## Geometric Modeling

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## Surface Parameterization

$\square$ Introduction
$\square$ Applications
$\square$ What is Parameterization?
$\square$ Parameterization Methods

## Problem

$\square$ 1-1 mapping from domain to surface $\square$ Original application:

- Texture mapping
$\square$ Images have a natural parameterization
- Goal: map onto surfaces
$\square$ Geometry processing
- Approximation
- Remeshing

- Data fitting
$\square$ Input: Piecewise (PL) triangular meshes


## Introduction

$\square$ A parameterization of a surface is a one-to-one mapping from a suitable domain to the surface.

## Introduction

$\square$ In general, the parameter domain itself will be a surface.
$\square$ Constructing a parameterization means mapping a surface into another.
$\square$ Usually the surfaces are either represented by or approximated by triangular meshes and the mappings are piecewise linear.

## Applications

$\square$ Scattered data fitting.
$\square$ Reparameterization of Spline surfaces
$\square$ Texture-mapping
$\square$ Mesh compression
$\square$ Surface approximations \& remeshing

## What is Parameterization?



## What is Parameterization?


$\square$ How to get $\mathbf{P}$ from $\mathbf{S}$ ?

- for each vertex of $\boldsymbol{S}$, find its $(u, v)$
- from $(u, v)$ of $P$, map image to $\boldsymbol{S}$
$\square$ A parameterization of a surface is a mapping $\rho:(x, y, z)->(u, v)$ from 3D space to 2D space


## Problem Definition

$\square$ Given a surface (mash) $S$ in $R^{3}$ and a domain $D$ find $\rho: D \leftrightarrow S$ (one-to-one)


## Recall: Applications

$\square$ Texture-mapping

- $\rho:(x, y, z)->(u, v)$ from 3D to 2D
$\square$ Remeshing

$$
\rho^{-1}:(u, v)->(x, y, z) \text { from 2D to 3D }
$$

## Typical Domains

sub-domain of $R^{2}$

- genus-0 + boundary
sphere
- closed genus-0
base mesh
- all (closed) models



## Applications

texture mapping

morphing

reconstruction


## Texture Mapping

$\square$ Real life objects not uniform in terms of color
$\square$ Texturing - define color for each point on object surface
$\square$ Map 2D texture to model surface:

- Have texture pattern defined over ( $u, v$ ) domain (Image)
- Assign (u,v) coordinates to each point on object surface



## Morphing

$\square$ Morphing requires one-to-one correspondence between the surfaces of the two models


## Normal / Bump Mapping



## Remeshing \& Surface Fitting



## More Applications



## Mesh Parameterization (2D)

$\square$ Problem definition:

- Input:
triangulated surface mesh in 3D
- Output:
- valid 2D mesh with same connectivity \& minimal metric distortion
- Mapping defined by vertex correspondence + barycentric coordinates
- Validity - no inverted (overlapping) triangles


## Parameterization in 2D

$\square$ Can do only for genus 0 surfaces with boundary
$\square$ Metrics preserved fully only for developable surfaces (Gaussian curvature $=0$ )


## Distortion

$\square$ Function of second fundamental form

$$
J=\left(\begin{array}{ll}
a & b \\
b & c
\end{array}\right)=\left(\begin{array}{ll}
\left(\frac{\partial f}{\partial u}\right)^{2} & \frac{\partial f}{\partial u} \frac{\partial f}{\partial v} \\
\frac{\partial f}{\partial u} \frac{\partial f}{\partial v} & \left(\frac{\partial f}{\partial v}\right)^{2}
\end{array}\right)
$$

$\square$ Isometry (0-distortion) if $J$ is identity matrix
$\square$ Components measure

- Shear (angles)
- Stretch (lengths)
- Conformality: angles + equal stretch


## Mapping Properties

$\square$ Validity: no folded over triangles
$\square$ Distortion - preserve (as mush as possible) lengths, angles, and areas

- isometric (length-preserving) mappings
- conformal (angle-preserving) mappings
- equiareal (area-preserving) mappings
$\square$ (Theorem)
isometric $\Leftrightarrow$ conformal + equiareal


## Barycentric Combination

$$
\begin{aligned}
\square \mathbf{p} & =\sum \lambda_{i} \mathbf{p}_{i}, \\
i & =1 \ldots n, \sum \lambda_{i}=1
\end{aligned}
$$


i.e., a combination of its neighbors

## Barycentric Combination

Choose $\mathbf{u}_{n+1}, \ldots, \mathbf{u}_{N}$ to be the vertices of any $K$-sided convex polygon in an anticlockwise sequence.

