# Mesh Parameterization

# **Richard Liu**

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**Surface parameterization:** construct a (bijective) map between two surfaces with similar topology

• Roots in cartography: how to make accurate map of Earth?

**Mesh parameterization:** construct a map between a triangular mesh and another surface (most often 2D plane)

# Mesh Parameterization Applications



Texture Mapping

Morphing

Databases



Normal Mapping



Mesh Completion



Remeshing





**Detail Transfer** 



Editing



Surface Fitting

Figure: Parameterization Applications

# Given a function $f: X \to Y$ ,

## Definition

f is **injective** or **one-to-one**, if  $\forall x, x' \in X$ ,  $f(x) = f(x') \Rightarrow x = x'$ .

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# Mathematical Framework

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#### Definition

f is **bijective** if f is both injective and surjective. Equivalently, f is bijective iff it is **invertible**.

Let  $\Omega \subset \mathbb{R}^2$  be a **simply connected** (without holes) region. Let  $f : \Omega \to \mathbb{R}^3$  be continuous and injective. The image of f is called a **surface** 

$$S = f(\Omega) = \{f(u, v) : (u, v) \in \Omega\}$$

We say that f is a **parameterization** of S over the **parameter domain**  $\Omega$ .

**Note:** By construction,  $f : \Omega \rightarrow S$  is trivially surjective. In practice injectivity is often what we care about.

## Example



# Example 2 з • Parameter domain: $\Omega = \{(u, v) \in \mathbb{R}^2 : u \in [0, 2\pi), v \in [0, 1]\}$ • Surface: $S = \{(x, y, z) \in \mathbb{R}^{:} x^{2} + y^{2} = 1, z \in [0, 1]\}$

# Mathematical Framework



# Mathematical Framework



#### Remark

A parameterization  $f : \Omega \to S$  is never unique. Given any bijection  $\gamma : \Omega \to \Omega$ ,  $g = f \circ \gamma$  is a parameterization of S over  $\Omega$ .

We can use f for deriving some key **intrinsic surface properties**, or properties that are independent of how the surface sits in space (extrinsic geometry).

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Another perspective: everything that is **knowable** to a tiny observable living on the surface (e.g. humans on the Earth)

# Mathematical Framework



[1.9] The intrinsic geometry of the surface of a crookneck squash: geodesics are the equivalents of straight lines, and triangles formed out of them may possess an angular excess of either sign, depending on how the surface bends:  $\mathcal{E}(\Delta_1) > 0$  and  $\mathcal{E}(\Delta_2) < 0$ .

A parameterization  $f : \Omega \subset \mathbb{R}^2 \to S \subset \mathbb{R}^3$  is **regular** if the tangent vectors  $f_u = \frac{\partial f}{\partial u}$  and  $f_v = \frac{\partial f}{\partial v}$  are always linearly independent.

**Note:**  $f_u$ ,  $f_v$  are functions from  $\mathbb{R}^2$  to  $\mathbb{R}^3$  and span the local tangent plane.

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$$n_f = \frac{f_u \times f_v}{||f_u \times f_v||}$$

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**Note:** Regularity is required for  $n_f$  to be nonzero everywhere. **Note:** The surface normal is always **independent** of the parameterization, making it an intrinsic property. We can also apply *f* towards deriving the **first and second fundamental forms.** They are fundamental precisely because they determine the key metric properties of a surface, such as the **gaussian curvature**, **mean curvature**, and **surface area**.

Given parameterization f, the **first fundamental form** is defined as

$$\mathsf{I}_{f} = \begin{pmatrix} f_{u} \cdot f_{u} & f_{u} \cdot f_{v} \\ f_{v} \cdot f_{u} & f_{v} \cdot f_{v} \end{pmatrix} = \begin{pmatrix} \mathsf{E} & \mathsf{F} \\ \mathsf{F} & \mathsf{G} \end{pmatrix}$$

Given parameterization f, the **first fundamental form** is defined as  $(f \cdot f - f \cdot f) - (F - F)$ 

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# Area of a Surface

Given parameterization  $f: \Omega \rightarrow S$ , the area A(S) can be found

$$A(S) = \int_{\Omega} \sqrt{\det(\mathsf{I}_f)} du dv$$

Given a twice-differentiable parameterization f, the **second** fundamental form is defined as

$$\Pi_{f} = \begin{pmatrix} f_{uu} \cdot n_{f} & f_{uv} \cdot n_{f} \\ f_{uv} \cdot n_{f} & f_{vv} \cdot n_{f} \end{pmatrix} = \begin{pmatrix} L & M \\ M & N \end{pmatrix}$$

The Gaussian curvature is

$$\mathcal{K} = \det(\mathsf{I}_f^{-1}\mathsf{I}\mathsf{I}_f) = \frac{\det\mathsf{I}\mathsf{I}_f}{\det\mathsf{I}_f} = \frac{LN - M^2}{EG - F^2}$$

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The mean curvature is

$$S = \frac{1}{2} \operatorname{trace}(\mathsf{I}_f^{-1}\mathsf{II}_f) = \frac{LG - 2MF + NE}{2(EG - F^2)}$$

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# Mathematical Framework



A surface S is **developable** if  $\forall p \in S$ , K(p) = 0, i.e. the Gaussian curvature is 0 everywhere on S.



