Digital Geometry -- Surface Deformations

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http://jjcao.github.io/DigitalGeometry/

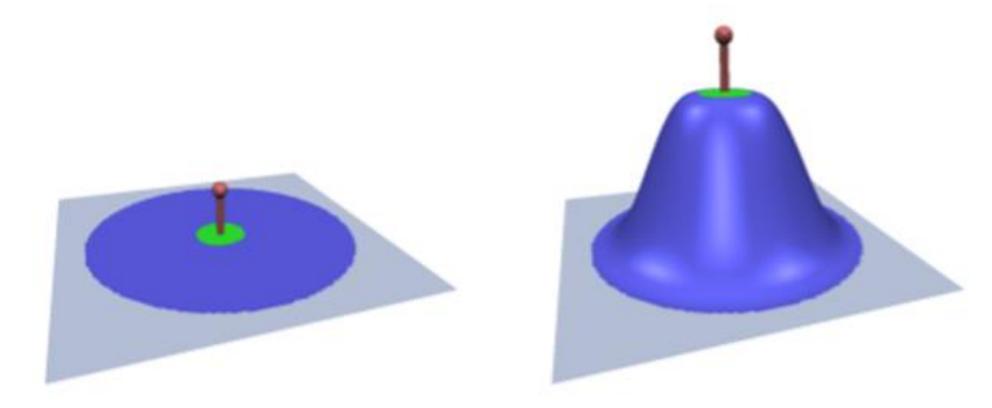
Overview

- Surface-based deformation
 - Energy minimization
 - Multiresolution editing
 - Differential coordinates

Physically-Based Deformation

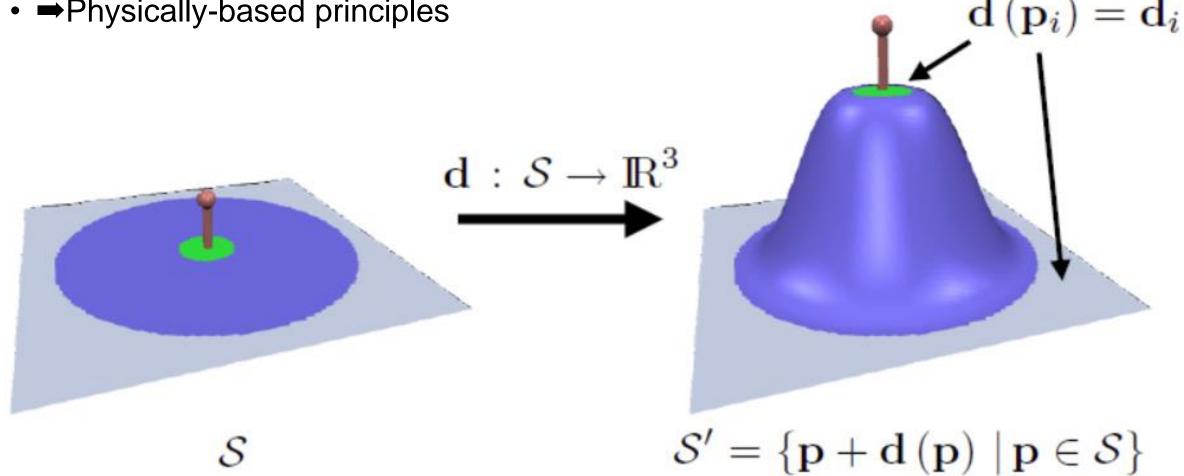
Modeling Metaphor

- Paint three surface regions
 - Support region (blue)
 - Fixed vertices (gray)
 - Handle regions (green)



Modeling Notation

- Mesh deformation by displacement function d
 - Interpolate prescribed constraints
 - Smooth, intuitive deformation
 - →Physically-based principles



Shell Deformation Energy

Stretching

- Change of local distances
- Captured by 1st fundamental form

Bending

- Change of local curvature
- Captured by 2nd fundamental form

Stretching & bending is sufficient

 Differential geometry: "1st and 2nd fundamental forms determine a surface up to rigid motion."

$$\left(\int_{\Omega} k_s \left\| \mathbf{I} - \overline{\mathbf{I}} \right\|^2 \right)$$

$$\mathbf{I} = \begin{bmatrix} \mathbf{x}_u^T \mathbf{x}_u & \mathbf{x}_u^T \mathbf{x}_v \\ \mathbf{x}_v^T \mathbf{x}_u & \mathbf{x}_v^T \mathbf{x}_v \end{bmatrix}$$

$$\left(\int_{\Omega} k_b \left\| \mathbf{I} \mathbf{I} - \bar{\mathbf{I}} \mathbf{I} \right\|^2 \right)$$

$$\mathbf{II} = \begin{bmatrix} \mathbf{x}_{uu}^T \mathbf{n} & \mathbf{x}_{uv}^T \mathbf{n} \\ \mathbf{x}_{vu}^T \mathbf{n} & \mathbf{x}_{vv}^T \mathbf{n} \end{bmatrix}$$

Energy Models [Botsch & Kobbelt, SIGGRAPH 04]

■ Nonlinear stretching & bending energies

$$\int_{\Omega} k_s \|\mathbf{I} - \mathbf{I}'\|^2 + k_b \|\mathbf{I} - \mathbf{I}'\|^2 du dv$$
stretching bending

■ Linearize terms → Quadratic energy

$$\int_{\Omega} k_s \underbrace{\left(\|\mathbf{d}_u\|^2 + \|\mathbf{d}_v\|^2\right)} + k_b \underbrace{\left(\|\mathbf{d}_{uu}\|^2 + 2\|\mathbf{d}_{uv}\|^2 + \|\mathbf{d}_{vv}\|^2\right)}_{\text{bending}} \mathrm{d}u \mathrm{d}v$$
stretching bending

Energy Models [Botsch & Kobbelt, SIGGRAPH 04]