For each $i \in\{1, \ldots, n\}$, choose any set of real numbers $\lambda_{i, j}$ for $j=1, \ldots, N$ such that

$$
\lambda_{i, j}=\left\{\begin{array}{ll}
>0 & (i, j) \in E \\
0 & (i, j) \notin E
\end{array} \quad \sum_{j=1}^{N} \lambda_{i, j}=1\right.
$$

and define $\mathbf{u}_{1}, \ldots, \mathbf{u}_{n}$ to be the solution of the linear system of equations.

$$
\mathbf{u}_{i}=\sum_{j=1}^{N} \lambda_{i, j} \mathbf{u}_{j}
$$

## Barycentric Combination

$$
\mathbf{u}_{i}=\sum_{j=1}^{N} \lambda_{i, j} \mathbf{u}_{j} \Rightarrow \mathbf{u}_{i}-\sum_{j=1}^{n} \lambda_{i, j} \mathbf{u}_{j}=\sum_{j=n+1}^{N} \lambda_{i, j} \mathbf{u}_{j}
$$

By considering the two components $u_{i}$ and $v_{i}$ of $\mathbf{u}_{i}$ separately this is equivalent to the matrix equation

$$
\mathbf{A u}=\mathbf{b}
$$

where the matrix $\mathbf{A}$ is $n \times n$ having elements

$$
a_{i, j}= \begin{cases}1 & i=j \\ -\lambda_{i, j} & i \neq j\end{cases}
$$

## 2D Barycentric Embeddings

$\square$ Fix 2D boundary to convex polygon
$\square$ Define embedding as a solution of

$$
\begin{array}{ll}
\mathbf{W} x=b_{x} \\
\mathbf{W} y=b_{y}
\end{array} \quad w_{i j}=\left\{\begin{array}{ll}
<0 & (i, j) \in E \\
-\sum_{j \neq i} w_{i j} & (i, i) \\
0 & \text { otherwise }
\end{array}\right\}
$$

- $\mathbf{W}$ is symmetric: $w_{i j}=w_{j i}$
- weights $w_{i j}$ control parameterization shape


## Weights - Uniform (Tutte)

$\square$ Set

$$
w_{i j}=1 \quad\left(\lambda_{i, j}=\frac{1}{d_{i}}\right)
$$

$\square$ No shape information - equilateral triangles
$\square$ Fastest to solve


## Weights - Shape Preserving (Floater)

For each $l \in\left\{1, \ldots, d_{i}\right\}, \overrightarrow{\mathbf{p}_{i}}$ will intersect $\overline{\mathbf{p}_{r(l)}} \mathbf{p}_{r(l)+l}$, then we can have $\mathbf{p}=\delta_{1} \mathbf{p}_{l}+\delta_{2} \mathbf{p}_{r(l)}+\delta_{3} \mathbf{p}_{r(l)+1} \quad \delta_{1}+\delta_{2}+\delta_{3}=1$

Define $\mu_{k, l \prime}$ for $k=1, \ldots, d_{i}$, by $\mu_{l, l}=\delta_{1}$, $\mu_{r(l), l}=\delta_{2}, \mu_{r(l)+1, l}=\delta_{3}$, and $\mu_{k, l}=0$ otherwise. Then for each $l$ we now find

$$
\mathbf{p}=\sum_{k=1}^{d_{i}} \mu_{k, l} \mathbf{p}_{k} \quad \sum_{k=1}^{d_{i}} \mu_{k, l}=1
$$

Finally define $\lambda_{i, j}=\frac{1}{d_{i}} \sum_{l=1}^{d_{i}} \mu_{j, l}$



## Weights - Harmonic Mapping

$\square$ Quasi - conformal : minimize angular distortion
$\square$ Locally, preserving angles preserves distances (ratios)
$\square$ For fixed boundary have unique harmonic map which minimizes conformal energy

- Rieman theorem: any $\mathrm{C}^{1}$ continuous surface in $R^{3}$ can be mapped conformally to fixed domain in $\mathrm{R}^{2}$
- Nearly true for meshes


## Weights - Harmonic Mapping

- Approximate harmonic map for fixed boundary
$\square$ Represent as configuration of springs on mesh edges

$$
E(v)=\frac{1}{2} \sum_{(i, j)} w_{i j}\left\|v_{i}-v_{j}\right\|^{2}
$$

- Spring coefficients

$$
w_{i j}=\frac{L_{i 1}^{2}+L_{j 1}^{2}-L_{i j}^{2}}{A_{1}}+\frac{L_{i 2}^{2}+L_{j 2}^{2}-L_{i j}^{2}}{A_{2}}
$$

$$
=\frac{\cot \left(\alpha_{i j}\right)+\cot \left(\beta_{i j}\right)}{2}
$$


$L_{i j}$ - edge Fength in 3D
$A_{i j k}$ - triangle area in 3D
$\alpha_{i j}$ and $\beta_{i j}$ - opposing angles in 3D

## Weights - Conformal Mapping

$\square$ Represent as configuration of springs on mesh edges

$$
E(v)=\sum_{(i, j)} w_{i j}\left\|u_{i}-u_{j}\right\|^{2}
$$

$\square \quad$ Spring coefficients

$$
w_{i j}=\cot \left(\alpha_{i j}\right)+\cot \left(\beta_{i j}\right)
$$

$\square \alpha_{i j}$ and $\beta_{i j}$ - opposing angles in 3D


## Barycentric Formulation

$\square E(v)$ minimum reached when gradient equal 0

$$
\frac{\partial E(v)}{\partial v_{i}}=\sum_{j} w_{i j}\left(v_{i}-v_{j}\right)=0
$$