# Three types of developable surfaces



The **Jacobian** of parameterization f is the 3 x 2 matrix of partial derivatives of f.

 $J_f = (f_u, f_v)$ 

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For any  $m \times n$  matrix J, the **singular value decomposition** (SVD) is given by

$$J = U \Sigma V^T$$

where  $\Sigma$  is an  $m \times n$  diagonal matrix, and U and V are  $m \times m$  and  $n \times n$  orthonormal matrices, respectively.

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By the above, the SVD of the Jacobian is

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

where  $\sigma_1$ ,  $\sigma_2$  are the **singular values**.

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### Remark

We can write the first fundamental form as

$$\mathsf{I}_{f} = J_{f}^{T} J_{f} = \begin{pmatrix} f_{u}^{T} \\ f_{v}^{T} \end{pmatrix} (f_{u} \ f_{v})$$

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It is clear  $I_f$  is symmetric.

Thus the eigenvalues of  $I_f$  are given by

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

#### Remark

For a matrix A, the singular values are the square roots of the eigenvalues of  $A^T A$ .
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 $\sigma_1$  and  $\sigma_2$  tell us **everything** about the **metric distortion** induced by the parameterization.

Parameterizations induce distortion in **lengths**, which can be further divided into distortion in **angles** and distortion in **areas**.

#### Properties of Parameterizations



Figure: SVD Decomposition of mapping  $\tilde{f}$ 

A parameterization is **conformal**, or **angle-preserving**, when the singular values of the Jacobian are equal, i.e.  $\sigma_1 = \sigma_2$ .

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#### Definition

A parameterization is **equiareal/authalic**, or **area-preserving**, when the singular values of the Jacobian multiply to 1, i.e.  $\sigma_1 \sigma_2 = 1$ .

A parameterization is conformal, or angle-preserving, when

 $\sigma_1 = \sigma_2.$ 

#### Definition

A parameterization is equiareal/authalic, or area-preserving, when  $\sigma_1 \sigma_2 = 1$ .

#### Definition

A parameterization is **isometric**, or **length-preserving** iff it is conformal and equiareal, i.e.  $\sigma_1 = \sigma_2 = 1$ .

#### So can we always find an isometric parameterization to the plane?

## So can we always find an isometric parameterization to the plane? **NOPE**

#### Theorem

(Gauss, 1827) Globally isometric parameterizations (from 3D to 2D) only exist for developable surfaces (i.e. K = 0 everywhere)

#### So how to find the "best" parameterization?

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Take bivariate non-negative function  $E : \mathbb{R}^2_+ \to \mathbb{R}_+$  that takes local distortion measures  $\sigma_1$  and  $\sigma_2$ , and has minimum defined according to objective.

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

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e.g. E global minimum at (1,1) = isometry objective

e.g. E minimal values along (x, x) for  $x \in \mathbb{R}_+ =$  conformal objective

Now let's consider triangle meshes specifically, which can be considered **piecewise linear surfaces**.

A **mesh** is a triangulation M = (V, E, F), where  $V = \{v_i\} \subset \mathbb{R}^3$ ,  $E = \{e_{ij}\}$ , and  $F = \{f_{ijk}\}$  are the vertex, edge, and face sets, respectively. More formally, edge  $e_{ij}$  represents the convex hull between vertices  $v_i$  and  $v_j$  (i.e. line segment), and face  $f_{ijk}$  is the convex hull of non-collinear points  $v_i, v_i, v_k$ .

We already mentioned **conformal** and **equiareal** maps. Another important property for applications to meshes is **bijectivity**.

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e.g. For texture mapping, want to be able to annotate parts of the texture with reference to unique region of surface

A mesh parameterization is **locally injective** if no triangles change orientation ("flip" or "fold over") during the parameterization.

#### Definition

A mesh parameterization is **globally bijective** if it is locally injective and the boundary of the parameterization does not intersect itself.