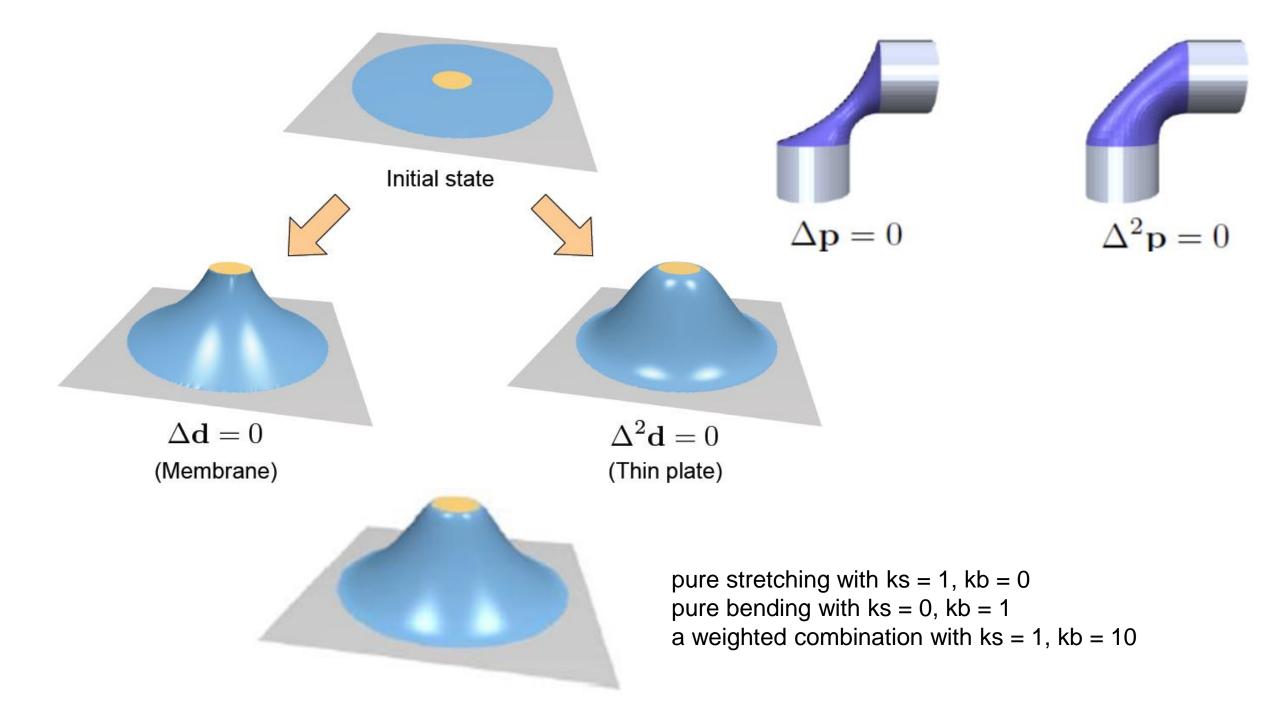
■ Minimize linearized bending energy:

$$E(\mathbf{d}) = \int_{\mathcal{S}} \|\mathbf{d}_{uu}\|^2 + 2\|\mathbf{d}_{uv}\|^2 + \|\mathbf{d}_{vv}\|^2 d\mathcal{S} \quad f(x) \to \min$$

■ Variational calculus → Euler-Lagrange PDE

$$\Delta^2 \mathbf{d} := \mathbf{d}_{uuuu} + 2\mathbf{d}_{uuvv} + \mathbf{d}_{vvvv} = 0 \qquad \left[f'(x) = 0 \right]$$

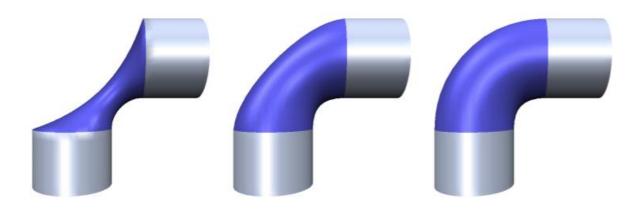
"Best" deformation that satisfies constraints



K = 1 & 2 is not enough

$$E_k(d) = \int_{\Omega} \|\nabla^k d\|^2 du dv$$

$$\int_{\mathbb{R}^3} \|\mathbf{d}_{xxx}\|^2 + \|\mathbf{d}_{xyy}\|^2 + \ldots + \|\mathbf{d}_{zzz}\|^2 \, dx \, dy \, dz$$



Minimizing $E_k => C_{k-1}$ blending

membrane surface (k = 1), thin-plate surface (k = 2), minimal curvature variation (k = 3).

[Botsch & Kobbelt, SIGGRAPH 04]

$$min E_k(d(x)) \Rightarrow \Delta^k d(x) = 0, x \in \Omega \setminus \delta\Omega$$

s.t. $\Delta^j d(x) = b_j(x), x \in \delta\Omega, j < k$

$$\qquad \qquad \left(\begin{array}{c|c} \bar{\Delta}^k \\ \hline 0 & I_{F+H} \end{array} \right) \left(\begin{array}{c} \mathbf{p} \\ \mathbf{f} \\ \mathbf{h} \end{array} \right) = \left(\begin{array}{c} \mathbf{0} \\ \mathbf{f} \\ \mathbf{h} \end{array} \right)$$

PDE Discretization

Euler-Lagrange PDE

Laplace discretization

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

$$\mathbf{d} = \mathbf{0}$$

$$\mathbf{d} = \delta \mathbf{h}$$

$$\Delta \mathbf{d}_{i} = \frac{1}{2A_{i}} \sum_{j \in \mathcal{N}_{i}} (\cot \alpha_{ij} + \cot \beta_{ij}) (\mathbf{d}_{j} - \mathbf{d}_{i})$$

$$\Delta^{2} \mathbf{d}_{i} = \Delta(\Delta \mathbf{d}_{i})$$

$$\mathbf{x}_{j}$$

Linear System

- Sparse linear system
 - Turn into symmetric positive definite system

$$\begin{pmatrix} \Delta^2 \\ 0 & \mathbf{I} & 0 \\ 0 & 0 & \mathbf{I} \end{pmatrix} \begin{pmatrix} \vdots \\ \mathbf{d}_i \\ \vdots \end{pmatrix} = \begin{pmatrix} 0 \\ \delta \mathbf{h}_i \end{pmatrix}$$

- Solve this system each frame
 - Use efficient linear solvers !!!
 - Sparse Cholesky factorization

