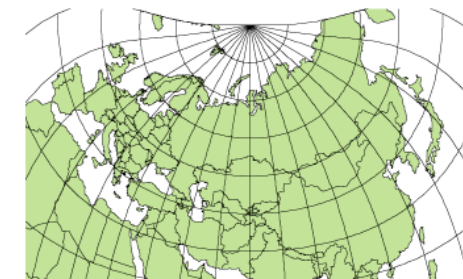
Gnomonische Projektion



Flächentreue Kegelprojektion



Transverse Mercator-Projektion

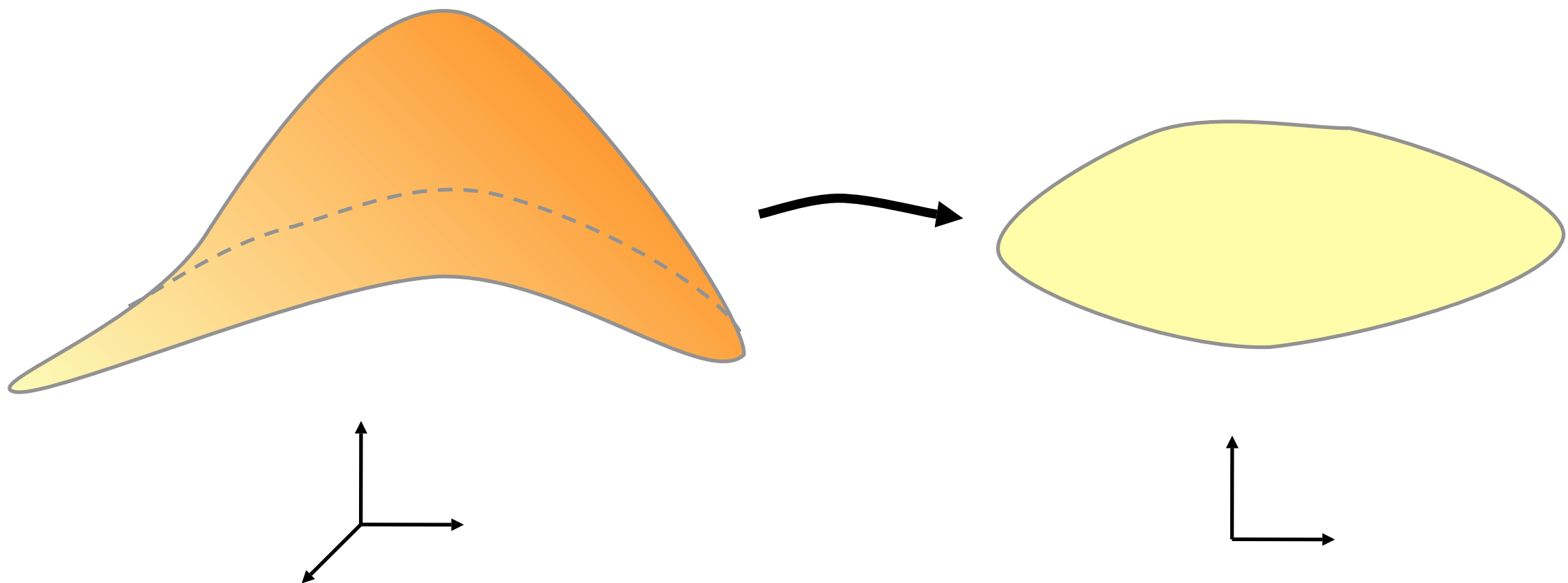


Cassini-Soldner-Projektion

[www.wikipedia.de]

# Surface Parameterization

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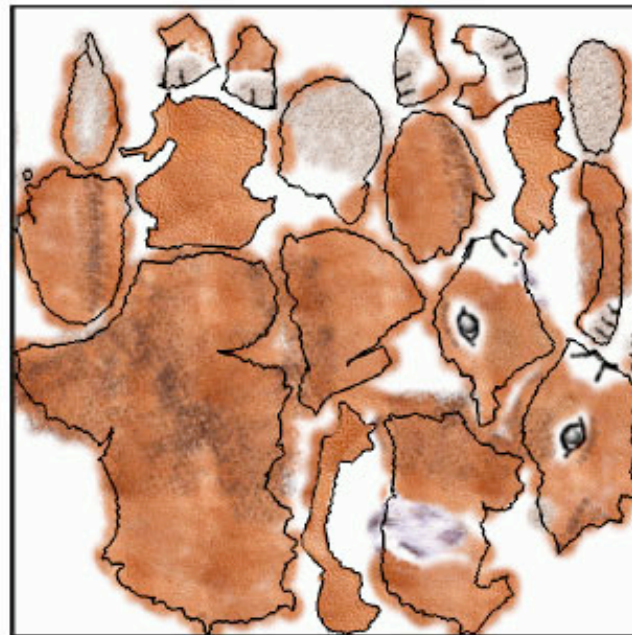
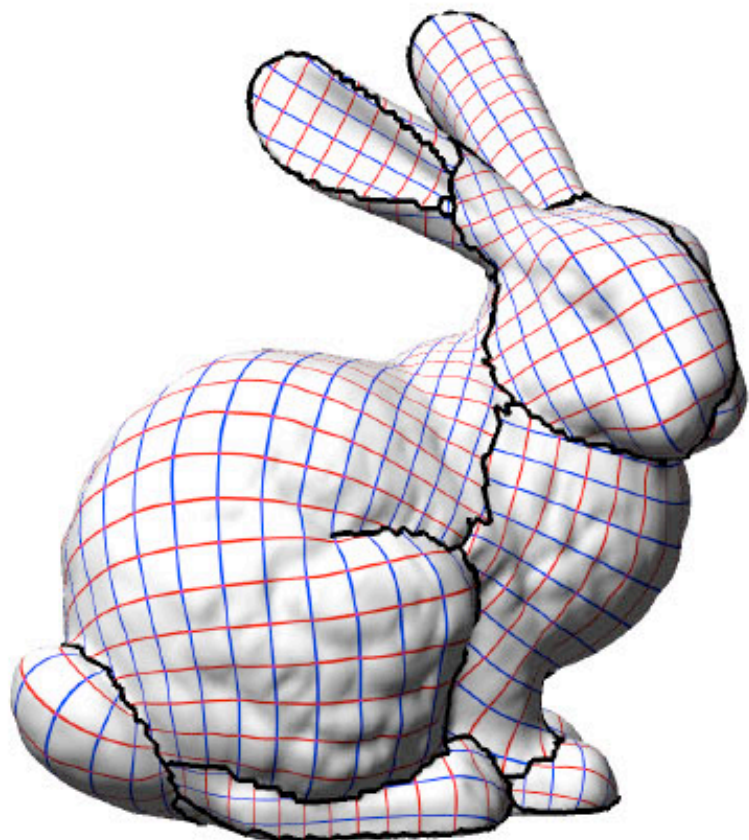




# Motivation

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- Texture mapping

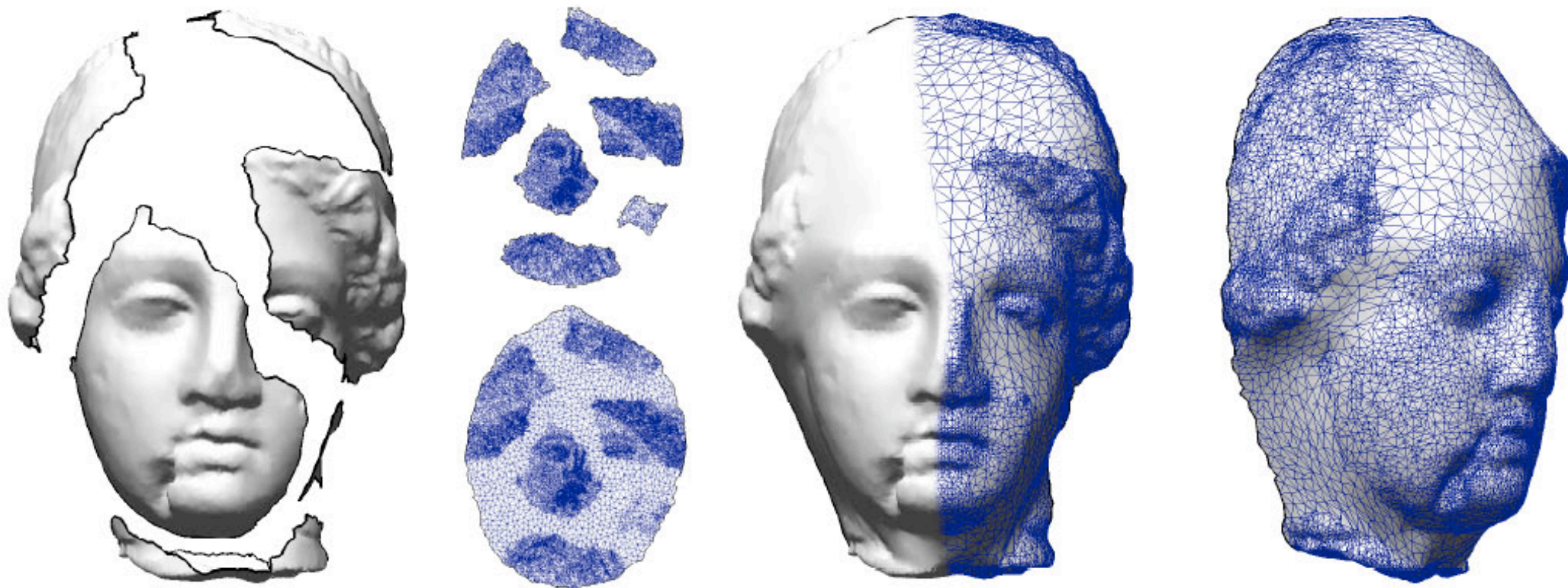


Lévy, Petitjean, Ray, and Maillot: *Least squares conformal maps for automatic texture atlas generation*, SIGGRAPH 2002

# Motivation

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- Many operations are simpler on planar domain