#### Mesh Parameterization Properties



Triangle Flip



Boundary Intersection

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In general, mesh parameterization methods can be characterized by the following set of properties:

- Distortion minimized: {angle (conformal), area (equiareal), distance (isometric)}
- Boundary: {fixed, free}
- Bijectivity: {global, local}

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- **②** Compute **barycentric coordinates** for the interior vertices
- Solve a **linear system** based around minimizing the spring energy of the mesh

# **Barycenteric coordinates** are simply a way of representing an interior point in a polygon (typically triangle) as a linear combination of its vertices.

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#### Definition

For a point x in the interior of a triangle  $f_{ijk} = \{v_i, v_j, v_k\}$ , values  $\lambda_i$ ,  $\lambda_j$ ,  $\lambda_k$  are **barycentric coordinates** of x with respect to the vertices of  $f_{ijk}$  if:

$$\mathbf{1} \quad \mathbf{x} = \lambda_i \mathbf{v}_i + \lambda_j \mathbf{v}_j + \lambda_k \mathbf{v}_k$$

$$\lambda_i + \lambda_j + \lambda_k = 1$$

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**Note:** above definition can be easily generalized to n-gons, but barycentric coordinates are **only unique** when *x* has 3 neighbors.

Fixed boundary methods primarily differ on how to construct the barycentric coordinates, and how to deal with the boundary.

Typically want to choose a **convex** parameter domain. Why?

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Typically want to choose a convex parameter domain. Why?

#### Theorem

Tutte (1963) For a parameterization  $f : \Omega \rightarrow S$  constructed by fixing the boundary and computing positive barycentric coordinates for the interior vertices, if  $\Omega$  is convex, then f is bijective.

**Tutte embeddings.** Tutte first to introduce the above-described framework into the mesh parameterization context with his seminal work on straight-line embeddings of planar graphs.

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- $\lambda_{ij} = 1/|N_i|$  defined uniformly (**not** barycentric)
- Guarantee bijectivity under certain constraints
- No guarantee of distortion minimization

## Harmonic parameterization. Eck et al.'s method makes use of harmonic coordinates, or cotangent weights (very famous).

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$$w_{ij} = \cot \gamma_{ij} + \cot \gamma_{ji}$$



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- Minimize harmonic energy  $(\triangle f(u, v) = 0)$
- Harmonic condition weaker than conformal
- Weights can be negative when angles are obtuse  $\Rightarrow$  non-bijective parameterization

#### Other Coordinates.

- Wachspress coordinates (Wachspress 1975)
- Mean value coordinates (Floater 2003)

#### Pros

• Weights can be computed for every interior vertex even if neighbors not coplanar or more than 3 vertex neighbors

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- Weights can be computed for every interior vertex even if neighbors not coplanar or more than 3 vertex neighbors
- Linear complexity

#### Cons

- High distortion when surface boundary highly non-convex
- Often no "natural" way of distributing parameter points along the boundary.

# Fixed Boundary Methods

#### Cons



Figure 4.1: A: a mesh cut in a way that makes it homeomorphic to a disk, using the *seamster* algorithm [Sheffer and Hart, 2002]; B: Tutte-Floater parameterization obtained by fixing the border on a square; C: parameterization obtained with a free-boundary parameterization [Sheffer and de Sturler, 2001].

Workarounds

• Virtual boundary: augment 3D boundary with extra triangles (Lee et al. 2002)

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- **Scaffolding:** similar idea, but iteratively remeshes virtual boundary based on some distortion energy (Jiang et al. 2017)

## Fixed Boundary Methods



Fig. 3.6 (a) Adding a virtual boundary to the original mesh. (b) Shape Preserving [32] parameterization of the original mesh. (c) Parameterization of the original mesh and its virtual boundary [74]. The virtual boundary vertices are fixed, allowing the real boundary vertices to move.

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**LSCM.** (Levy et al. 2002) The least squares conformal maps method seeks to minimize the following **conformal energy** 

$$E_{LSCM} = E_C = \frac{1}{2} \int_S ||f_v - \operatorname{rot}_{90}(f_u X)||^2 dp = \frac{(\sigma_1 - \sigma_2)^2}{2}$$

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Intuition: the gradient vectors  $f_u$  and  $f_v$  are **orthogonal** and **have the same norm**.

### Free Boundary Methods



Figure 4.9: A conformal parameterization transforms an elementary circle into an elementary circle.

