





Mesh Smoothing

Filter out high frequency components for noise removal



Desbrun, Meyer, Schroeder, Barr: Implicit Fairing of Irregular Meshes using Diffusion and Curvature Flow, SIGGRAPH 99





Advanced Filtering / Signal Processing



Pauly, Kobbelt, Gross: Point-Based Multi-Scale Surface Representation, ACM TOG 2006



Guskow, Sweldens, Schroeder: Multiresolution Signal Processing for Meshes, SIGGRAPH 99



Fair Surface Design



Hole-filling with energy-minimizing patches







Mesh deformation



Botsch, Kobbelt: An intuitive framework for real-time freeform modeling, SIGGRAPH 04



Kobbelt, Campagna, Vorsatz, Seidel: Interactive Multi-Resolution Modeling on Arbitrary Meshes, SIGGRAPH 98



Outline

- Motivation
- Smoothing as Diffusion
- Smoothing as Energy Minimization
- Alternative Approaches



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- Motivation
- Smoothing as Diffusion
 - Spectral Analysis
 - Laplacian Smoothing
 - Curvature Flow
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Filter Design

- Assume high frequency components = noise
- Low-pass filter



Filter Design

- Assume high frequency components = noise
- Low-pass filter





Filter Design

- Assume high frequency components = noise
- Low-pass filter
 - damps high frequencies (ideal: cut off)
 - e.g., by convolution with Gaussian (spatial domain)
 - multiply with Gaussian (frequency domain)
- Fourier Transform



Spectral Analysis and Filter Design

Univariate: Fourier Analysis

$$F(\varphi) = \frac{1}{2\pi} \int_{-\infty}^{\infty} f(t) e^{-i\varphi t} dt$$
 frequency domain spatial domain

- Example: Low-pass filter
 - Damp (ideally cut off high frequencies)
 - Multiply F with Gaussian (= convolve f with Gaussian)
- Are there "geometric frequencies"?



Spectral Analysis and Filter Design

Univariate: Fourier Analysis

$$F(\varphi) = \frac{1}{2\pi} \int_{-\infty}^{\infty} f(t) e^{-i\varphi t} dt$$

Generalization

$$\Delta e^{i\varphi t} = \frac{\partial^2}{\partial t^2} e^{i\varphi t} = -\varphi^2 e^{i\varphi t}$$

- $e^{i\varphi t}$ are eigenfunctions of the Laplacian
- use them as basis functions for geometry



Spectral Analysis

• Eigenvalues of Laplacian ≅ frequencies



B. Vallet, B. Levy. *Spectral geometry processing with manifold harmonics*. Technical report, INRIA-ALICE, 2007.



Spectral Analysis

Low-pass filter
 reconstruction from
 eigenvectors associated with low frequencies



B. Vallet, B. Levy. *Spectral geometry processing with manifold harmonics*. Technical report, INRIA-ALICE, 2007.



Spectral Analysis

- Eigenvalues of Laplace matrix ≅ frequencies
- Decomposition in frequency bands is used for mesh deformation
 - often too expensive for direct use in practice!
 difficult to compute eigenvalues efficiently
- For smoothing apply diffusion...



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Diffusion

Diffusion equation



diffusion constant $\frac{\partial}{\partial t}x = \frac{1}{\mu}\Delta x$ \int Laplace operator



Diffusion

Diffusion equation







Laplacian Smoothing

Discretization of diffusion equation

$$\frac{\partial}{\partial t}\mathbf{p}_i = \mu \Delta \mathbf{p}_i$$

- · Leads to simple update rule
 - iterate

$$\mathbf{p}'_i = \mathbf{p}_i + \mu \ dt \ \Delta \mathbf{p}_i$$
 explicit Euler integration

- until convergence















Laplacian Smoothing



0 Iterations



5 Iterations



20 Iterations







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Curvature Flow

- Curvature is independent of parameterization
- Flow equation

surface normal



• We have $\Delta_S \mathbf{p} = -2H\mathbf{n}$

Laplace-Beltrami operator



Curvature Flow

• Mean curvature flow $\frac{\partial}{\partial t}$

$$\frac{\partial}{\partial t}\mathbf{p} = \mu \Delta_S \mathbf{p}$$

- use discrete Laplace-Beltrami operator (cot weights)