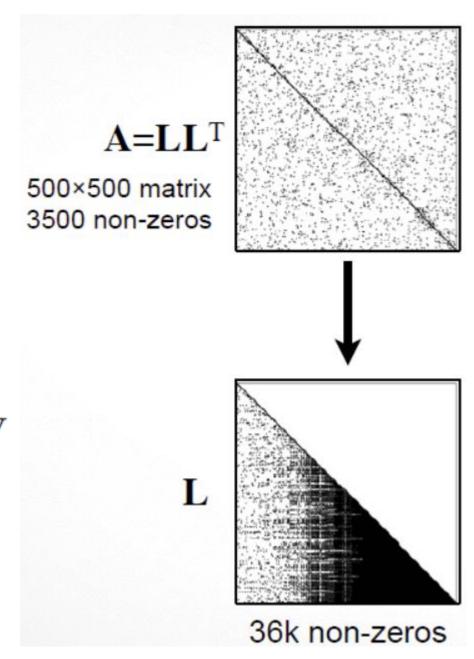
Sparse SPD Solvers

- Dense Cholesky factorization
 - Cubic complexity
 - High memory consumption (doesn't exploit sparsity)
- Iterative conjugate gradients
 - Quadratic complexity
 - Need sophisticated preconditioning
- Multigrid solvers
 - Linear complexity
 - But rather complicated to develop (and to use)
- Sparse Cholesky factorization?

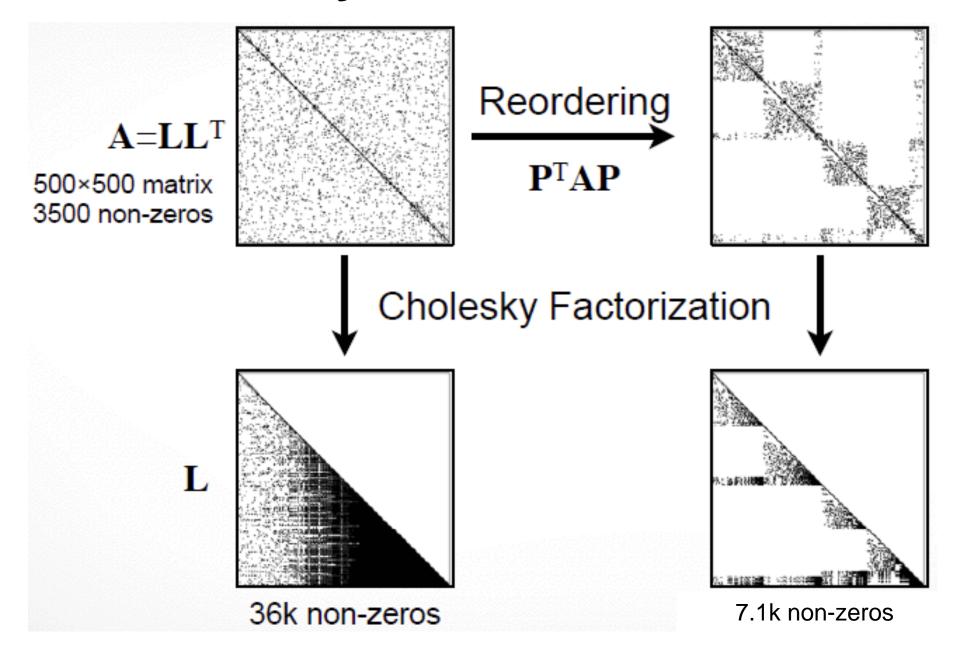
Dense Cholesky Solver

Solve Ax = b

- 1. Cholesky factorization $\mathbf{A} = \mathbf{L}\mathbf{L}^T$
- 2. Solve system $y = L^{-1}b$, $x = L^{-T}y$



Sparse Cholesky Factorization



Sparse Cholesky Solver

Solve
$$Ax = b$$

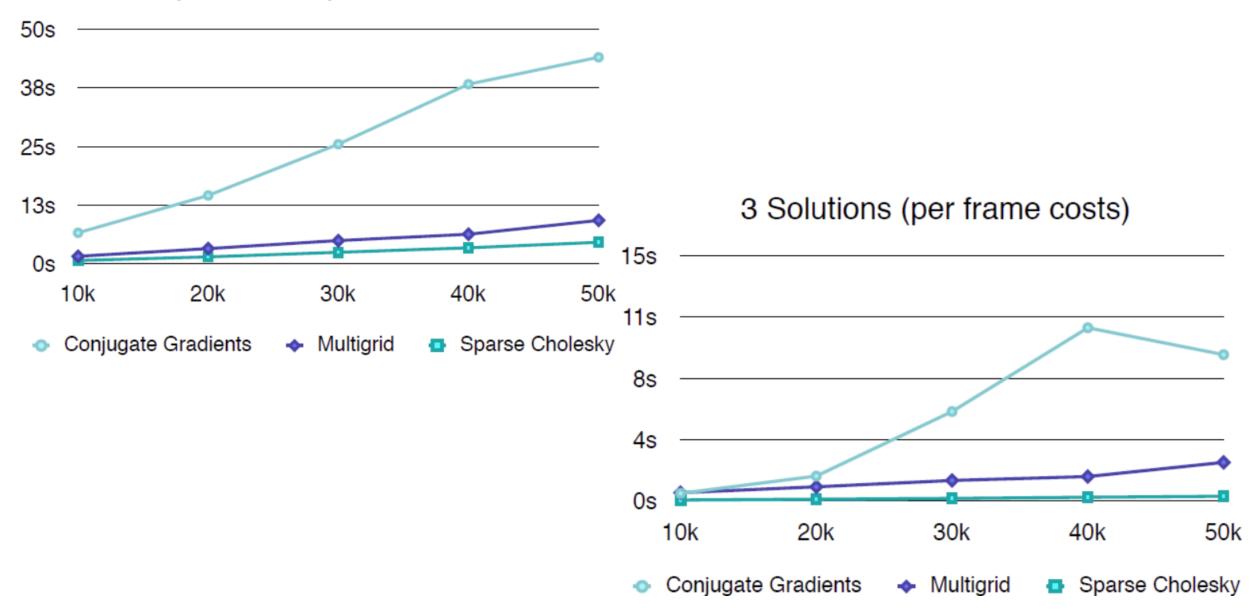
Pre-computation

- 1. Matrix re-ordering $\tilde{\mathbf{A}} = \mathbf{P}^T \mathbf{A} \mathbf{P}$
- 2. Cholesky factorization $\tilde{\mathbf{A}} = \mathbf{L}\mathbf{L}^T$
- 3. Solve system $\mathbf{y} = \mathbf{L}^{-1}\mathbf{P}^T\mathbf{b}$, $\mathbf{x} = \mathbf{P}\mathbf{L}^{-T}\mathbf{y}$