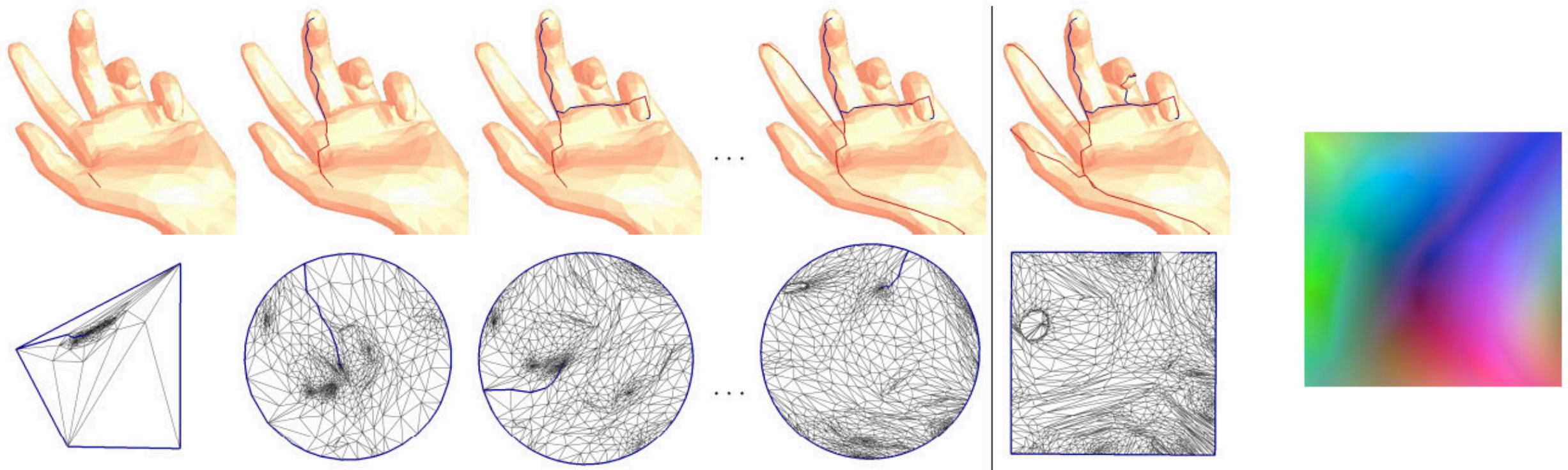


Lévy: *Dual Domain Extrapolation*, SIGGRAPH 2003



# Motivation

- Exploit regular structure in domain

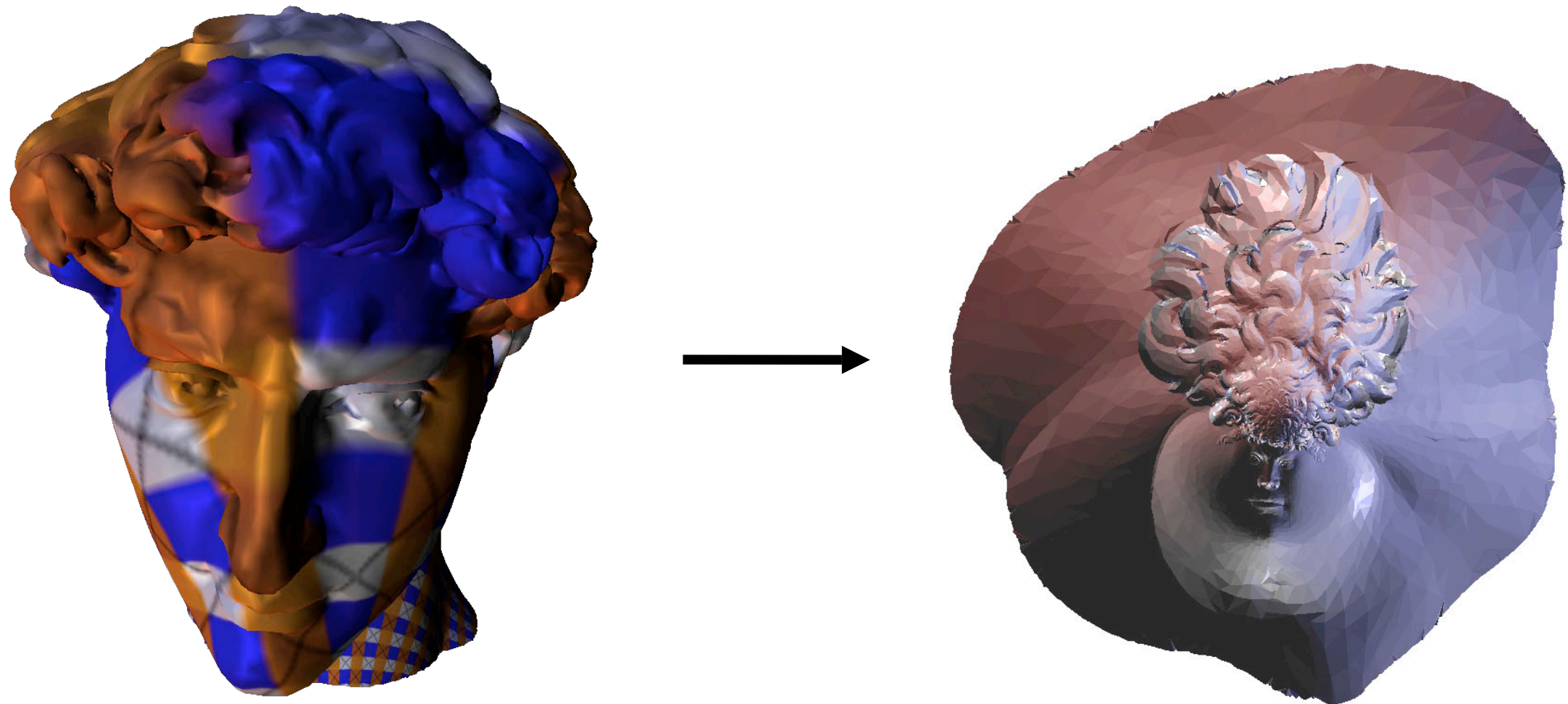


Gu, Gortler, Hoppe: *Geometry Images*, SIGGRAPH 2002

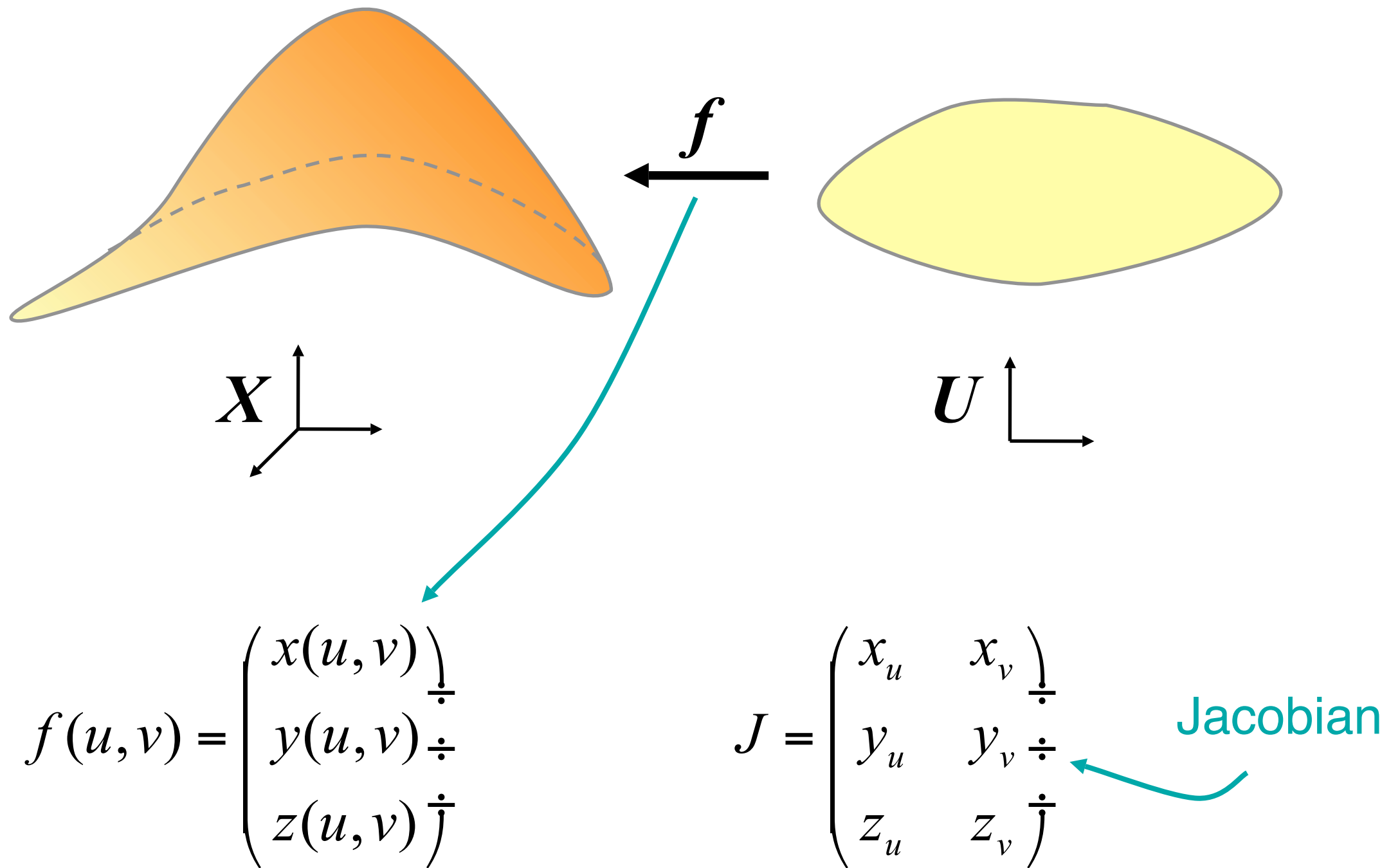


# Surface Parameterization

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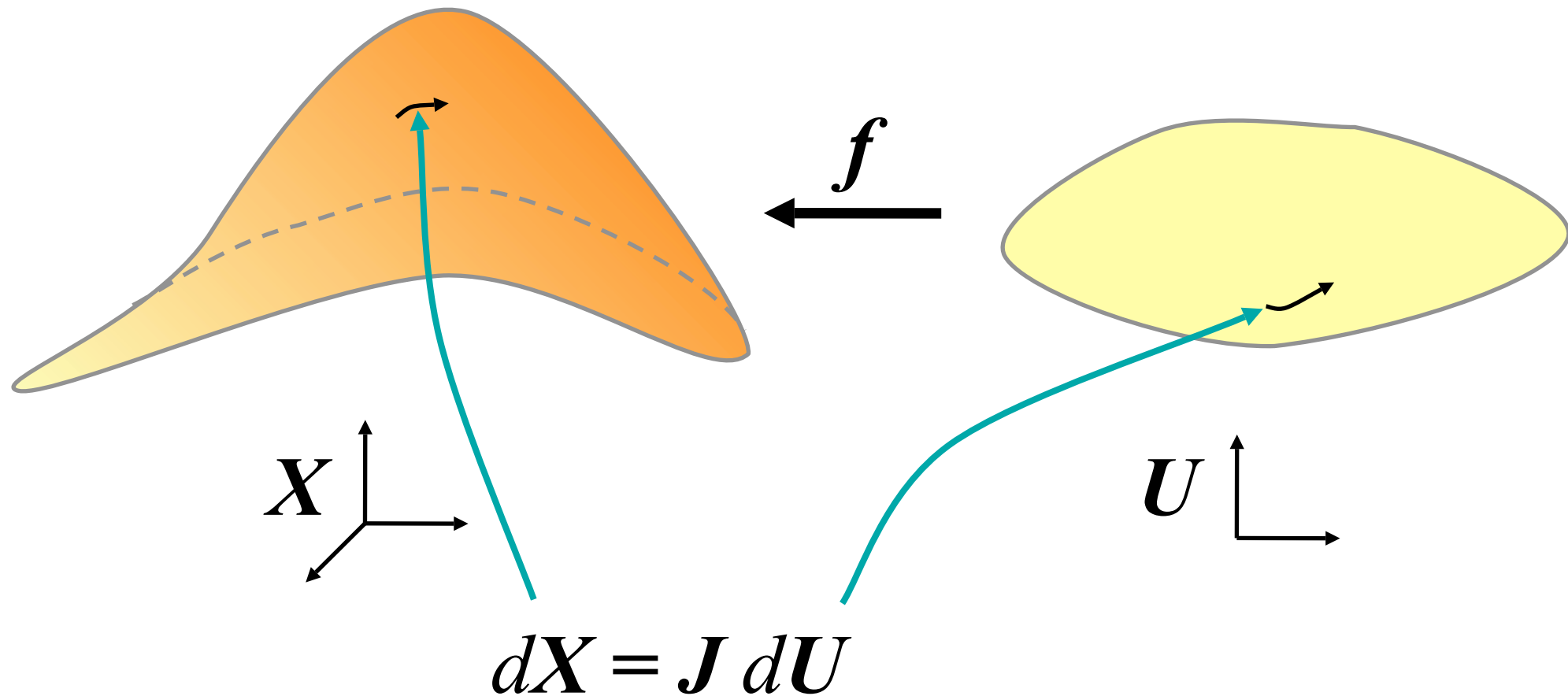


