

Laplacian Mesh Processing

(includes material from Olga Sorkine, Yaron Lipman, Marc Pauly, Adrien Treuille, Marc Alexa and Daniel Cohen-Or)

Recap: Laplacian in Euclidean space

- **Laplacian** (of **scalar**-valued function):
 - In operator form:
$$\Delta = \left(\frac{\partial^2}{\partial x_1^2}, \frac{\partial^2}{\partial x_2^2}, \dots, \frac{\partial^2}{\partial x_n^2} \right)$$
 - Maps **scalar** field to **scalar** field
- $\Delta f = \nabla \cdot \nabla f = \frac{\partial^2 f}{\partial x_1^2} + \frac{\partial^2 f}{\partial x_2^2} + \dots + \frac{\partial^2 f}{\partial x_n^2}$
- Divergence of gradient of f



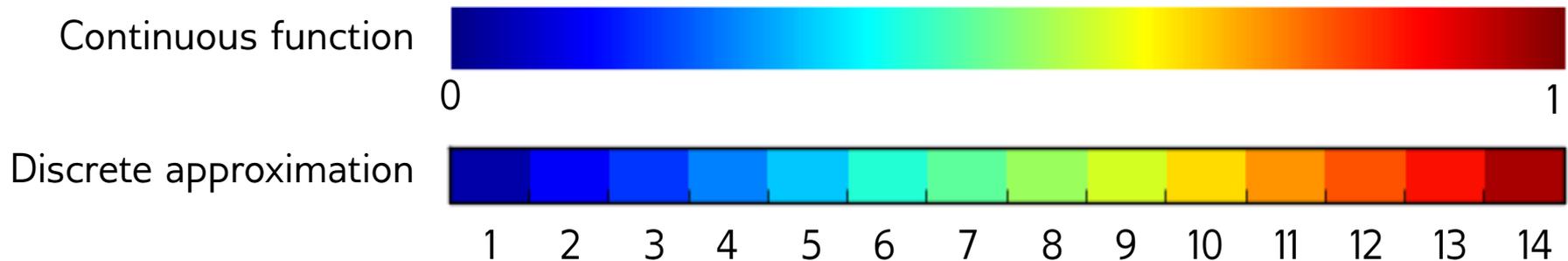
Original function



After applying Laplacian

Recap: Laplacian in Euclidean space

- The Laplacian can be discretized



$$\frac{1}{h} \begin{bmatrix} -1 & 1 & 0 & \dots & 0 & 0 \\ 0 & -1 & 1 & \dots & 0 & 0 \\ & 0 & -1 & \dots & 0 & 0 \\ \vdots & & & \ddots & & \vdots \\ 0 & 0 & & & -1 & 1 \\ 0 & 0 & \dots & & 0 & -1 \end{bmatrix}$$

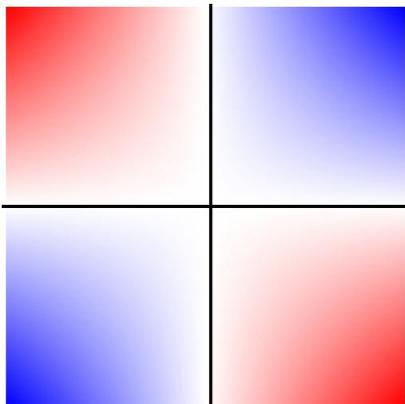
1D discrete gradient

$$\frac{1}{h^2} \begin{bmatrix} -2 & 1 & 0 & \dots & 0 & 0 \\ 1 & -2 & 1 & \dots & 0 & 0 \\ & 1 & -2 & \dots & 0 & 0 \\ \vdots & & & \ddots & & \vdots \\ 0 & 0 & & & -2 & 1 \\ 0 & 0 & \dots & & 1 & -2 \end{bmatrix}$$