Per-frame computation

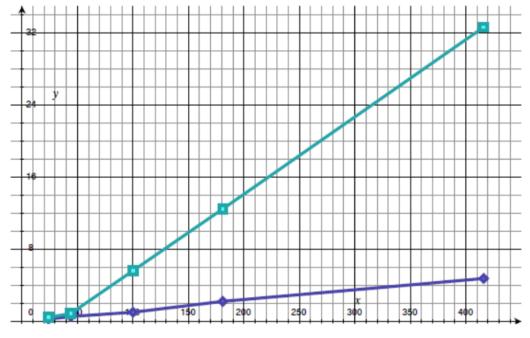
Bi-Laplace Systems

Setup + Precomp. + 3 Solutions



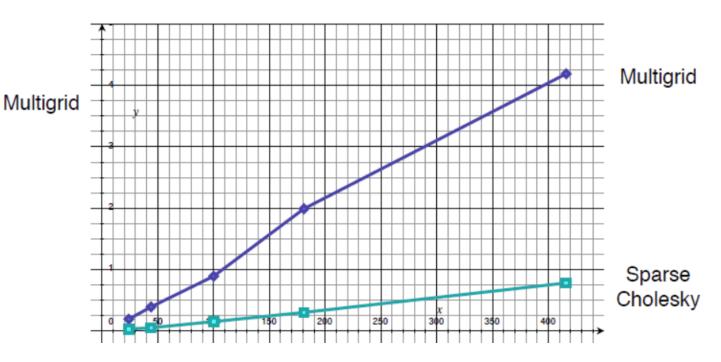
Laplace Systems

Setup + Precomp. + 3 Solutions



Sparse Cholesky

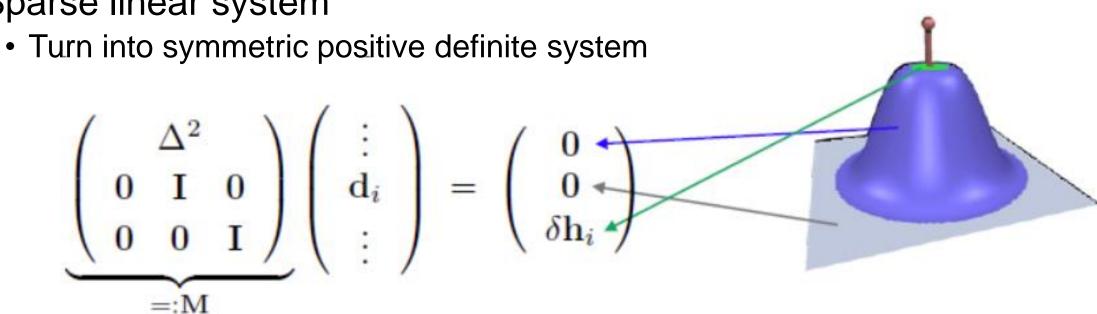
3 Solutions (per frame costs)



[Shi et al, SIGGRAPH 06]

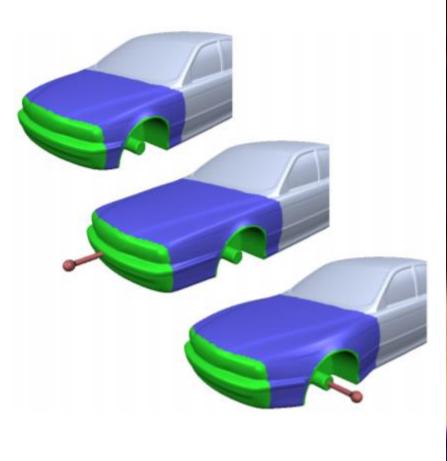
Linear System

Sparse linear system



- Can be turned into symm. pos. def. system
 - Right hand sides changes each frame
 - Sparse Cholesky factorization
 - Very efficient implementations publicly available

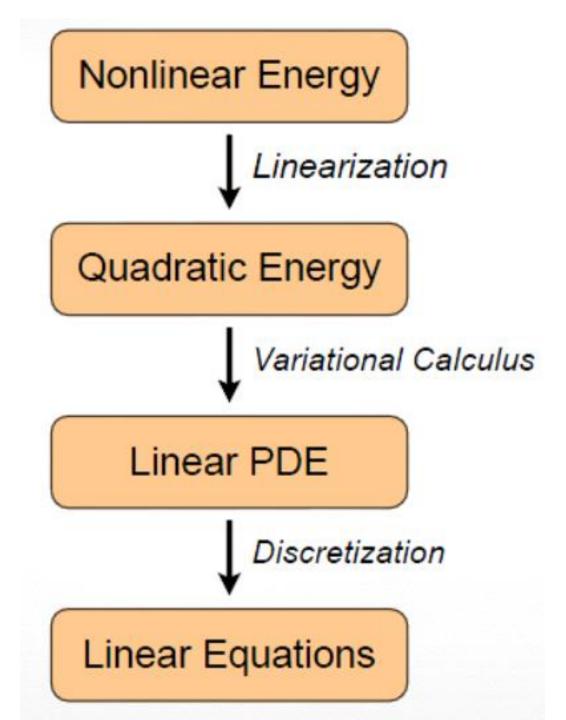
CAD-Like Deformation & Facial Animation





[Botsch & Kobbelt, SIGGRAPH 04]