# Surface Parameterization



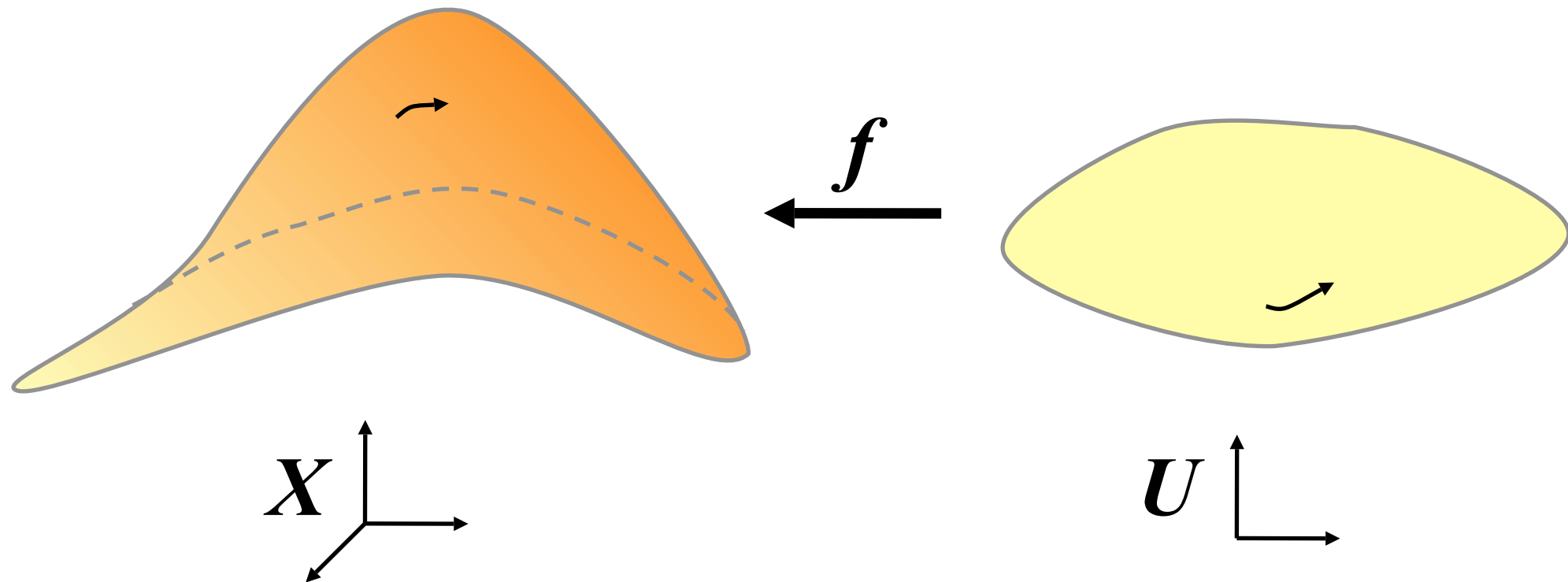
# Surface Parameterization

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# Surface Parameterization



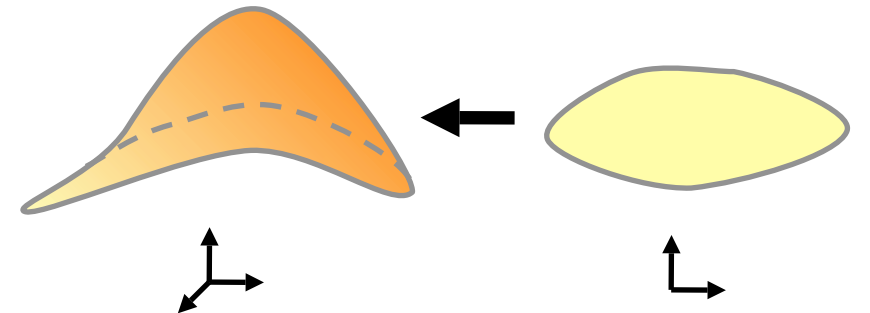
$$dX = J dU$$

$$\|dX\|^2 = dU \underbrace{J^T J}_{\text{First Fundamental Form}} dU$$

$$\mathbf{I} = \begin{pmatrix} x_u x_u & x_u x_v \\ x_u x_v & x_v x_v \end{pmatrix}$$

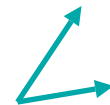
# Characterization of Mappings

- By first fundamental form  $I$ 
  - Eigenvalues  $\lambda_{1,2}$  of  $I$
  - Singular values  $\sigma_{1,2}$  of  $J$  ( $\sigma_i^2 = \lambda_i$ )



- *Isometric*

- $I = Id$ ,  $\lambda_1 = \lambda_2 = 1$



- *Conformal*

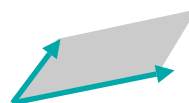
- $I = \mu Id$ ,  $\lambda_1 / \lambda_2 = 1$



angle preserving

- *Equiareal*

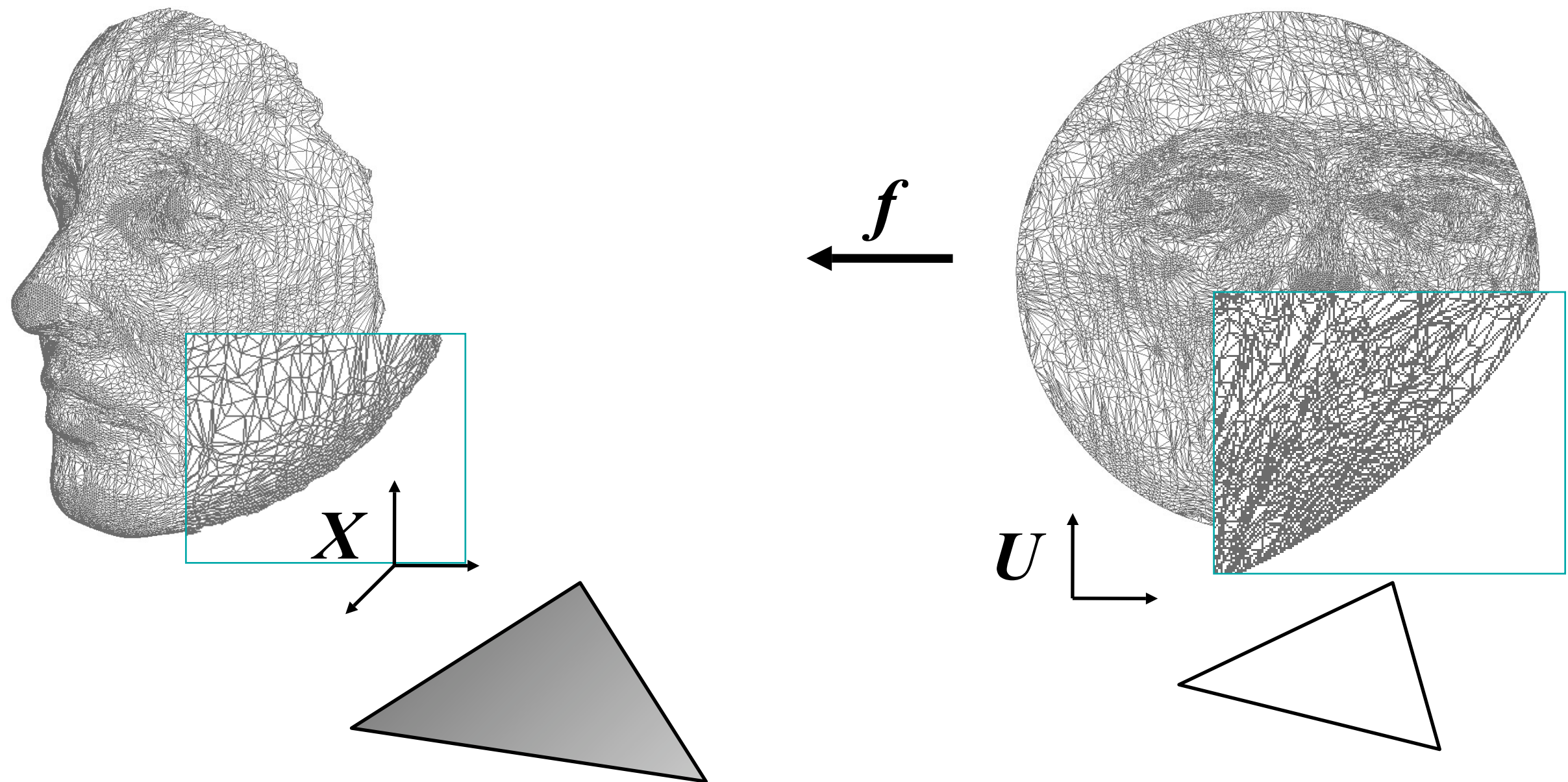
- $\det I = 1$ ,  $\lambda_1 \lambda_2 = 1$



area preserving

# Piecewise Linear Maps

- Mapping = 2D mesh with same connectivity

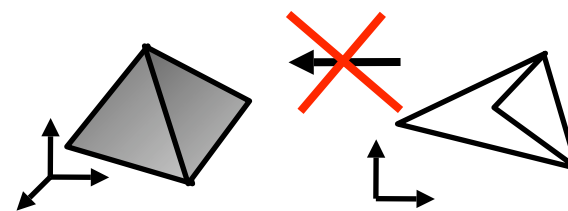




# Objectives

- Isometric maps are rare
- Minimize distortion w.r.t. a certain measure

- Validity (bijective map)



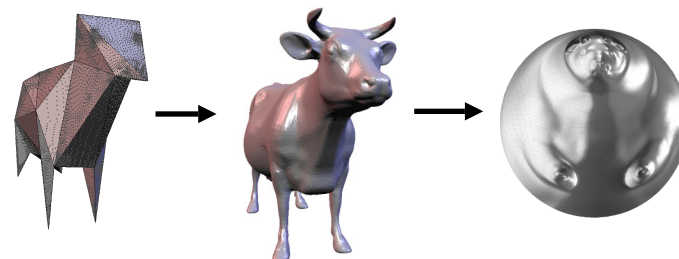
triangle flip

- Boundary



fixed / free?