**DCP.** (Desbrun et al 2002) Discrete conformal parameterization minimizes the **dirichlet energy**.

#### Definition

Given a parameterization  $f : \Omega \subset \mathbb{R}^2 \to S \subset R^3$ , the **Dirichlet** energy measures the integral of the squared norm of the gradients.

$$E_D = \frac{1}{2} \int_S ||f_u||^2 + ||f_v||^2 dp$$

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The Dirichlet energy can also be expressed in terms of the singular values  $\sigma_1$ ,  $\sigma_2$  of the Jacobian

$$E_D = \frac{\sigma_1^2 + \sigma_2^2}{2}$$

Using the singular value definitions, we can easily see that **DCP** and **LSCM** are equivalent methods.

$$E_D - E_C = \sigma_1 \sigma_2 = \det(J) = \frac{\operatorname{Area}(\Omega)}{\operatorname{Area}(S)}$$

Using the singular value definitions, we can easily see that **DCP** and **LSCM** are equivalent methods.

$$E_D - E_C = \sigma_1 \sigma_2 = \det(J) = \frac{\operatorname{Area}(\Omega)}{\operatorname{Area}(S)}$$

*Recall:*  $\Omega$  is the parameter domain (2D) and *S* is the surface (3D). So Dirichlet and conformal energies are the same up to a fixed boundary (choice of pinned vertices) in the parameter domain.

• Require two pinned vertices to avoid trivial solution (heuristic: two diameter vertices)

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- Require two pinned vertices to avoid trivial solution (heuristic: two diameter vertices)
- LSCM energy a **flawed** metric: scaled by area of parameter domain
- No guarantee of local or global bijectivity
- Linear (fast) and empirically lower distortion than fixed boundary methods

#### LSCM/DCP. Extensions

• Spectral conformal parameterization (Mullen et al. 2008): find solution to minimizing conformal energy **without** needing to pin vertices

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- Spectral conformal parameterization (Mullen et al. 2008): find solution to minimizing conformal energy **without** needing to pin vertices
  - Find Fiedler vector solution u from  $L_c u = \lambda B u$

#### LSCM/DCP. Extensions

- Spectral conformal parameterization (Mullen et al. 2008): find solution to minimizing conformal energy **without** needing to pin vertices
- Hierarchical LSCM (Ray and Levy 2003): Speed-up using hierarchical solver

**MIPS.** (Hormann and Greiner 2000) First method to compute natural boundary. Minimizes the *Dirichlet energy per parameter-space area* 

$$\mathcal{K}_{\mathcal{F}}(J_{\mathcal{T}}) = ||J_{\mathcal{T}}||_{\mathcal{F}}||J_{\mathcal{T}}^{-1}||_{\mathcal{F}} = \frac{\sigma_1^2 + \sigma_2^2}{\sigma_1 \sigma_2} = \frac{\mathsf{trace}(\mathsf{I}_{\mathcal{T}})}{\det J_{\mathcal{T}}}$$

# **MIPS.** (Hormann and Greiner 2000) Iteratively move each vertex to reduce energy, checks for flips, and checks for boundary overlaps.

#### MIPS. Properties

- Nonlinear (slow)
- Global bijectivity

**Angle Based Flattening.** (Sheffer and de Sturler 2000) Based on the observation: a planar triangulation is defined by the corner angles of triangles (up to similarity).

Unlike previous methods, problem is defined in angle space.

**Angle Based Flattening.** (Sheffer and de Sturler 2000) Minimize the objective

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

where  $\beta_i$  are the known 3D angles and  $\alpha_i$  are the unknown 2D angles.

# **Angle Based Flattening.** (Sheffer and de Sturler 2000) Require constraints on 2D angles for "valid triangulation"

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- All angles positive
- Angles in each triangle sum to  $\pi$
- Sum of angles around each vertex is  $2\pi$
- Edges shared by adjacent triangles have same length

#### Angle Based Flattening. Properties

• Locally bijective (but not global)

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- Locally bijective (but not global)
- Non-linear (slow) and unstable for large meshes

#### Angle Based Flattening. Extensions

 Zayer et al (2003): Enforce convex boundaries on parameter domain ⇒ global bijectivity

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- Zayer et al (2003): Enforce convex boundaries on parameter domain ⇒ global bijectivity
- Kharevych et al (2006): Introduce cone singularities ⇒ global parameterization. Continuous up to translation and rotation, except at singularities.

# Free Boundary Methods

#### Angle Based Flattening. Extensions




Parameterization with uniform weights [128] on a circular domain.



Parameterization with harmonic weights [28] on a circular domain.



Parameterization with mean value weights [33] on a circular domain.



Parameterization with LSCM [79].

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## Comparisons



Stretch minimizing parameterization [107].

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- Ricci flows
- Circle packing
- Discrete conformal equivalence
- Cone singularities
- Etc...

## Mesh Parameterization: Theory and Practice (2008) Mesh Parameterization Methods and Their Applications