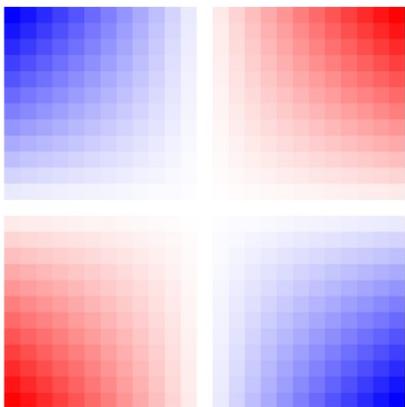
1D discrete Laplace operator

Laplacian in 2D Euclidean space

- The Laplacian can be discretized



Continuous function



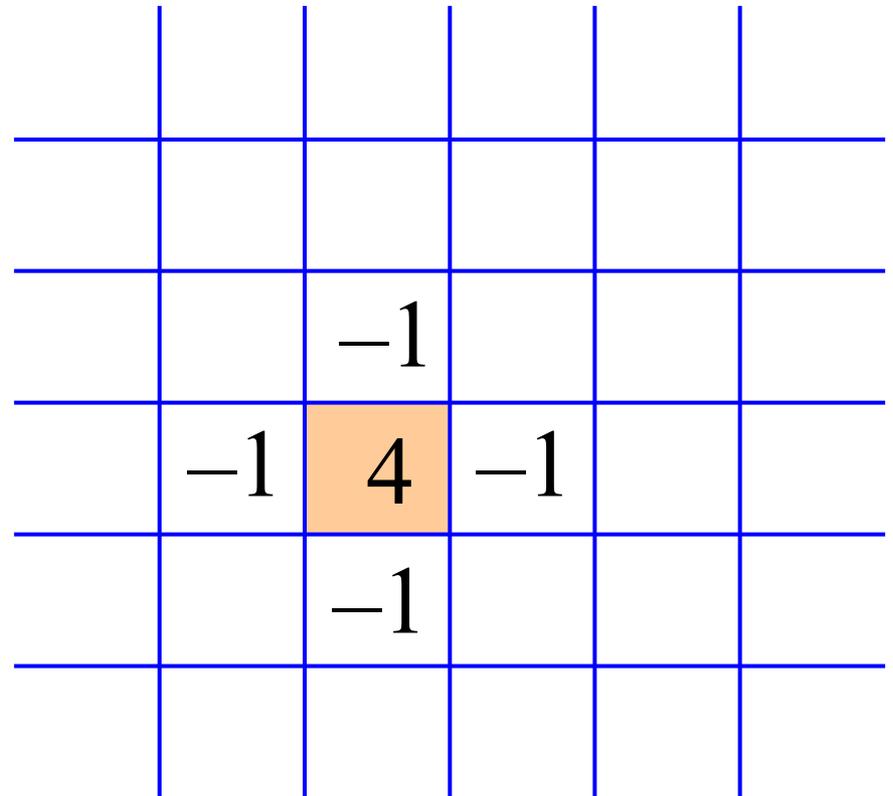
Discrete approximation

$$\begin{bmatrix} -4 & 1 & \cdot & 1 & \cdot & \cdot & \cdot & \cdot & \cdot \\ 1 & -4 & 1 & \cdot & 1 & \cdot & \cdot & \cdot & \cdot \\ \cdot & 1 & -4 & \cdot & \cdot & 1 & \cdot & \cdot & \cdot \\ 1 & \cdot & \cdot & -4 & 1 & \cdot & 1 & \cdot & \cdot \\ \cdot & 1 & \cdot & 1 & -4 & 1 & \cdot & 1 & \cdot \\ \cdot & \cdot & 1 & \cdot & 1 & -4 & \cdot & \cdot & 1 \\ \cdot & \cdot & \cdot & 1 & \cdot & \cdot & -4 & 1 & \cdot \\ \cdot & \cdot & \cdot & \cdot & 1 & \cdot & 1 & -4 & 1 \\ \cdot & \cdot & \cdot & \cdot & \cdot & 1 & \cdot & 1 & -4 \end{bmatrix}$$

2D discrete Laplace operator

Laplacian in 2D Euclidean space

- The Laplacian is computed via differences of a cell from its neighbors
- We can think of the grid as a graph connecting adjacent cells



Thought for the Day #1

Can you express applying the Laplacian (or other differential operator) as a convolution?

Graph Laplacian

- For a general graph, we can compute a similar Laplace operator
 - The function f is represented by its values at graph vertices
 - The discrete Laplace operator is applied on graph neighborhoods centred at the vertices
 - If the graph is a grid, we should recover the standard Euclidean Laplacian

Graph Laplacian

- Let $G = (V, E)$ be an undirected graph
- Let $N_1(i)$ represent the 1-ring neighborhood of vertex i , that is, $N_1(i) = \{j \mid (i, j) \in E\}$
- The degree d_i of vertex i is $|N_1(i)|$
- Let D be the diagonal matrix with $D_{ii} = d_i$
- The adjacency matrix A of G is $A_{ij} = \begin{cases} 1 & (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$

Graph Laplacian

- The (topological) **Laplacian matrix** L of the graph is defined as

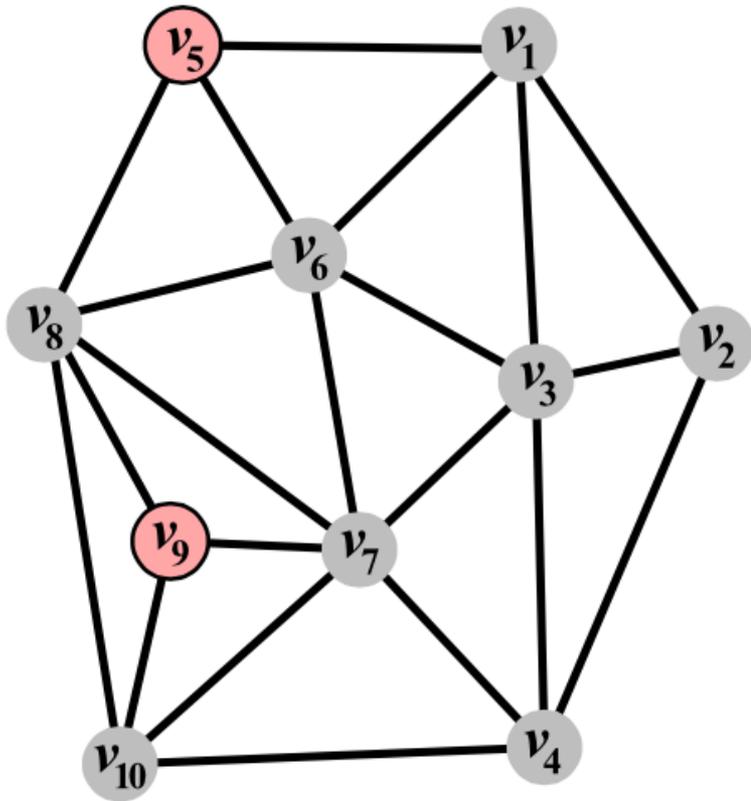
$$L = I - D^{-1}A$$

- Commonly, we can multiply by D and consider the **symmetric Laplacian** $L_s = DL = D - A$ instead

$$(L_s)_{ij} = \begin{cases} d_i & i = j \\ -1 & (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$