- Domain



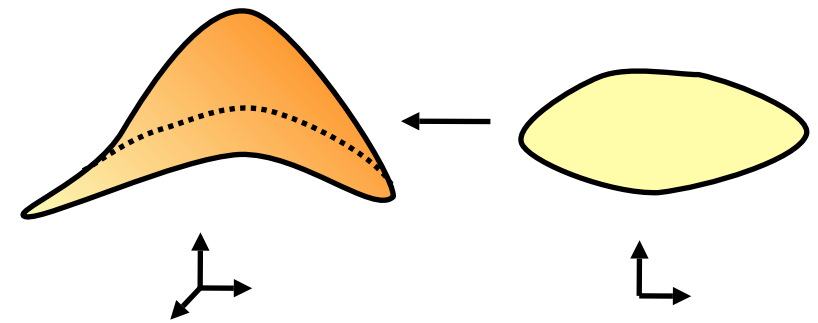
e.g., spherical

- Numerical solution

linear / non-linear?

# Discrete Harmonic Maps

- $f$  is *harmonic* if  $\Delta f = 0$
- Solve Laplace equation



$$\begin{cases} \Delta u = 0 \\ \Delta v = 0 \\ (u, v)|_{\partial\Omega} = (u_0, v_0) \end{cases}$$

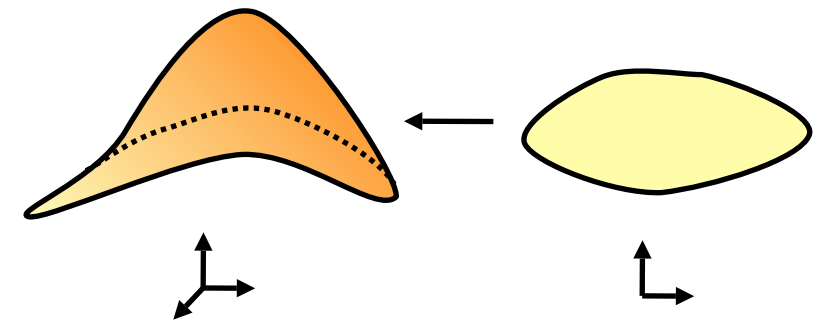
$u$  and  $v$  are *harmonic*

Dirichlet boundary conditions

- In 3D: "fix planar boundary and smooth"

# Discrete Harmonic Maps

- $f$  is *harmonic* if  $\Delta f = 0$
- Solve Laplace equation
- Yields linear system (again)



$$L(p_i) = \sum_{j \in N_i} w_{ij} (p_j - p_i) = 0 \quad \text{vertices } 1 \leq i \leq n$$

- *Convex combination maps*
  - *Normalization*
  - *Positivity*

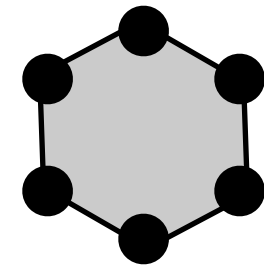
$$\sum_{j \in N_i} w_{ij} = 1$$
$$w_{ij} > 0$$



# Convex Combination Maps

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- Every (interior) planar vertex is a *convex combination* of its neighbors
- Guarantees *validity* if boundary is mapped to a convex polygon (e.g., rectangle, circle)
- Weights
  - Uniform (*barycentric mapping*)
  - Shape preserving [Floater 1997]
  - Mean Value Coordinates [Floater 2003]
    - Use mean value property of harmonic functions



Reproduction of  
planar meshes

# Conformal Maps

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- Planar *conformal mappings*  $f(x, y) = \begin{pmatrix} u(x, y) \\ v(x, y) \end{pmatrix}$  satisfy the *Cauchy-Riemann conditions*

$$\frac{\partial u(x, y)}{\partial x} = \frac{\partial v(x, y)}{\partial y} \quad \text{and} \quad \frac{\partial u(x, y)}{\partial y} = -\frac{\partial v(x, y)}{\partial x}$$

# Conformal Maps

---

- Planar *conformal mappings*  $f(x, y) = \begin{pmatrix} u(x, y) \\ v(x, y) \end{pmatrix}$  satisfy the *Cauchy-Riemann conditions*

$$u_x = v_y \quad \text{and} \quad u_y = -v_x$$

- Differentiating once more by  $x$  and  $y$  yields

$$u_{xx} = v_{xy} \quad \text{and} \quad u_{yy} = -v_{xy} \quad \Rightarrow \quad u_{xx} + u_{yy} = \Delta u = 0$$

and similar  $\Delta v = 0$

- conformal  $\Rightarrow$  harmonic



# Discrete Conformal Maps

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- Planar *conformal mappings*  $f(x, y) = \begin{pmatrix} u(x, y) \\ v(x, y) \end{pmatrix}$  satisfy the *Cauchy-Riemann conditions*

$$u_x = v_y \quad \text{and} \quad u_y = -v_x$$

- *In general, there are no conformal mappings for piecewise linear functions!*

# Discrete Conformal Maps

---

- Planar *conformal mappings*  $f(x, y) = \begin{pmatrix} u(x, y) \\ v(x, y) \end{pmatrix}$  satisfy the *Cauchy-Riemann conditions*

$$u_x = v_y \quad \text{and} \quad u_y = -v_x$$

- Conformal energy (*per triangle*  $T$ )

$$E_T = (u_x - v_y)^2 + (u_y + v_x)^2$$

- Minimize  $\sum_{T \in \mathcal{T}} E_T A_T \rightarrow \min$

# Discrete Conformal Maps

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- Least-squares conformal maps [Lévy et al. 2002]

$$\sum_{T \in \mathcal{T}} E_T A_T \rightarrow \min \quad \text{where} \quad E_T = (u_x - v_y)^2 + (u_y + v_x)^2$$

- Satisfy Cauchy-Riemann conditions in *least-squares* sense
- Leads to solution of linear system
- *Alternative formulation leads to same solution...*

# Discrete Conformal Maps

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- Same solution is obtained for

$$\Delta_S u = 0$$

cotangent weights

$$\Delta_S v = 0$$

$$n \times \nabla u \big|_{\partial\Omega} = c$$

Neumann boundary conditions

$$n \times \nabla v \big|_{\partial\Omega} = c$$

$$(u, v)_{|\partial\Omega_0} = (u_0, v_0)$$

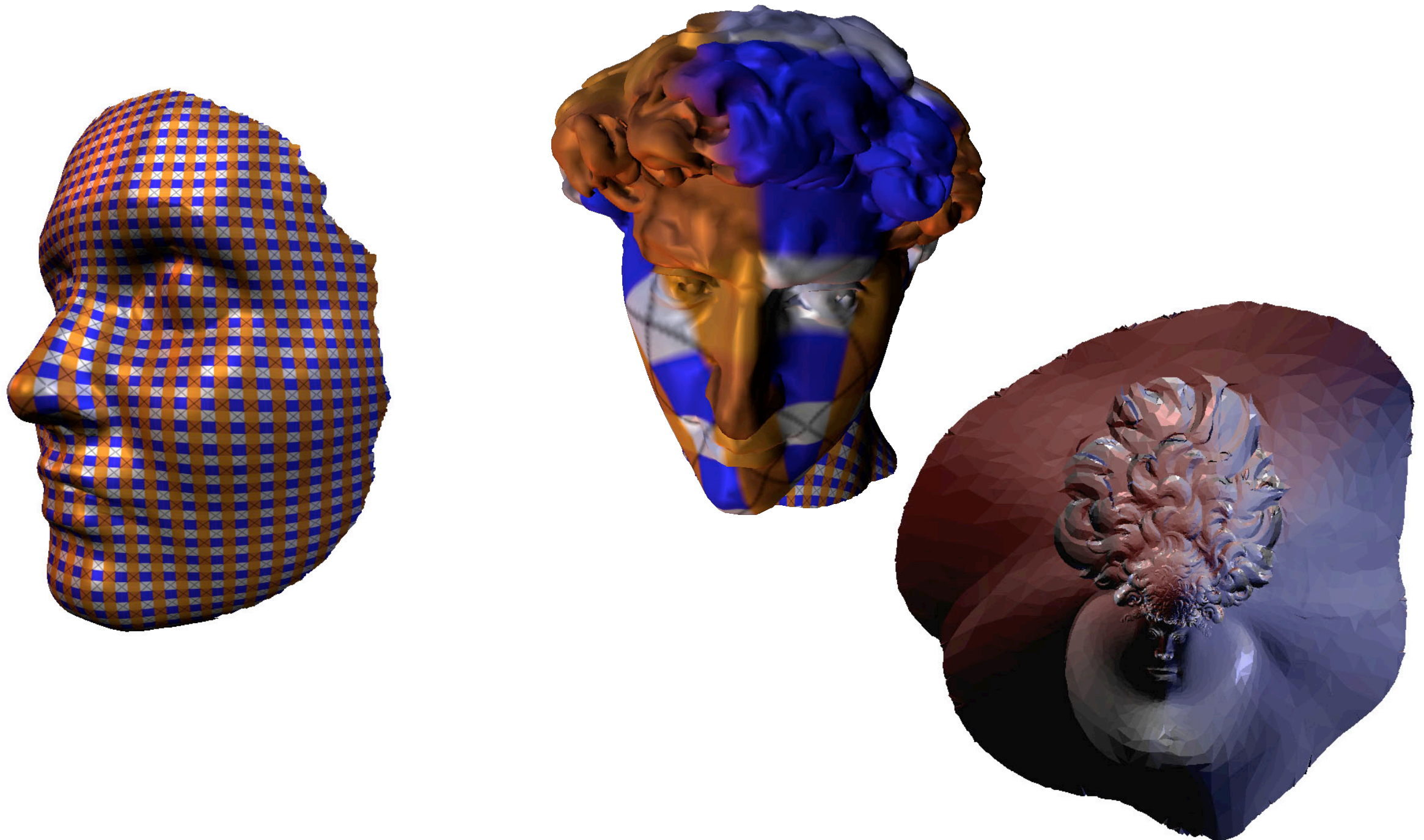
+ fixed vertices

*Discrete Conformal Maps*

[Desbrun et al. 2002]