Derivation Steps

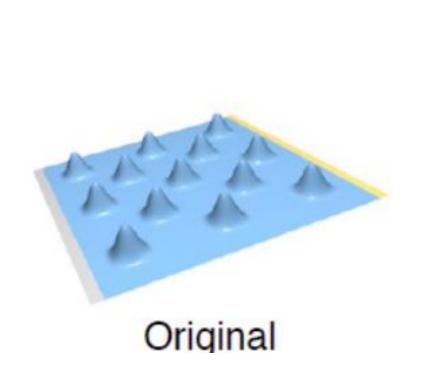


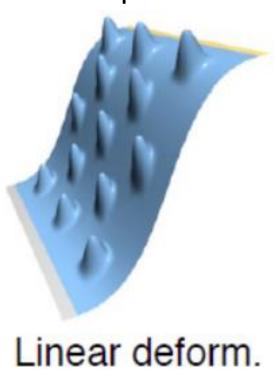
Overview

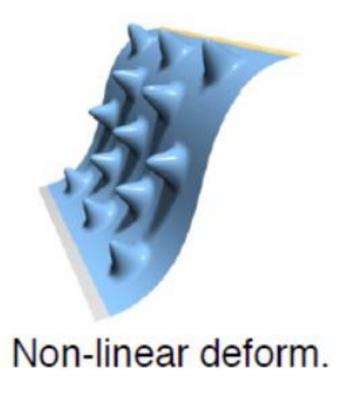
- Surface-based deformation
 - Energy minimization
 - Multiresolution editing
 - Differential coordinates

Multiresolution Modeling

- Even pure translations induce local rotations!
 - → Inherently non-linear coupling
- Alternative approach
 - Linear deformation + multi-scale decomposition...





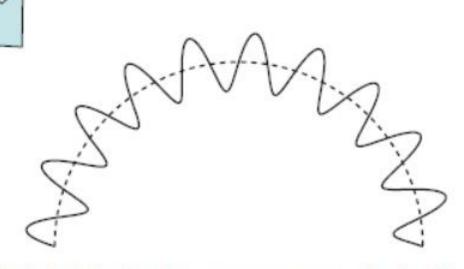


Multiresolution Editing



Frequency decomposition

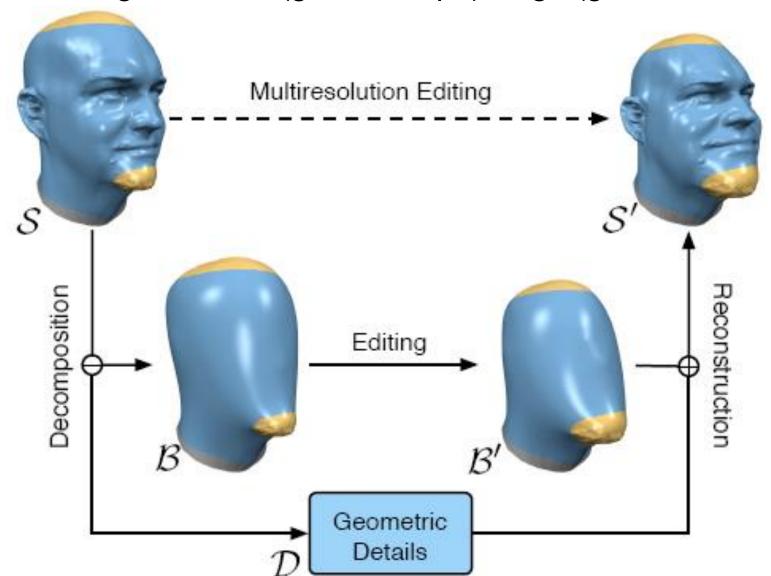
Change low frequencies



Add high frequency details, stored in local frames

Multiresolution Editing [Michael Garland el at 99]

■ Geometric signal – Low (global shape) /High (geometric details)



3 operators:

- Decomposition
- Deformation
- reconstruction

Displacement Vectors

- Decomposition: $\mathbf{p}_i = \mathbf{b}_i + \mathbf{h}_i, \quad \mathbf{h}_i \in \mathbb{R}^3,$
 - · Represent hi via global vs local frame

$$\mathbf{h}_i = \alpha_i \, \mathbf{n}_i + \beta_i \, \mathbf{t}_{i,1} + \gamma_i \, \mathbf{t}_{i,2}.$$

Reconstruction:

$$\mathbf{p}'_{i} = \mathbf{b}'_{i} + \alpha_{i} \mathbf{n}'_{i} + \beta_{i} \mathbf{t}'_{i,1} + \gamma_{i} \mathbf{t}'_{i,2}.$$

Choose t i heuristicaly



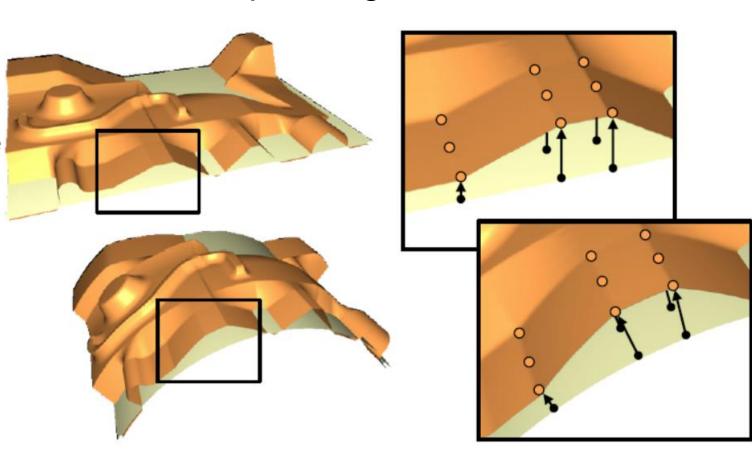
Store hi via global vs local frame

Normal Displacements

- long displacement vectors might lead to instabilities, in particular for bending deformations
- Displacement vectors should connect vertices pi of S to their closest surface points on B instead of their corresponding vertices bi of B.

$$\mathbf{p}_i = \mathbf{b}_i + h_i \cdot \mathbf{n}_i, \quad h_i \in \mathbb{R}.$$

- How to compute bi
- Resampling may introduce alias artifacts
- Local Newton iteration