Verify that this gives the correct grid Laplacian

Graph Laplacian



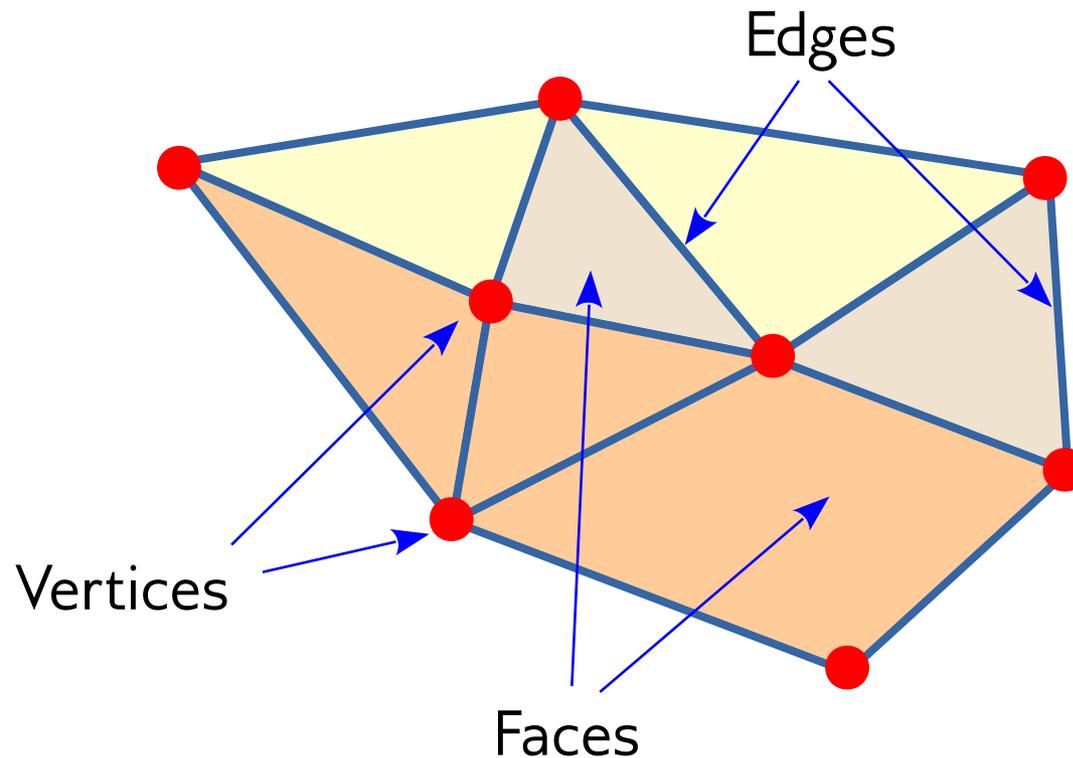
Graph

4	-1	-1		-1	-1				
-1	3	-1	-1						
-1	-1	5	-1		-1	-1			
	-1	-1	4			-1			-1
-1				3	-1		-1		
-1		-1			4	-1	-1		
		-1	-1		-1	6	-1	-1	-1
				-1	-1	-1	6	-1	-1
						-1	-1	3	-1
			-1			-1	-1	-1	4