$\square$ Barycentric embedding formulation
$\square$ Can have negative weights - does not guarantee validity


## Weights - Mean Value

$\square$ Set

$$
w_{i j}=\frac{\left(\tan \left(\gamma_{i j} / 2\right)+\tan \left(\delta_{i j} / 2\right)\right) / 2}{\left\|v_{i}-v_{j}\right\|}
$$


$\square$ Result visually identical to conformal
$\square$ No negative weights - always valid


## Comparison



## Practical Implementation

## $\square$ Boundary

- Popular options: Square, circle, triangle
- Application specific
$\square$ Reconstruction - rectangle
$\square$ Mapping to base mesh-triangle
$\square$ Spreading points along boundary
$\square$ Cord length
$\square$ Solve $\mathbf{W} x=b_{x}$

$$
\mathbf{W} y=b_{y}
$$

- Right hand side determined from boundary vertices


## Practical Implementation

$\square$ Solving linear system expensive $\left(O\left(n^{3}\right)\right)$
$\square$ Use iterative solution:

- Get initial guess for interior nodes
- While conditions not met:
$\square$ Set each interior node to weighted average of neighbors:

$$
v_{i}=\frac{1}{\sum_{(i, j)} w_{i j}} \sum_{(i, j)} w_{i j} v_{j}
$$

- Stopping conditions:
$\square$ Convergence: vertices do not move
- Exceed maximal number of iterations
- Parameterization is valid
$\square$ Solution exists \& is reached thanks to matrix structure


## Fixed vs. Free Boundary

$\square$ Fixed

- Useful when boundary fixed a priory (e.g. mapping to base mesh)
- Increase distortion
$\square$ Free
- Typically less distortion


## Local Unfolding

$\square$ While not all mesh flattened

- Select seed triangle \& map as is to 2D
- Define front - boundary of unfolded patch
- Assign cost to each vertex adjacent to boundary-amount of distortion caused by mapping it to 2D
- Map best current vertex to 2D (if cost below threshold), add it to front \& recompute adjacent costs


## Local Unfolding

$\square$ Advantages

- Bounded distortion
- Simple
- Drawback
- Generate long seams parameterization/texture discontinuities


## Angle Based Flattening (ABF)

$\square$ Fact:

- Triangular 2D mesh is defined by its angles
$\square$ Define problem in angle space
$\square$ Angle based formulation:
- Distortion as function of angles
- Validity - set of angle constraints



## Constrained Minimization


$\square$ Objective: minimize (relative) deviation of angles

$$
F(\alpha)=\sum_{i, j} w_{i}^{j}\left(\alpha_{i}^{j}-\beta_{i}^{j}\right)^{2}
$$

$\square$ Initial choice for weights:

$$
w_{i}^{j}=\beta_{i}^{j-2}
$$

## Constraints

$$
\left.\left.\begin{array}{l}
g^{1}(\alpha) \equiv \alpha_{i}^{j} \geq \varepsilon \\
g^{2}(\alpha) \equiv \alpha_{i}^{1}+\alpha_{i}^{2}+\alpha_{i}^{3}=\pi \\
g^{3}(\alpha) \equiv \sum_{k} \alpha_{i}^{j(k)}=2 \pi \\
g^{4}(\alpha) \equiv \prod_{k} \sin \left(\alpha_{i}^{j(k)-1}\right)-\prod_{k} \sin \left(\alpha_{i}^{j(k)+1}\right)=0 \\
\frac{l_{1}}{l_{2}} \cdots \frac{l_{6}}{l_{1}}=\frac{\sin \left(\alpha_{1}\right)}{\sin \left(\alpha_{2}\right)} \\
\sin \left(\alpha_{1}\right) \\
\sin \left(\alpha_{2}\right)
\end{array}\right) \frac{\sin \left(\alpha_{6}\right)}{\sin \left(\alpha_{1}\right)}\right)
$$

## Solution

$\square$ Use Lagrange Multipliers
$F^{*}(\alpha, \mu)=F(\alpha)+\mu_{1} g^{2}(\alpha)+\mu_{2} g^{3}(\alpha)+\mu_{3} g^{4}(\alpha)$
$\square$ Solve the min-max problem (minimum on $\alpha$, maximum on $\mu$ )
$\square$ Reached when all derivatives are zero

- Have non-linear system of equations
$\square$ Use Newton method to solve


## ABF Summary

$\square$ Advantages

- No fixed boundary - less distortion
- No flipped triangles
- Proven to converge to solution for any valid input
$\square$ Drawbacks
- Expensive - solve non linear system
- Linear sub-systems can't be solved iteratively FAST
- Can have boundary overlaps
- Can't handle multiple boundary loops