Normal Displacements – compute bi via Newton iteration

find the barycentric coordinates of the base point bi as the root of the function

$$f(\alpha, \beta, \gamma) = (\mathbf{p}_i - \mathbf{b}_i) \times \mathbf{n}_i$$

• where

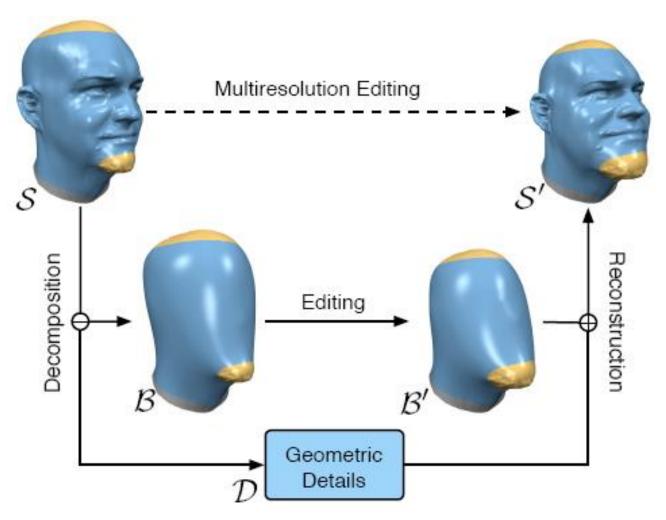
$$\mathbf{b}_{i} = \alpha \mathbf{a} + \beta \mathbf{b} + \gamma \mathbf{c}.$$

$$\mathbf{n}_{i} = \frac{\alpha \mathbf{n}_{a} + \beta \mathbf{n}_{b} + \gamma \mathbf{n}_{c}}{\|\alpha \mathbf{n}_{a} + \beta \mathbf{n}_{b} + \gamma \mathbf{n}_{c}\|}.$$

- The process is **initialized** with the triangle closest to pi. If a barycentric coordinate becomes negative during the Newton iteration, one proceeds to the respective neighboring triangle.
- Then graph of S and B is no longer restricted to be identical, which can be exploited to remesh B for the sake of higher numerical robustness.

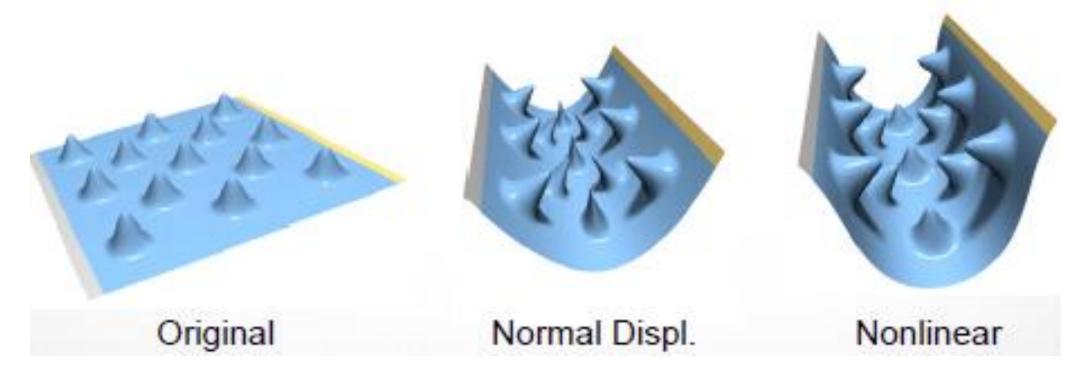
Multiresolution Editing [Michael Garland el at 99]

 the general displacements are in average about 9 times longer than normal displacements



Limitations

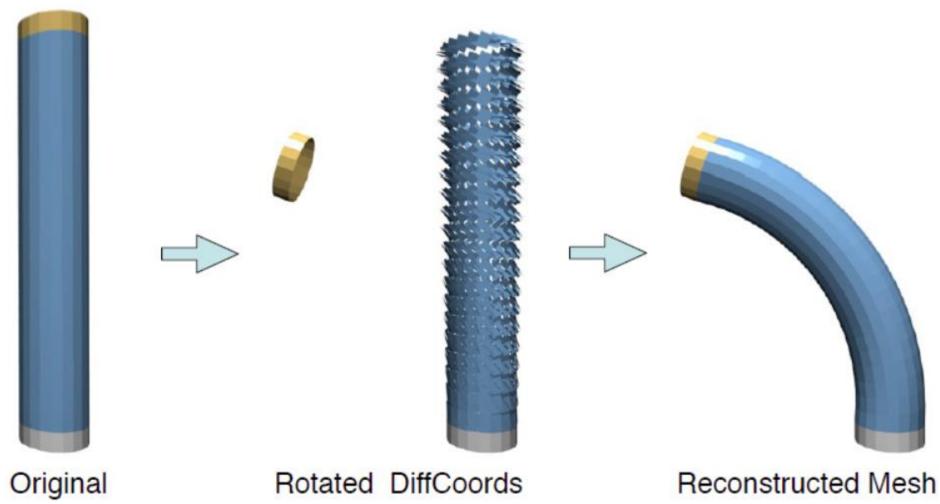
- Neighboring displacements are not coupled
 - Surface bending changes the angle between neighboring displacement vectors
 - Leads to volume changes or self-intersections
- Multiresolution hierarchy difficult to compute
 - Complex topology or Complex geometry might require more hierarchy levels



Differential Coordinates

- Manipulate differential coordinates instead of spatial coordinates
 - Gradients, Laplacians, local frames
 - Intuition: Close connection to surface normal
- Find mesh with desired differential coords
 - Formulate as energy minimization

Using gradient-based editing to bend the cylinder by 90



Rotating the handle and propagating its **damped local rotation** to the individual triangles (resp. their gradients JT) breaks up the mesh (center),

but solving the **Poisson system** reconnects it and yields the desired result (right).

Gradient-Based Editing

Manipulate gradient field of a function (surface)

$$\mathbf{g} = \nabla \mathbf{p} \qquad \mathbf{g} \mapsto \mathbf{g}'$$

Find function f' whose gradient is (close to) g'

$$\int_{\mathcal{S}} \|\nabla \mathbf{p}' - \mathbf{g}'\|^2 \, d\mathcal{S} \to \min$$

Variational calculus yields Euler-Lagrange PDE

$$\Delta \mathbf{p}' = \operatorname{div} \mathbf{g}'$$

Gradient-Based Editing

Use piecewise linear coordinate function

$$\mathbf{p}(u,v) = \sum_{v_i} \mathbf{p}_i \cdot \phi_i(u,v)$$

Its gradient is

$$\nabla \mathbf{p}(u, v) = \sum_{v_i} \mathbf{p}_i \cdot \nabla \phi_i(u, v)$$

$$\downarrow^{1} \qquad \phi_1 \qquad \phi_2 \qquad \phi_3 \qquad \downarrow^{1} \qquad \downarrow^{1$$