Symmetric Laplacian L_s

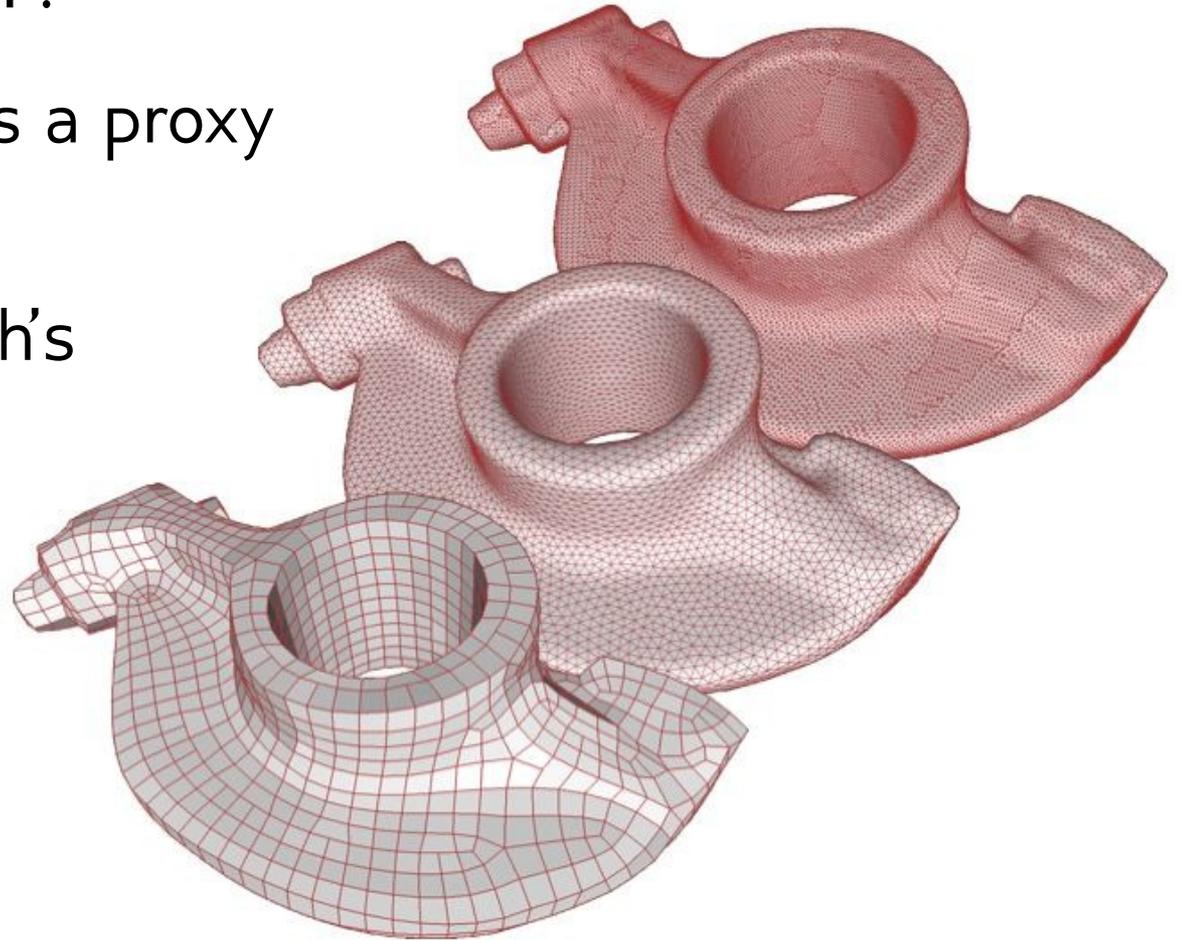
Mesh (Topological) Laplacian

- **Recall:** a mesh is a graph
- ... so we can compute its topological Laplacian



Refining the mesh

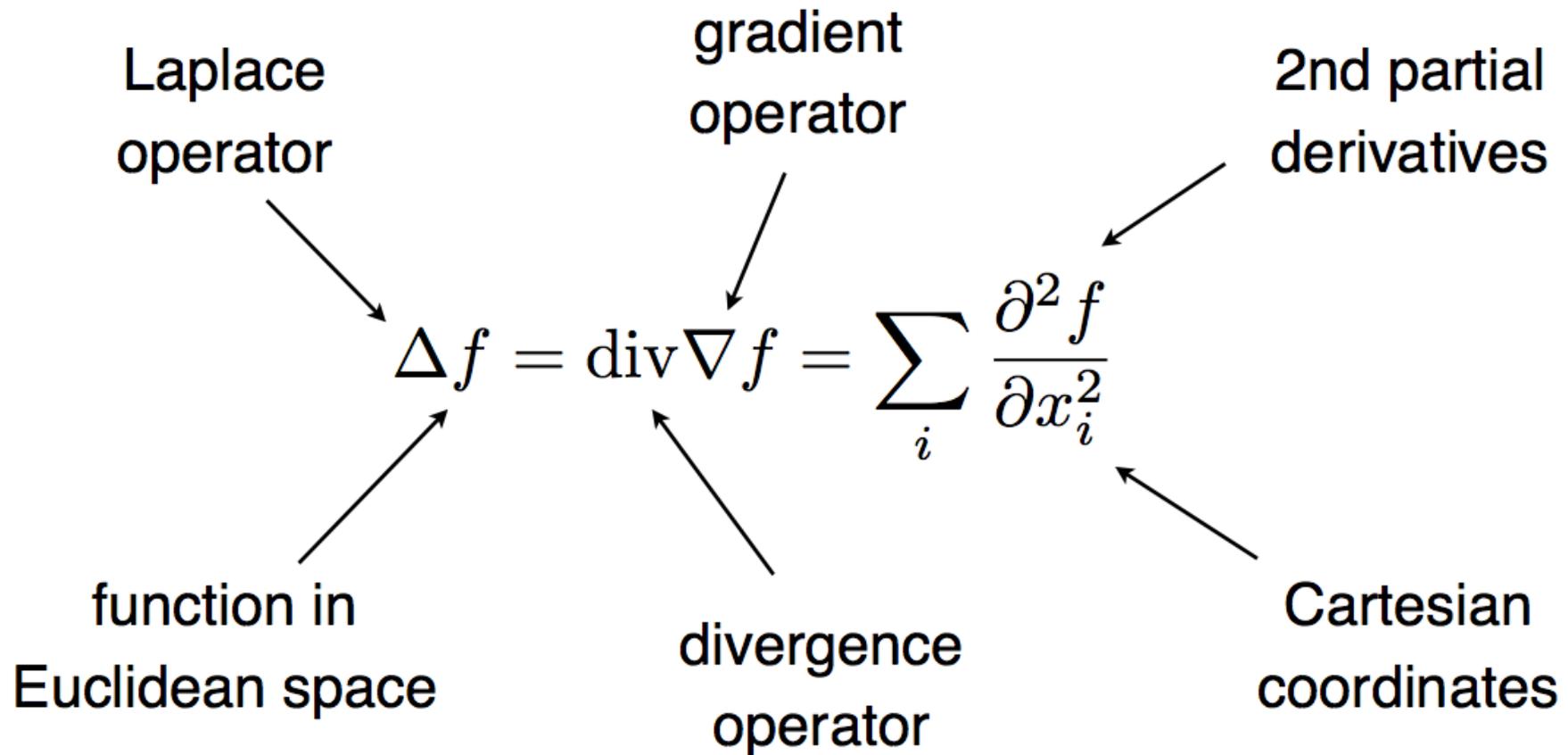
- What happens to the Laplacian as we make the mesh finer and finer?
 - The graph serves as a proxy for the surface
 - Eventually, the graph's metric properties should reduce to geometric surface properties



Thought for the Day #2

The *topological* Laplacian does not care about the embedding metric, i.e. the lengths of edges in the graph. How, then, can the Laplacian be sensitive to the *geometry* of the surface?