# Discrete Conformal Maps

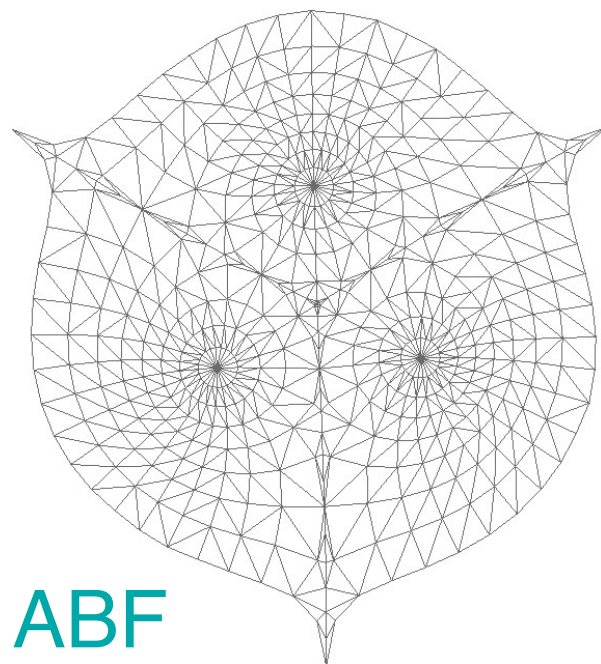
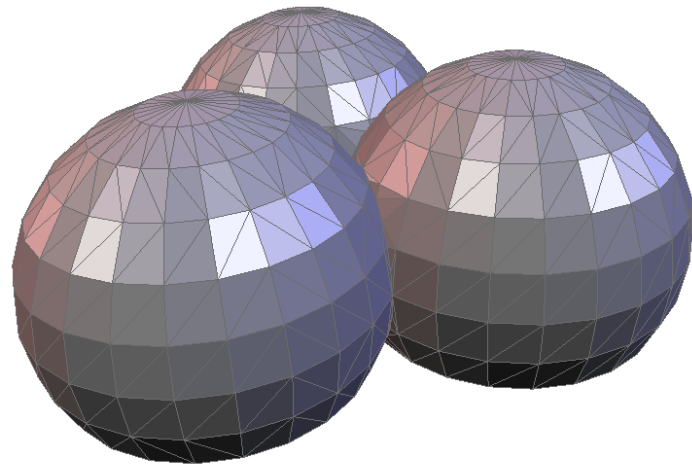
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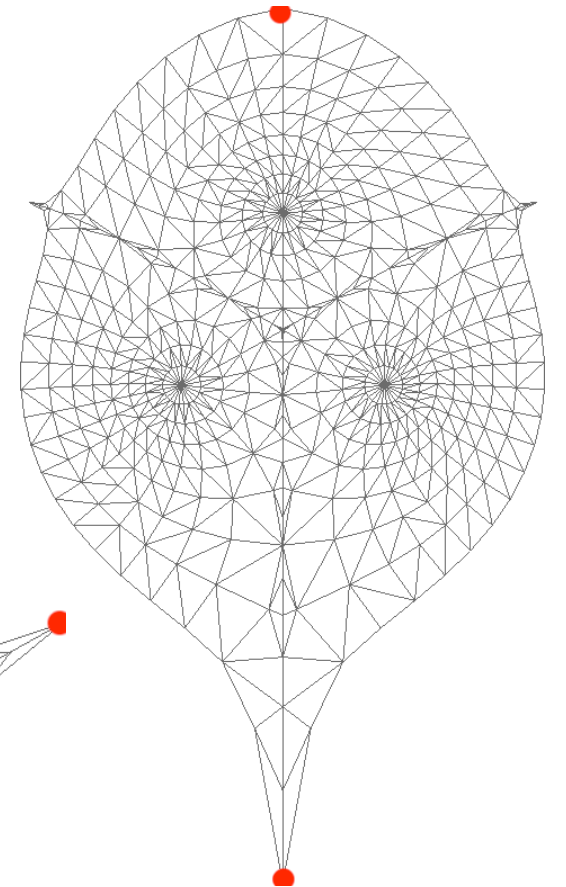
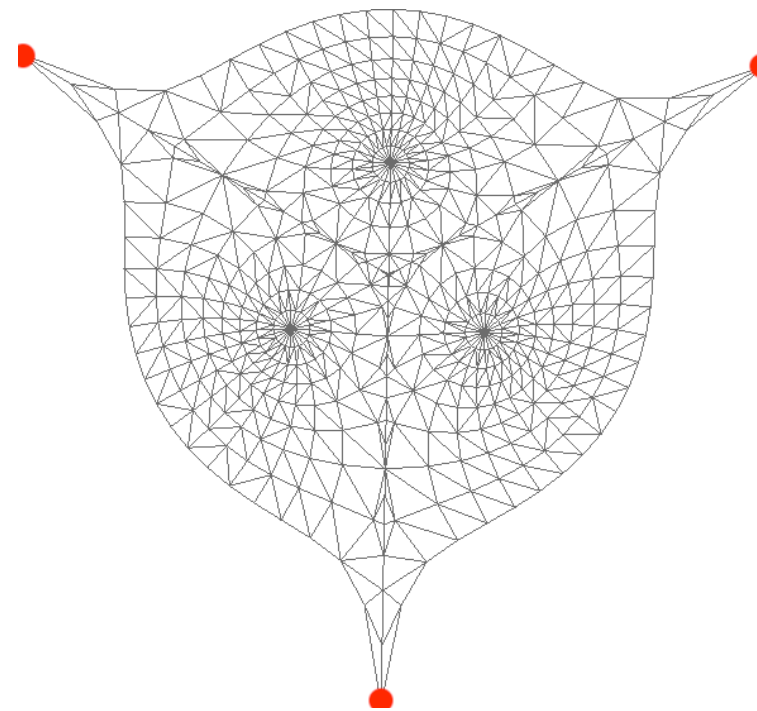
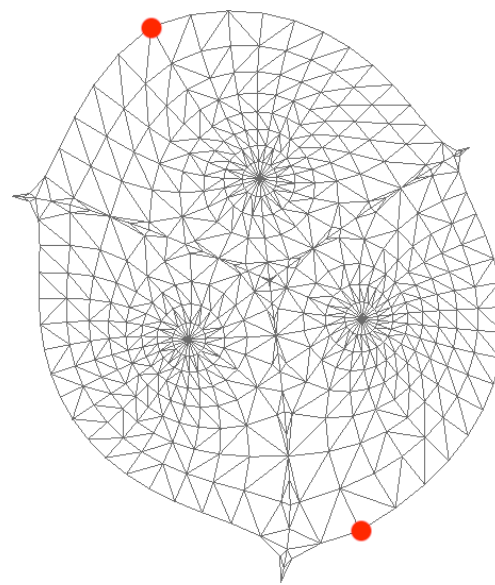


# Discrete Conformal Maps

- Free boundary depends on choice of *fixed* vertices ( $>1$ )



ABF





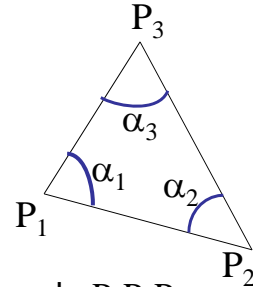
## LSCM – Geometric Interpretation

- Algebraic Interpretation:
  - Minimize conformal energy

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

- Geometric Interpretation:

- Use triangle similarity
- Given angles  $\alpha_1, \alpha_2, \alpha_3$  of a triangle  $P_1P_2P_3$  in 2D we have



$$P_3 - P_1 = \frac{\sin \alpha_2}{\sin \alpha_3} R_{\alpha_1} (P_2 - P_1),$$

$$R_\alpha = \begin{pmatrix} \cos \alpha & \sin \alpha \\ -\sin \alpha & \cos \alpha \end{pmatrix}$$



University of  
British Columbia



## LSCM

- In map from 3D to 2D might be impossible to keep angles exactly
  - Use least-squares

$$\min \sum_i \left( P_3^i - P_1^i - \frac{\sin \alpha_2^i}{\sin \alpha_3^i} R_{\alpha_1^i} (P_2^i - P_1^i) \right)^2$$

- To solve need to fix two vertices
  - Obtain linear system
  - Choice of vertices affects solution
- Can have flips



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# Angle Based Flattening [Sheffer&de Sturler 2000]

- Preserve angles  $\Rightarrow$  specify problem in angles

- Constraints

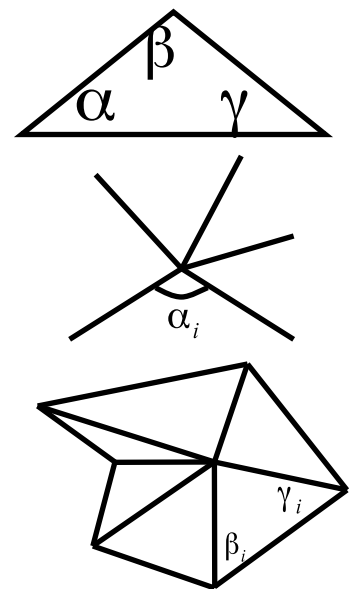
- triangle
- Internal vertex
- Wheel consistency

$$\alpha + \beta + \gamma - \pi = 0$$

$$\sum_i \alpha_i - 2\pi = 0$$

$$\prod_i \sin(\beta_i) - \prod_i \sin(\gamma_i) = 0$$

ensure validity



- Objective function

$$f(x) = \sum_{i=1}^N w_i (\alpha_i - \alpha_i^*)^2$$

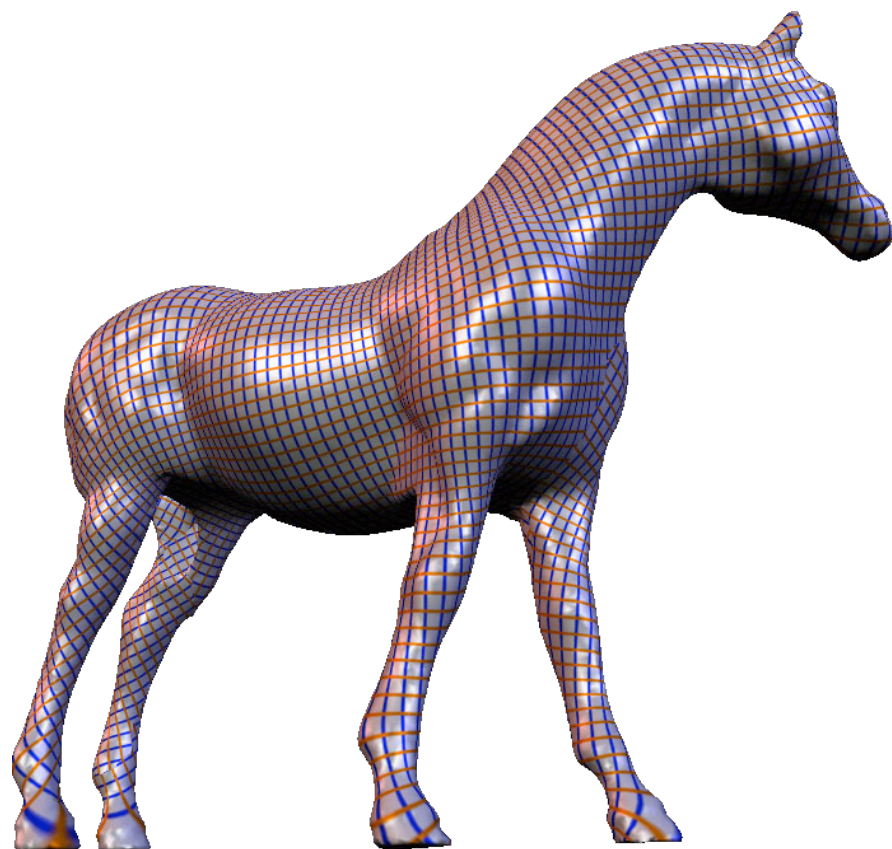
preserve angles 2D ~3D

"optimal" angles (uniform scaling)