Compare to uniform discretization of Laplacian



Comparison Original Umbrella Laplace-Beltrami 31 Mark Pauly ÉCOLE POLYTECHNIQUE Fédérale de Lausanne

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- Smoothing as Energy Minimization
 - membrane energy
 - thin-plate energy
- Alternative Approaches



Energy Minimization

- Penalize "un-aesthetic behavior"
- Measure fairness
 - principle of the simplest shape
 - physical interpretation
- Minimize energy functional
 - examples: membrane / thin plate energy



Non-Linear Energies

• Membrane energy (surface area)

$$\int_{\mathcal{S}} \mathrm{d}s \to \min \quad \text{with} \quad \delta \mathcal{S} = \mathbf{c}$$

• Thin-plate surface (curvature)

$$\int_{\mathcal{S}} \kappa_1^2 + \kappa_2^2 ds \rightarrow \min \quad \text{with} \quad \delta \mathcal{S} = \mathbf{c}, \ \mathbf{n}(\delta \mathcal{S}) = \mathbf{d}$$

• Too complex... simplify energies



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Membrane Surfaces

Surface parameterization

 $\mathbf{p}:\Omega\subset {\rm I\!R}^2\to {\rm I\!R}^3$

• Membrane energy (surface area)

$$\int_{\Omega} \left\| \mathbf{p}_u \right\|^2 + \left\| \mathbf{p}_v \right\|^2 \mathrm{d}u \mathrm{d}v \to \min$$



Membrane Surfaces

Surface parameterization

$$\mathbf{p}:\Omega\subset {\rm I\!R}^2\to {\rm I\!R}^3$$

• Membrane energy (surface area)

$$\int_{\Omega} \|\mathbf{p}_u\|^2 + \|\mathbf{p}_v\|^2 \,\mathrm{d} u \mathrm{d} v \to \min$$

Variational calculus

$$\Delta \mathbf{p} = 0$$



Thin-Plate Surfaces

Surface parameterization

$$\mathbf{p}:\Omega\subset {\rm I\!R}^2\to {\rm I\!R}^3$$

• Thin-plate energy (curvature)

$$\int_{\Omega} \left\| \mathbf{p}_{uu} \right\|^2 + 2 \left\| \mathbf{p}_{uv} \right\|^2 + \left\| \mathbf{p}_{vv} \right\|^2 \mathrm{d}u \mathrm{d}v \to \min$$

Variational calculus

$$\Delta^2 \mathbf{p} = 0$$



Energy Functionals



Analysis

Minimizer surfaces satisfy Euler-Lagrange PDE

$$\Delta_{\mathcal{S}}^{k}\mathbf{p}=0$$

• They are stationary surfaces of Laplacian flows

$$\frac{\partial \mathbf{p}}{\partial t} = \Delta_{\mathcal{S}}^k \mathbf{p}$$

• Explicit flow integration corresponds to iterative solution of linear system



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Alternative Approaches

- Anisotropic Diffusion
 - Data-dependent
 - Non-linear
- Normal filtering



- Smooth normal field and reconstruct (mesh editing)
- Non-linear PDEs
 - Avoid parameter dependence for fair surface design
- Bilateral Filtering



Example of Bilateral Filtering



Jones, Durand, Desbrun: Non-iterative feature preserving mesh smoothing, SIGGRAPH 2003

Literature

- Taubin: A signal processing approach to fair surface design, SIGGRAPH 1996
- Desbrun, Meyer, Schroeder, Barr: *Implicit Fairing of Irregular Meshes using Diffusion and Curvature Flow*, SIGGRAPH 99
- Botsch, Kobbelt: An Intuitive Framework for Real-Time Freeform Modeling, SIGGRAPH 2004
- Fleishman, Drori, Cohen-Or: *Bilateral mesh denoising*, SIGGRAPH 2003
- Jones, Durand, Desbrun: *Non-iterative feature preserving mesh smoothing*, SIGGRAPH 2003