Once again: the Euclidean Laplacian



The Laplace-Beltrami operator

- Extension of continuous Laplacian to manifolds (“smooth curved surfaces/spaces”)

The diagram illustrates the equation for the Laplace-Beltrami operator on a manifold \mathcal{S} . The equation is $\Delta_{\mathcal{S}} f = \text{div}_{\mathcal{S}} \nabla_{\mathcal{S}} f$. Four arrows point from descriptive text to the components of the equation: 'Laplace-Beltrami' points to $\Delta_{\mathcal{S}}$, 'function on manifold \mathcal{S} ' points to f , 'divergence operator' points to $\text{div}_{\mathcal{S}}$, and 'gradient operator' points to $\nabla_{\mathcal{S}}$.

$$\Delta_{\mathcal{S}} f = \text{div}_{\mathcal{S}} \nabla_{\mathcal{S}} f$$

Laplace-Beltrami

gradient operator

function on manifold \mathcal{S}

divergence operator

The Laplace-Beltrami operator

- **Example:** Let \mathbf{x} be the coordinate function:
 $\mathbf{x}(p) = \text{position of } p$

Laplace-Beltrami

coordinate function

divergence operator

gradient operator

mean curvature

surface normal

$$\Delta_S \mathbf{x} = \text{div}_S \nabla_S \mathbf{x} = -2H\mathbf{n}$$

Here, applying the Laplacian to a vector-valued function means applying it to each coordinate individually

The Laplace-Beltrami operator

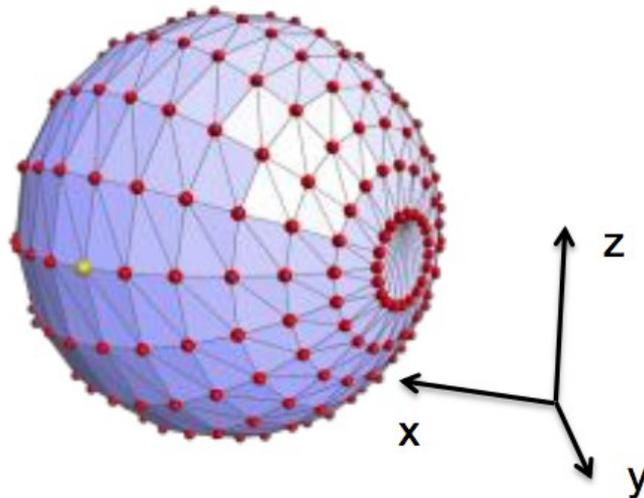
- Just like in 2D Euclidean space, if we know the Laplace-Beltrami operator of the surface, and the function value at a single point, we can solve for the function at all points on the surface
- Why is this important?
 - The LB operator captures global properties of the surface in terms of local properties
 - It expresses a local invariant
 - E.g. for the coordinate function, it expresses the surface in terms of its local curvature and orientation
 - We can modify the properties locally in one small region and recompute the global properties, trying to preserve the local property everywhere else

The Laplace-Beltrami operator

- The topological Laplacian is an approximation to the discrete Laplace-Beltrami operator
- We shall see better approximations soon