# Angle Based Flattening

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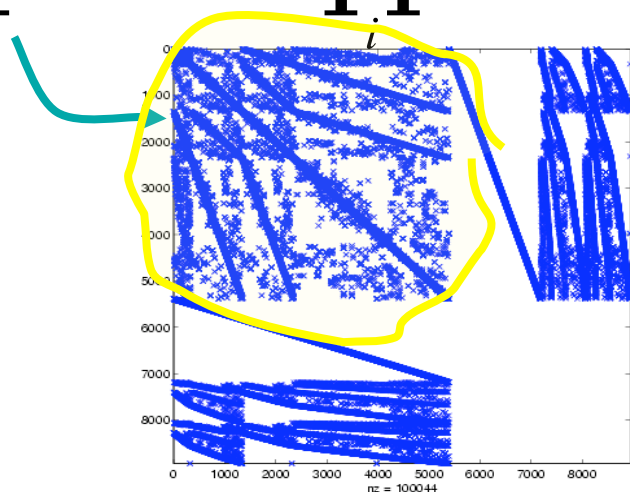
- Free boundary
- Validity: no local self-intersections
- Non-linear optimization



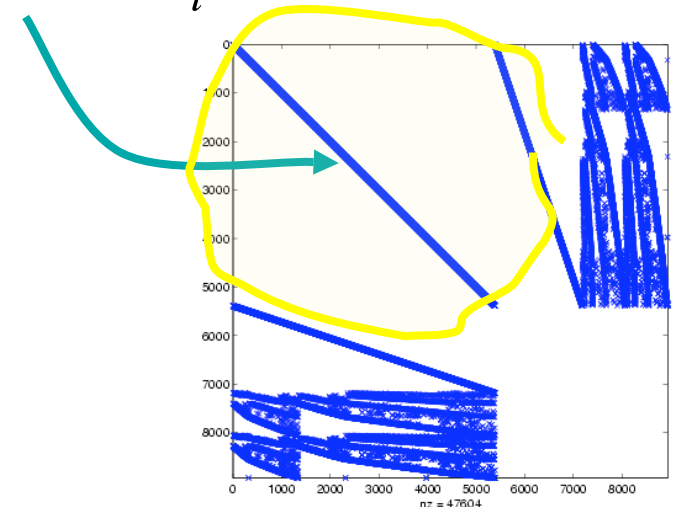
# Angle Based Flattening

- Free boundary
- Non-linear optimization
  - Newton iteration
  - Solve linear system in every step

$$\prod_i \sin(\alpha_i) - \prod_i \sin(\beta_i) = 0$$



$$\prod_i \log \sin(\alpha_i) - \prod_i \log \sin(\beta_i) = 0$$

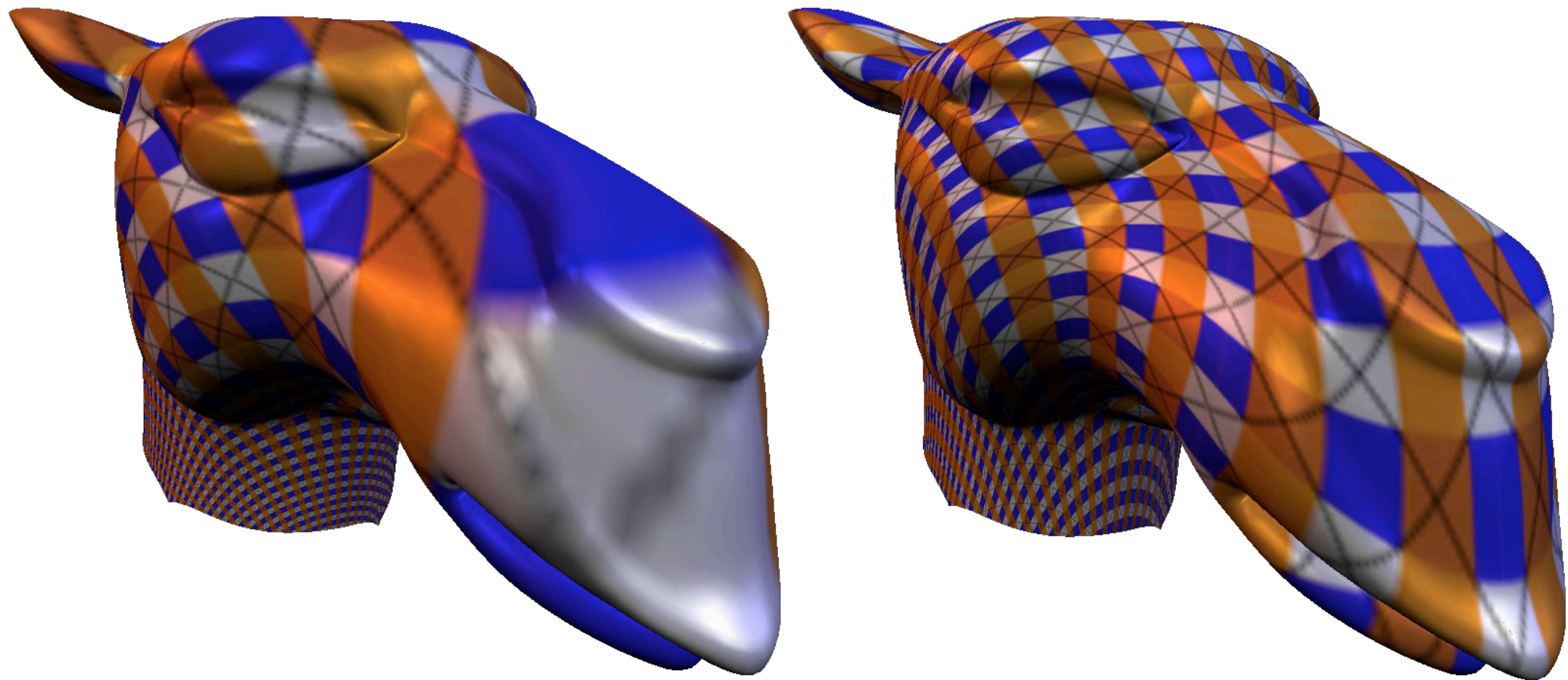


[Zayer et al. 2005]



# And how about area distortion?

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# Reducing Area Distortion

- Energy minimization based on

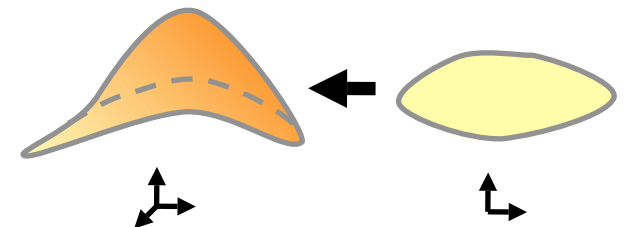
- MIPS [Hormann & Greiner 2000]

- *modification* [Degener et al. 2003]

- "Stretch" [Sander et al. 2001]



- *modification* [Sorkine et al. 2002]



$$\|J\|_F \|J^{-1}\|_F = \frac{\sigma_1}{\sigma_2} + \frac{\sigma_2}{\sigma_1}$$

$$\det J + \frac{1}{\det J} = \sigma_1 \sigma_2 + \frac{1}{\sigma_1 \sigma_2}$$

$$\|J\|_F = \sqrt{\sigma_1 + \sigma_2} \quad \text{or} \quad \|J\|_\infty = \sigma_1$$

$$\max \left\{ \sigma_1, \frac{1}{\sigma_2} \right\}$$

# Non-Linear Methods

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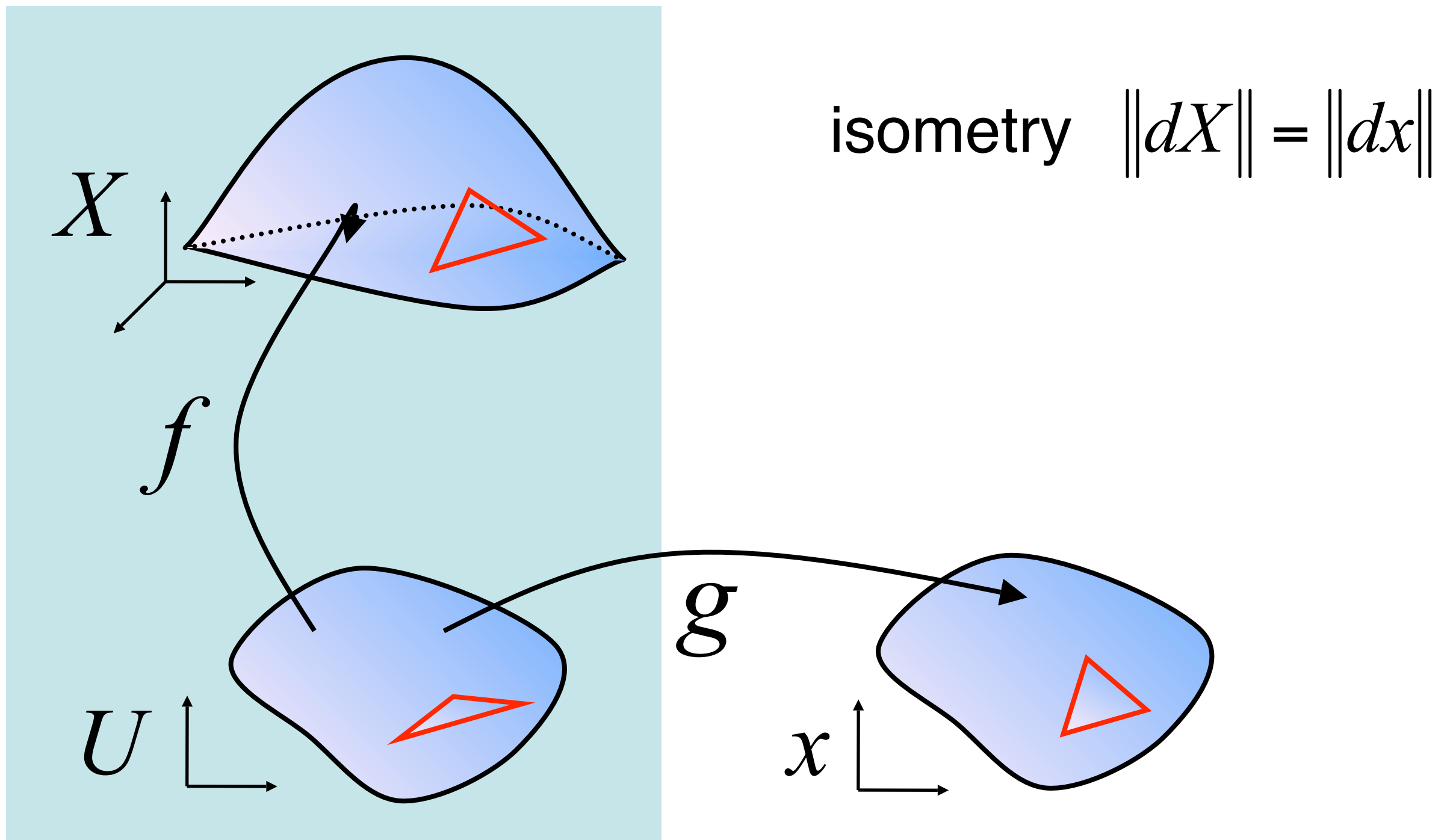
- Free boundary
- Direct control over distortion
- No convergence guarantees
- May get stuck in local minima
- May not be suitable for large problems
- May need feasible point as initial guess
- May require hierarchical optimization even for moderately sized data sets