Gradient-Based Editing
$$\nabla \mathbf{p}|_T = \begin{bmatrix} \nabla \mathbf{p}_x|_T \\ \nabla \mathbf{p}_y|_T \\ \nabla \mathbf{p}_z|_T \end{bmatrix} =: \mathbf{J}_T \in \mathbb{R}^{3 \times 3}$$
 • Constant per triangle
$$\nabla \mathbf{p}|_{f_j} =: \mathbf{G}_j \in \mathbb{R}^{3 \times 3}$$

$$\mathbf{G}_j =: \mathbf{J}_T$$

$$\left(egin{array}{c} \mathbf{G}_1 \ dots \ \mathbf{G}_F \end{array}
ight) \ = \underbrace{\mathbf{G}}_{\in \mathrm{I\!R}^{3F imes V}} \cdot \left(egin{array}{c} \mathbf{p}_1^T \ dots \ \mathbf{p}_V^T \end{array}
ight)$$

Manipulate per-face gradients

$$\mathbf{G}_j \mapsto \mathbf{G}_j'$$

$$\mathbf{J}_T' = \mathbf{M}_T \mathbf{J}_T$$

Gradient-Based Editing

- Reconstruct mesh from changed gradients
 - Overdetermined problem $\mathbf{G} \in \mathbb{R}^{3F \times V}$
 - Weighted least squares system
 - Linear Poisson (Laplace) system

$$\mathbf{G}^{T}\mathbf{D}\mathbf{G} \cdot \begin{pmatrix} \mathbf{p}_{1}^{\prime T} \\ \vdots \\ \mathbf{p}_{V}^{\prime T} \end{pmatrix} = \mathbf{G}^{T}\mathbf{D} \cdot \begin{pmatrix} \mathbf{G}_{1}^{\prime} \\ \vdots \\ \mathbf{G}_{F}^{\prime} \end{pmatrix}$$

$$\operatorname{div} \nabla = \Delta \begin{pmatrix} \mathbf{p}_{1}^{\prime T} \\ \vdots \\ \mathbf{G}_{F}^{\prime T} \end{pmatrix}$$

Laplacian-Based Editing

Manipulate Laplacians of a surface

$$\delta_i = \Delta(\mathbf{p}_i) \ , \quad \delta_i \mapsto \delta_i'$$

Find surface whose Laplacian is (close to) δ'

$$\int_{\mathcal{S}} \left\| \Delta \mathbf{p}' - \boldsymbol{\delta}' \right\|^2 d\mathcal{S} \rightarrow \min$$

Variational calculus yields Euler-Lagrange PDE

$$\Delta^2 \mathbf{p}' = \Delta \delta'$$

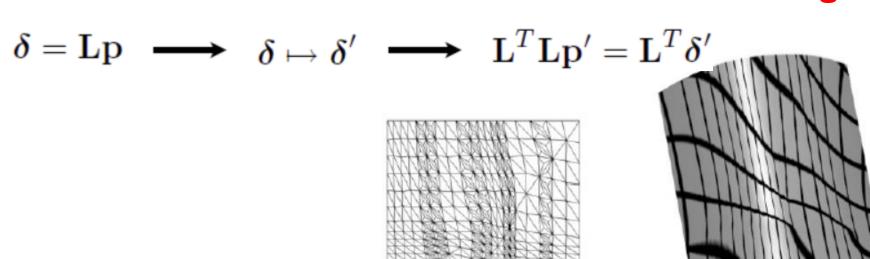
Discretization

Discretize Euler-Lagrange PDE

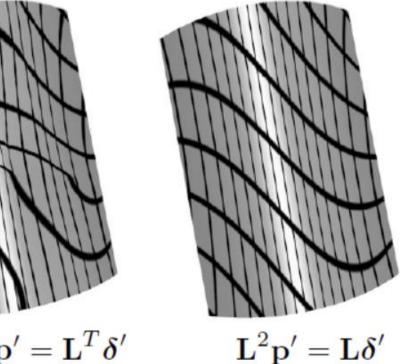
$$\Delta^2 \mathbf{p}' = \Delta \delta' \longrightarrow \mathbf{L}^2 \mathbf{p}' = \mathbf{L} \delta'$$

Frequently used (wrong) version

Wrong



Irregular mesh



 $\mathbf{L}^T \mathbf{L} \mathbf{p}' = \mathbf{L}^T \boldsymbol{\delta}'$

Connection to Plate Energy?

Neglect change of Laplacians for a moment...

$$\int \|\Delta \mathbf{p}' - \boldsymbol{\delta}\|^2 \to \min \qquad \longrightarrow \qquad \Delta^2 \mathbf{p}' = \Delta \boldsymbol{\delta}$$

- Basic formulations equivalent!
- Differ in detail preservation
 - Rotation of Laplacians
 - Multi-scale decomposition

$$\int \|\mathbf{d}_{uu}\|^2 + 2\|\mathbf{d}_{uv}\|^2 + \|\mathbf{d}_{vv}\|^2 \to \min \quad \longleftarrow \quad \Delta^2 \mathbf{d} = 0$$

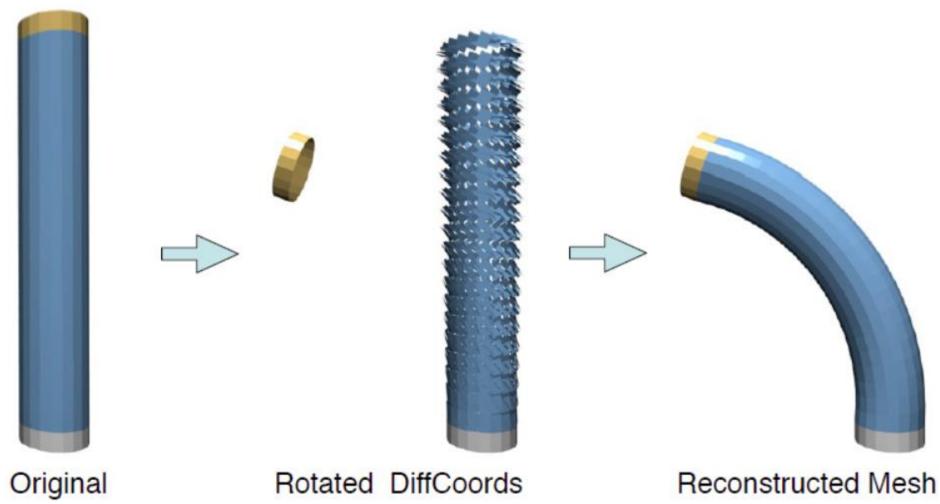
$$\Delta^{2}\mathbf{p}' = \Delta\delta$$

$$\mathbf{p}' = \mathbf{p} + \mathbf{d}$$

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

$$\Delta^{2}(\mathbf{p} + \mathbf{d}) = \Delta^{2}\mathbf{p}$$

Using gradient-based editing to bend the cylinder by 90



Rotating the handle and propagating its **damped local rotation** to the individual triangles (resp. their gradients JT) breaks up the mesh (center),

but solving the **Poisson system** reconnects it and yields the desired result (right).