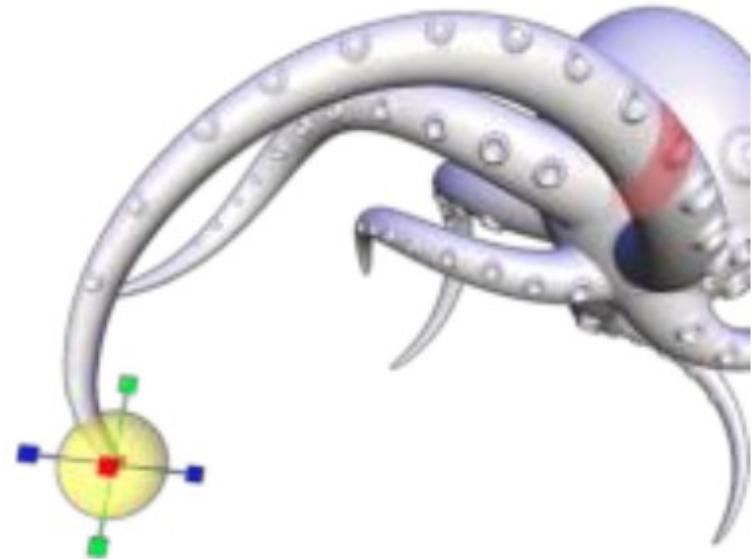
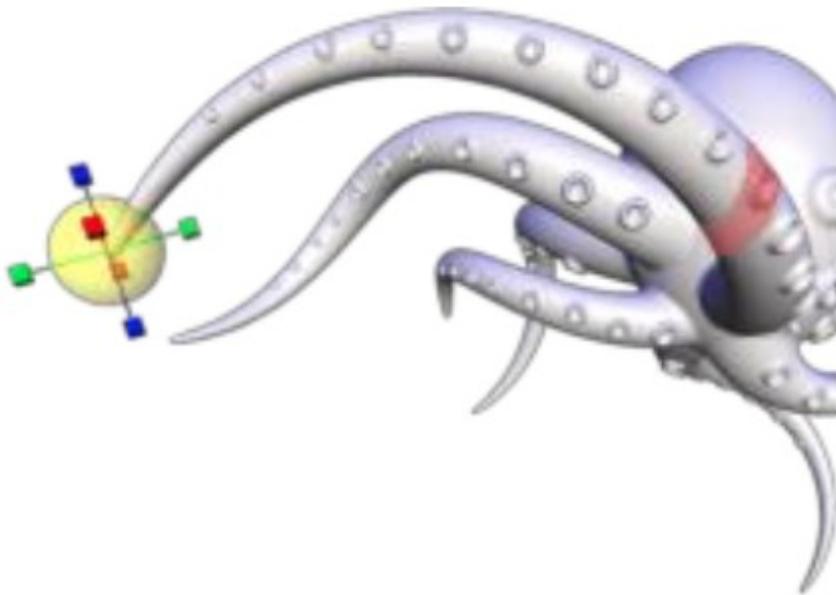
Laplacian Mesh Editing: Motivation

- Meshes are great, but:
 - Geometry is represented in a single *global* coordinate system
 - Single Cartesian coordinate of a vertex doesn't say much