# Linear Methods

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- Efficient solution of a sparse linear system
- Guaranteed convergence
- Fixed convex boundary
- May suffer from area distortion for complex meshes
- *An alternative approach to reducing area distortion...*
  - *How accurately can we reproduce a surface on the plane?*
  - *How do we characterize the mapping?*

# Reducing Area Distortion





# Reducing Area Distortion

- Quasi-harmonic maps [Zayer et al. 2005]

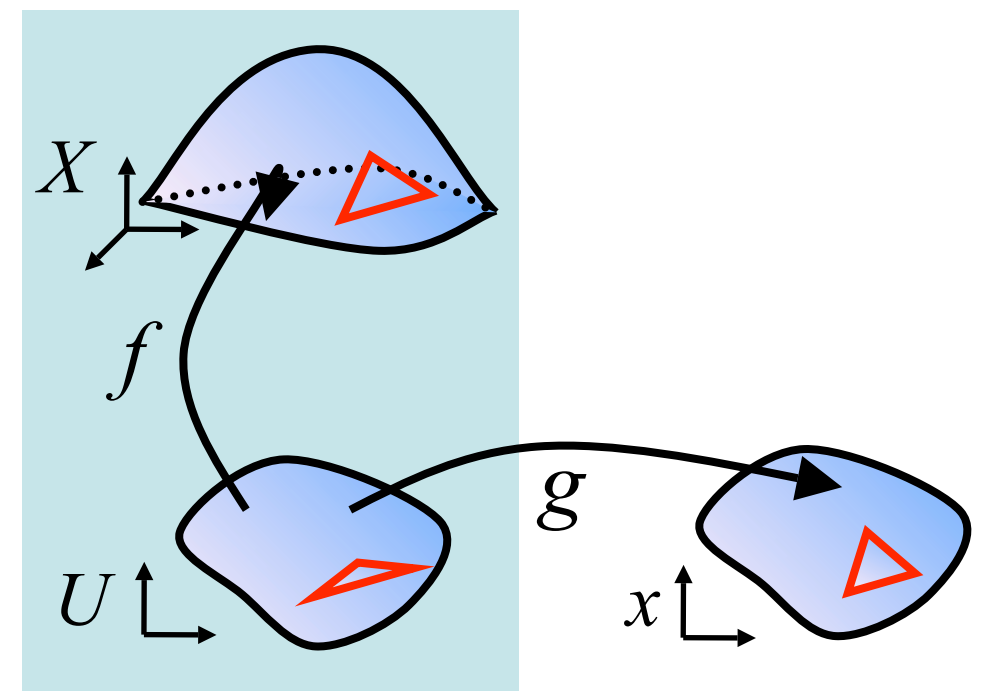


$$\int C \nabla g \times \nabla g \, d\mathbb{R} \min$$

$$\operatorname{div}(C \nabla g) = 0$$

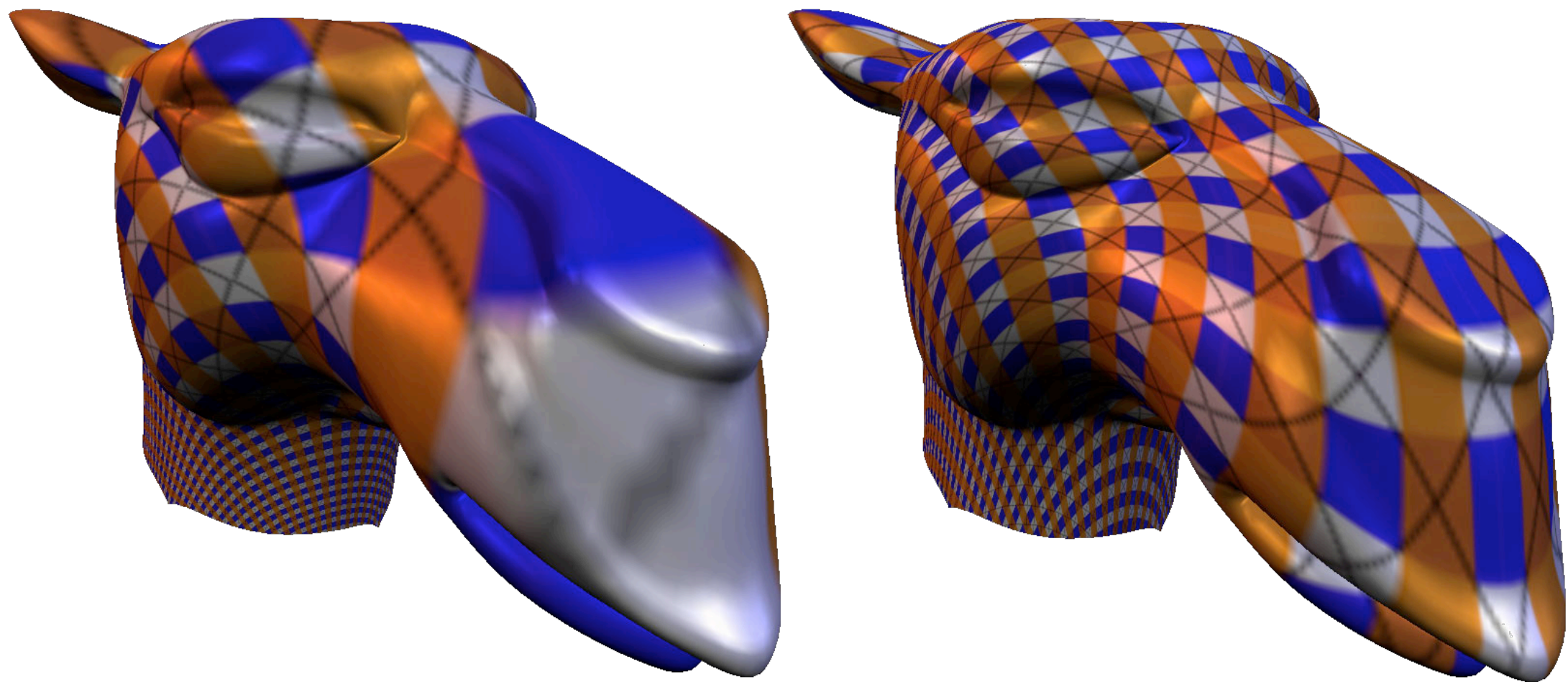
estimate from  $f$

- Iterate (*few iterations*)
  - Determine tensor  $C$  from  $f$
  - Solve for  $g$



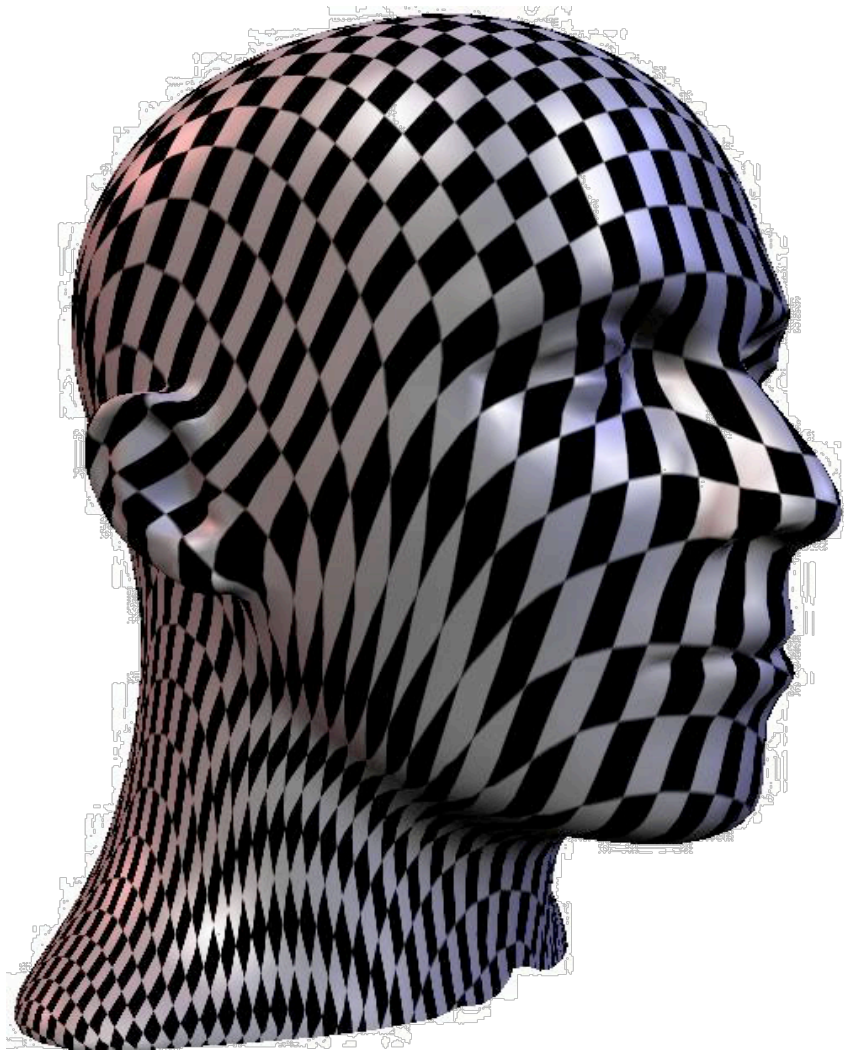
# Examples

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# Examples

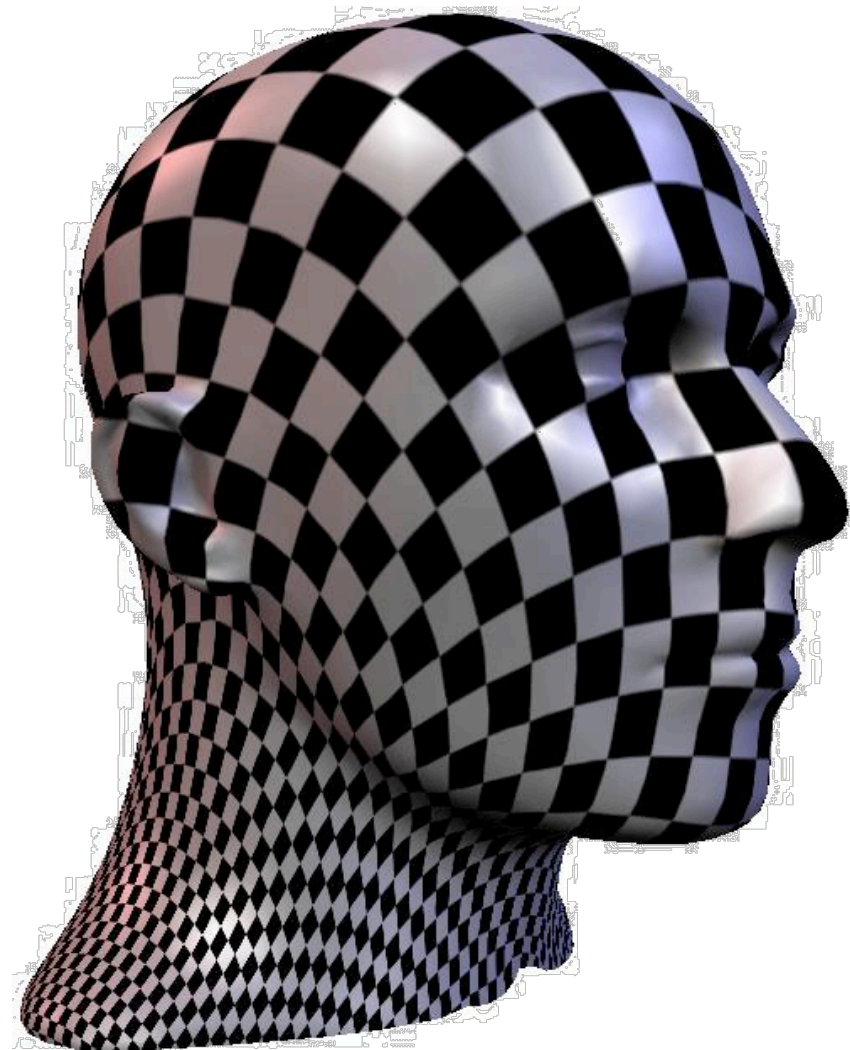
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$$\sqrt{\sigma_1 + \sigma_2} \rightarrow \min$$