Differential Coordinates

- Which differential coordinate δi?
 - Gradients
 - Laplacians
 - •
- How to get local transformations $T_i(\delta_i)$?
 - Smooth propagation
 - Implicit optimization
 - ...

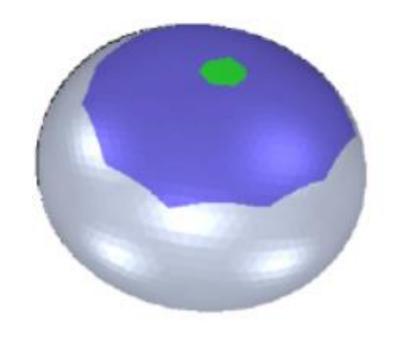
Construct Scalar Field

Construct smooth scalar field [0,1]

s(x)=1: Full deformation (handle)

s(x)=0: No deformation (fixed part)

s(x)∈(0,1): Damp handle transformation (in between)

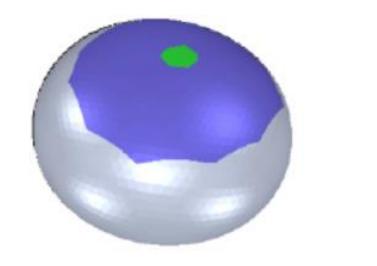




Construct Scalar Field

Construct a smooth harmonic field

$$\begin{aligned} &-\text{ Solve } & \Delta(s) = 0 \\ &-\text{ with } & s\left(\mathbf{p}\right) = \left\{ \begin{array}{ll} 1 & \mathbf{p} \in \text{ handle} \\ 0 & \mathbf{p} \in \text{ fixed} \end{array} \right. \end{aligned}$$





Damp Handle Transformation

Full handle transformation

- Rotation: $R(\mathbf{c}, \mathbf{a}, \alpha)$

– Scaling: S(s)

Damped by scalar λ

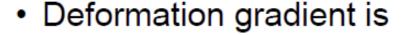
- Rotation: $R(\mathbf{c}, \mathbf{a}, \lambda \cdot \alpha)$

- Scaling: $S(\lambda \cdot s + (1-\lambda) \cdot 1)$

Damp Handle Transformation

Handle has been transformed <u>affinely</u>

$$T(x) = Ax + t$$



$$\nabla \mathbf{T}(\mathbf{x}) = \mathbf{A}$$

Extract rotation R and scale/shear S

$$\mathbf{A} = \mathbf{U} \mathbf{\Sigma} \mathbf{V}^T \quad \Rightarrow \quad \mathbf{R} = \mathbf{U} \mathbf{V}^T, \ \mathbf{S} = \mathbf{V} \mathbf{\Sigma} \mathbf{V}^T$$

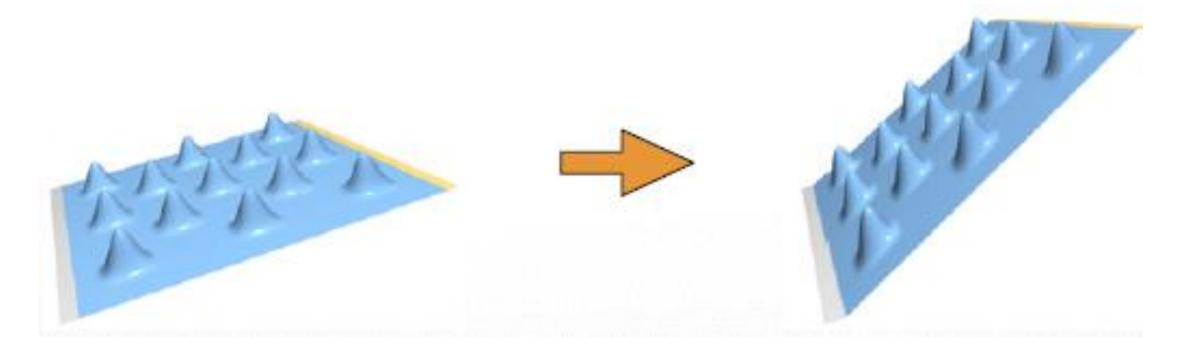
polar decomposition

$$\mathbf{RS} = \mathbf{U}\mathbf{V}^T\mathbf{V}\mathbf{\Sigma}\mathbf{V}^T = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T = \mathbf{M}.$$

$$\mathbf{M}_i = \operatorname{slerp}(\mathbf{R}, \mathbf{Id}, s_i) \cdot ((1 - s_i)\mathbf{S} + s_i\mathbf{Id}),$$

Limitations

- Differential coordinates work well for rotations
 - Represented by deformation gradient
- Translations don't change deformation gradient
 - Translations don't change differential coordinates
 - "Translation insensitivity"



Implicit Optimization

• Simultaneously optimize: new vertex positions p' & local rotations Mi

$$E(\mathbf{p}') = \sum_{i=1}^{n} A_i \|\mathbf{M}_i \, \boldsymbol{\delta}_i - \Delta \mathbf{p}_i'\|^2$$

 Local transformations are restricted to linearized similarity transformations

$$\mathbf{M}_i = egin{bmatrix} s_i & -h_{i,z} & h_{i,y} \ h_{i,z} & s_i & -h_{i,x} \ -h_{i,y} & h_{i,x} & s_i \end{bmatrix}$$

Extracting (si; hi) as linear combinations of p'i.

$$\mathbf{M}_i(\mathbf{p}_i - \mathbf{p}_j) = \mathbf{p}'_i - \mathbf{p}'_j, \quad \forall \mathbf{p}_j \in \mathcal{N}_1(\mathbf{p}_i)$$

Thanks