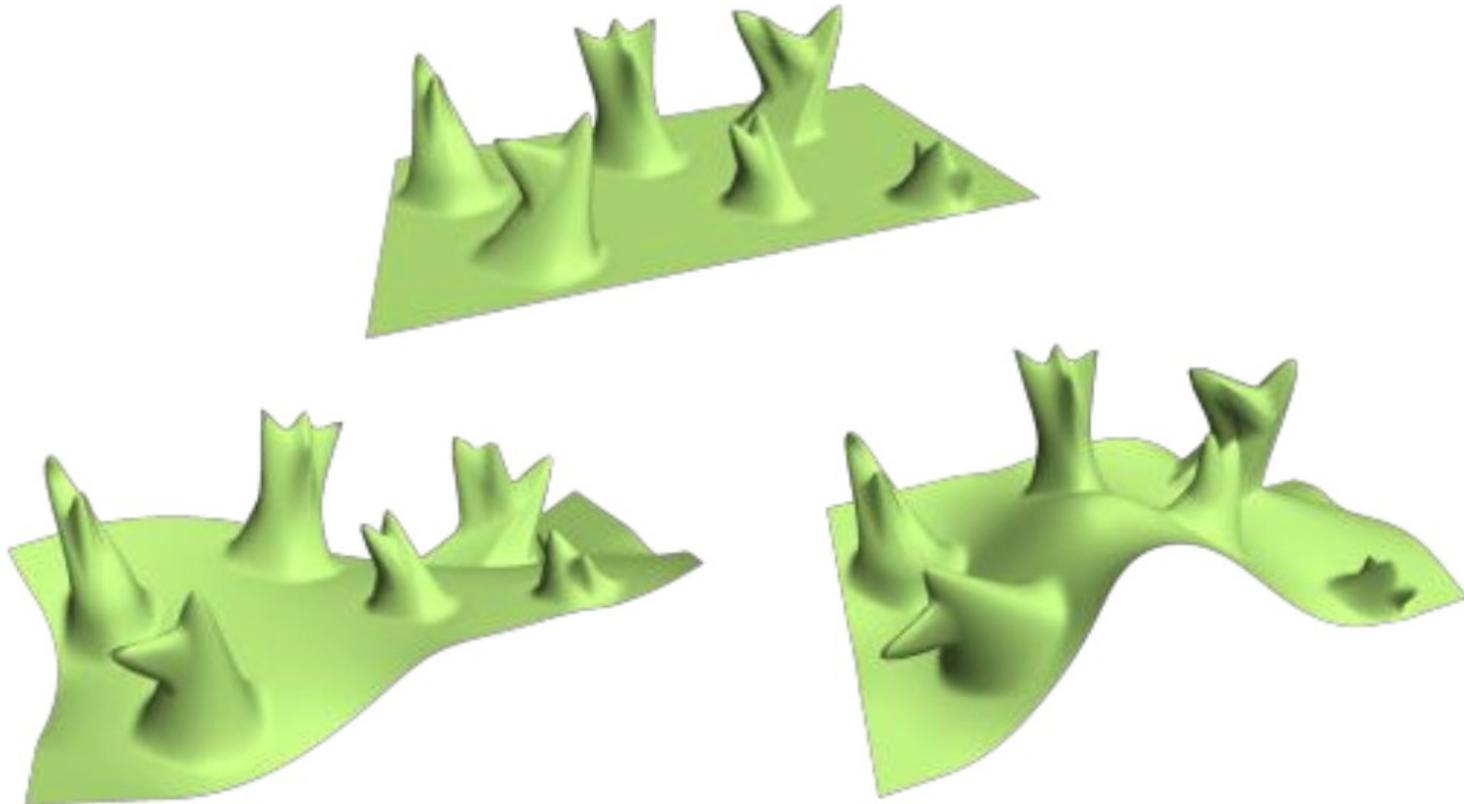
Laplacian Mesh Editing: Motivation

- Meshes are difficult to edit



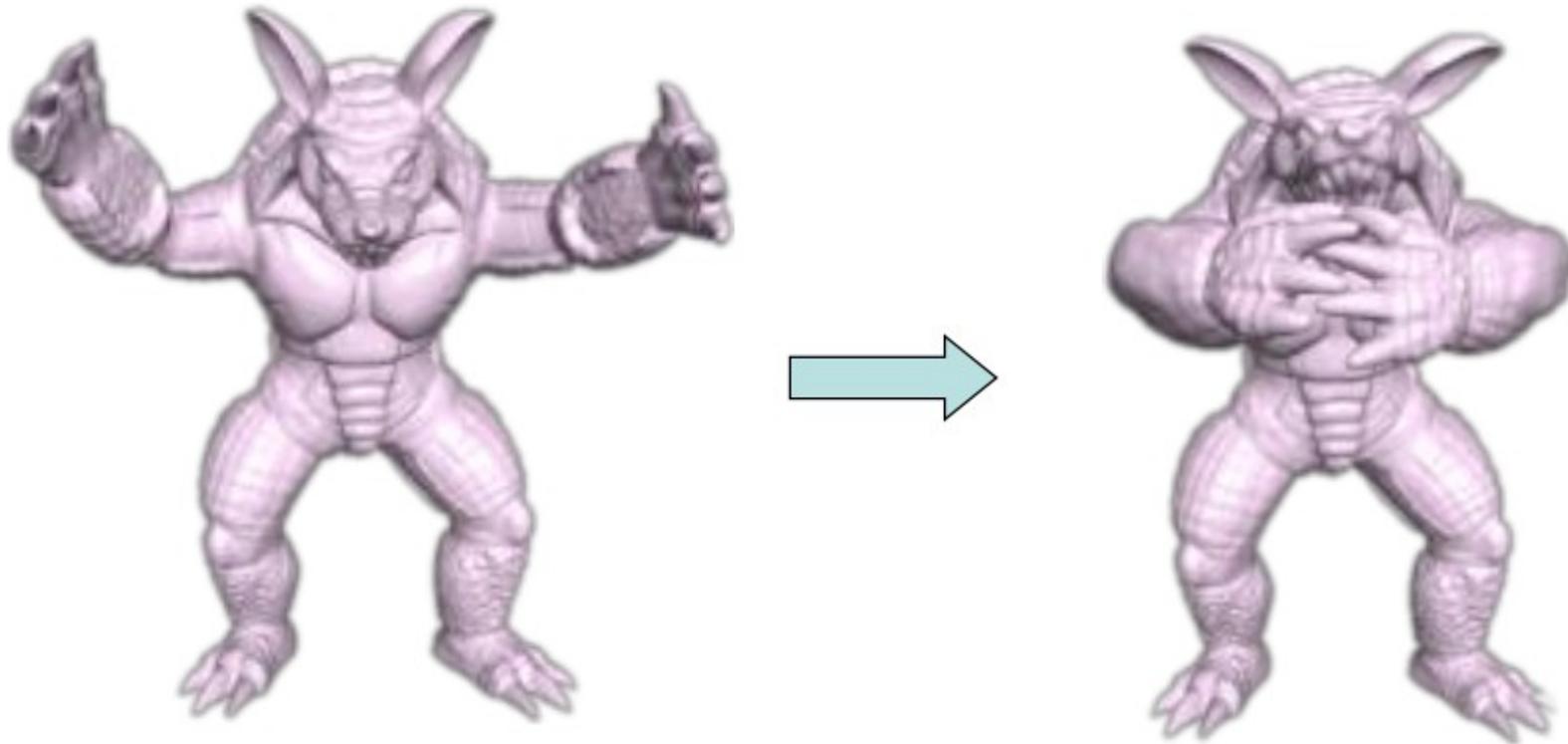
Laplacian Mesh Editing: Motivation

- Meshes are difficult to edit



Laplacian Mesh Editing: Motivation

- Meshes are difficult to edit



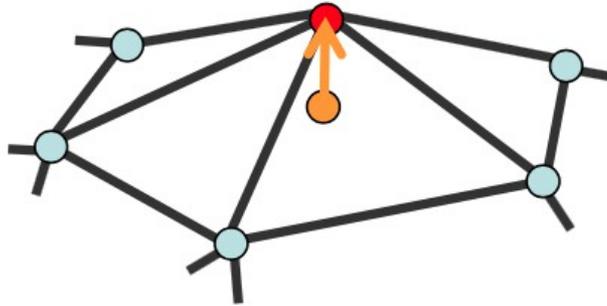
Differential Coordinates

- Represent a point *relative* to its neighbors
- Represent *local detail* at each surface point
- Linear operator takes us from global to differential
- Useful for operations on surfaces where surface details are important