Stretch metric minimization

Using [Yoshizawa et. al 2004]

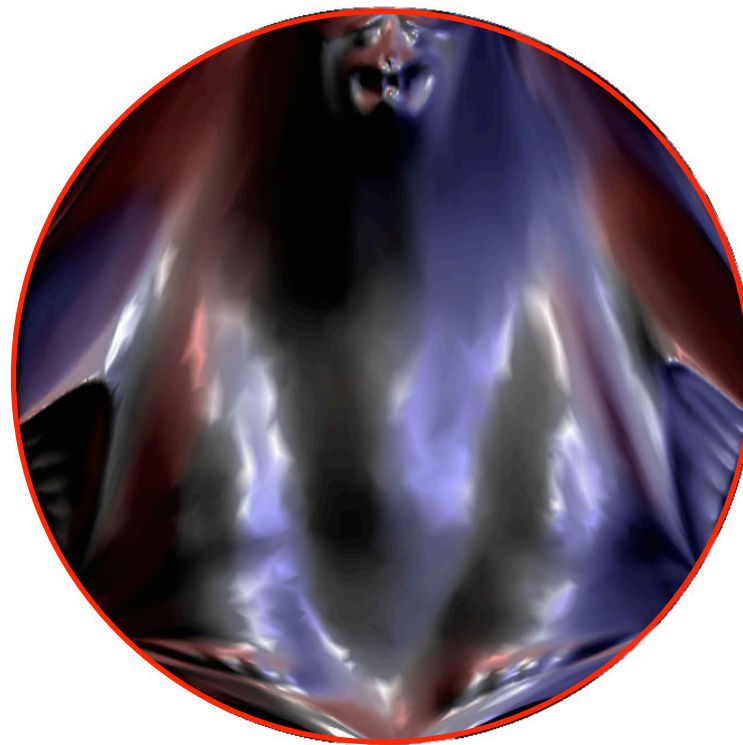
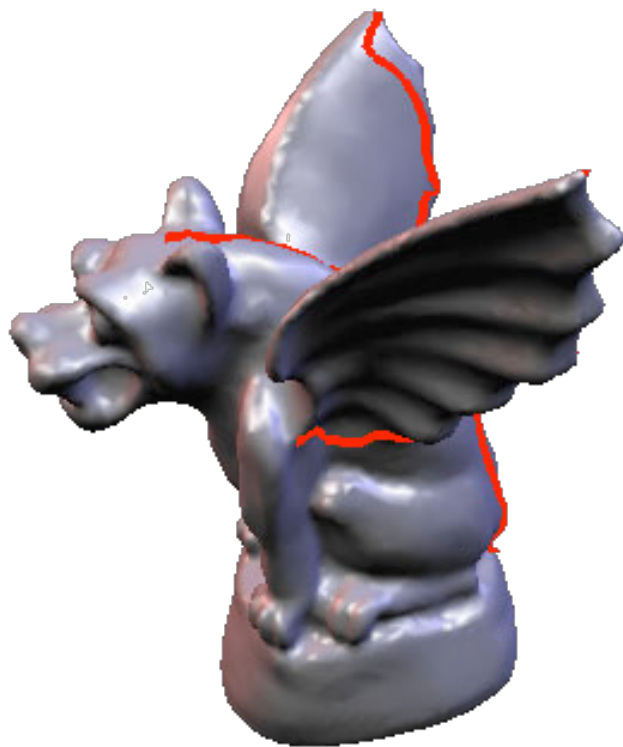




# Reducing Area Distortion

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- Introduce cuts  $\Rightarrow$  area distortion vs. continuity
- Often cuts are unavoidable (e.g., open sphere)

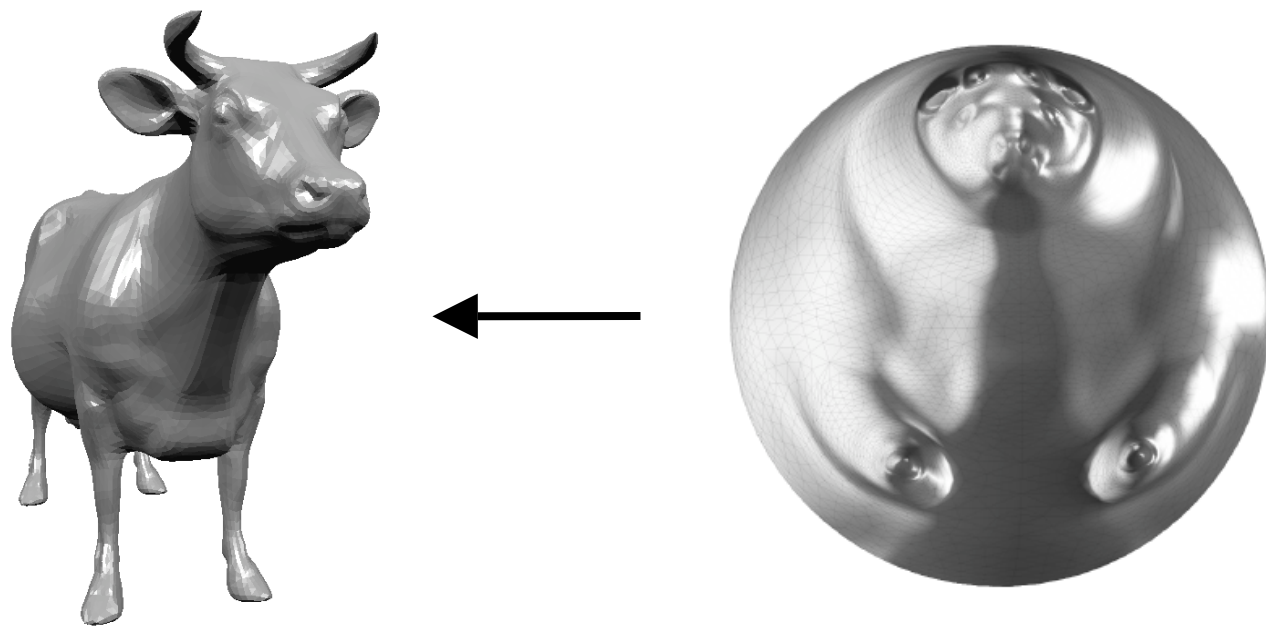


Treatment of boundary  
is important!

# Spherical Parameterization

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- Sphere is natural domain for genus-0 surfaces
- Additional constraint  $\|U\|^2 = 1$



- Naïve approach
  - Laplacian smoothing and back-projection
  - Obtain minimum for degenerate configuration



# Spherical Parameterization

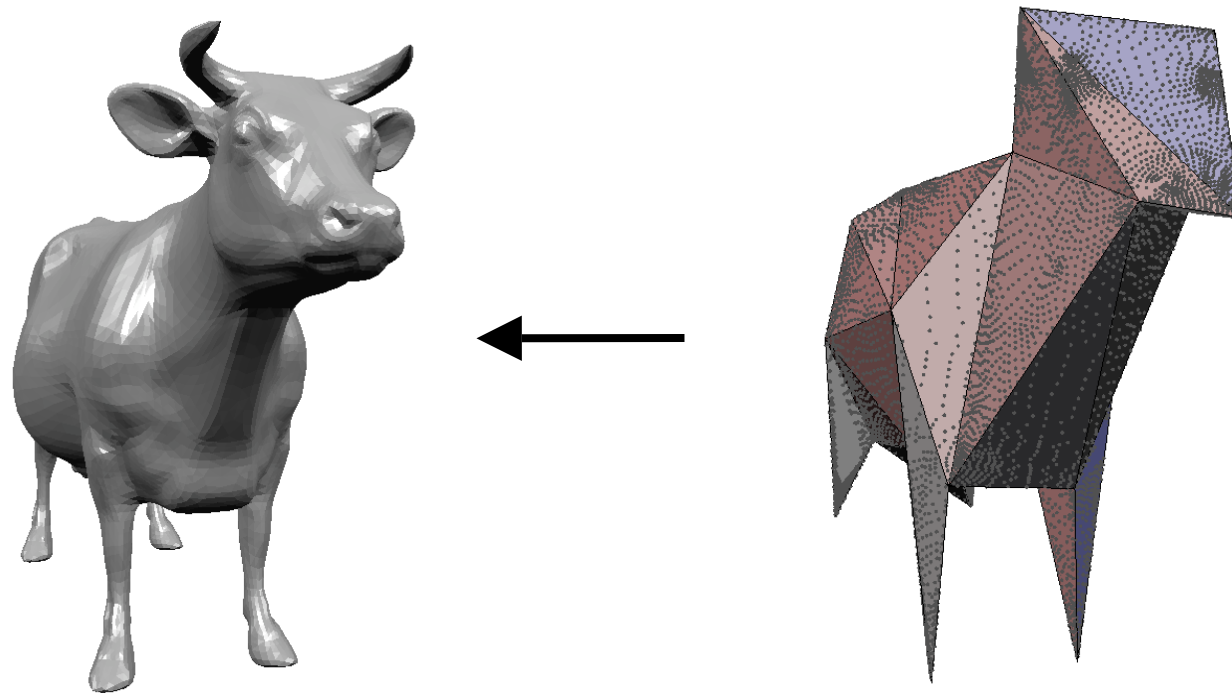
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- (Tangential) Laplacian Smoothing and back-projection
  - Minimum energy is obtained for *degenerate* solution
- Theoretical guarantees are expensive
  - [Gotsman et al. 2003]
- A compromise?!
  - Stereographic projection
  - Smoothing in curvilinear coordinates

# Arbitrary Topology

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- Piecewise linear domains
  - *Base mesh* obtained by *mesh decimation*
  - Piecewise maps
  - Smoothness



# Literature

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- Floater & Hormann: *Surface parameterization: a tutorial and survey*, Springer, 2005
- Lévy, Petitjean, Ray, and Maillot: *Least squares conformal maps for automatic texture atlas generation*, SIGGRAPH 2002
- Desbrun, Meyer, and Alliez: *Intrinsic parameterizations of surface meshes*, Eurographics 2002
- Sheffer & de Sturler: *Parameterization of faceted surfaces for meshing using angle based flattening*, Engineering with Computers, 2000.