Differential Coordinates

- Detail = surface - **smooth**(surface)
- Smoothing = averaging

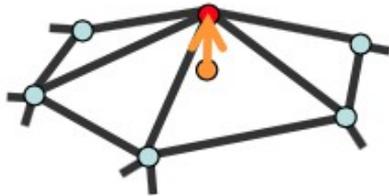


$$\delta_i = \mathbf{v}_i - \frac{1}{d_i} \sum_{j \in N(i)} \mathbf{v}_j$$

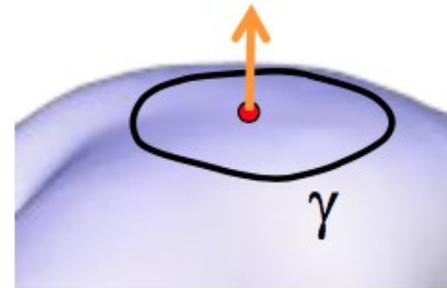
$$\delta_i = \sum_{j \in N(i)} \frac{1}{d_i} (\mathbf{v}_i - \mathbf{v}_j)$$

Connection to the smooth case

- The direction of δ_i approximates the normal
- The magnitude of δ_i approximates the mean curvature



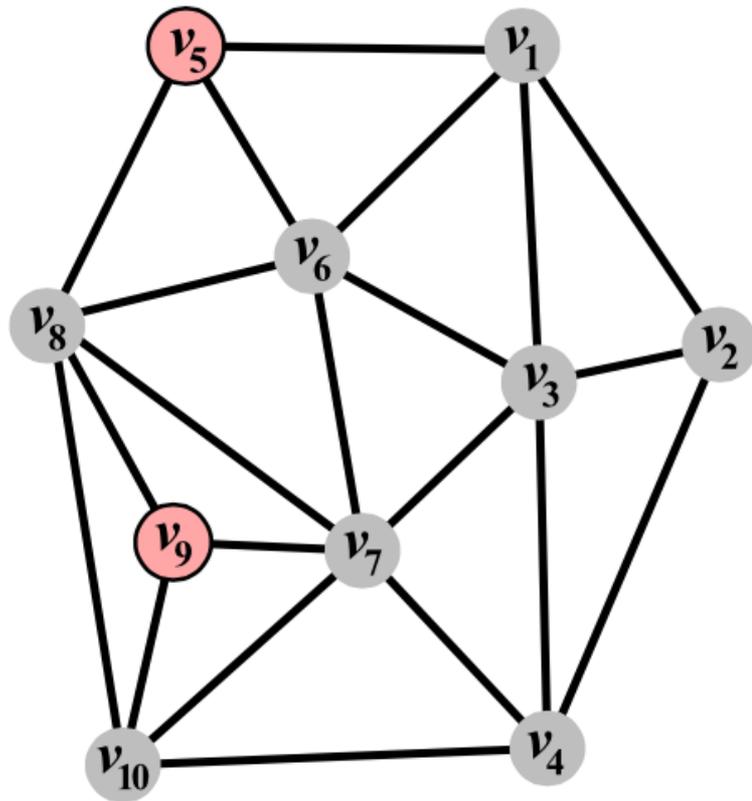
$$\delta_i = \frac{1}{d_i} \sum_{\mathbf{v} \in N(i)} (\mathbf{v}_i - \mathbf{v})$$



$$\frac{1}{\text{len}(\gamma)} \int_{\mathbf{v} \in \gamma} (\mathbf{v}_i - \mathbf{v}) ds$$

$$\lim_{\text{len}(\gamma) \rightarrow 0} \frac{1}{\text{len}(\gamma)} \int_{\mathbf{v} \in \gamma} (\mathbf{v}_i - \mathbf{v}) ds = H(\mathbf{v}_i) \mathbf{n}_i$$

Connection to graph Laplacian



Graph

4	-1	-1		-1	-1				
-1	3	-1	-1						
-1	-1	5	-1		-1	-1			
	-1	-1	4			-1			-1
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				-1	-1	-1	6	-1	-1
						-1	-1	3	-1
			-1			-1	-1	-1	4

Symmetric Laplacian L_s

Weighting Schemes

$$\delta_i = \frac{\sum_{j \in N(i)} w_{ij} (\mathbf{v}_i - \mathbf{v}_j)}{\sum_{j \in N(i)} w_{ij}}$$

- Ignore geometry

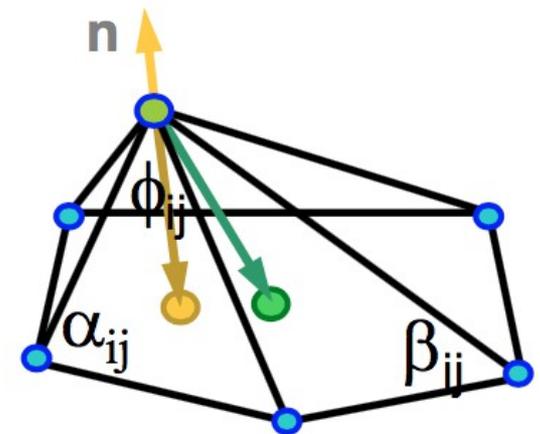
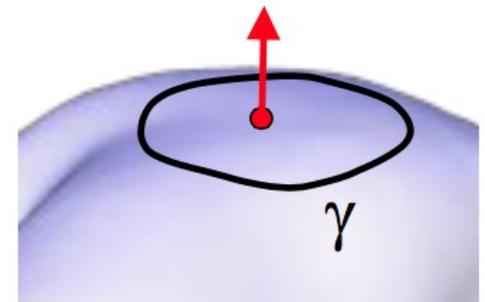
$$\delta_{\text{umbrella}} : w_{ij} = 1$$

- Integrate over circle around vertex

$$\delta_{\text{mean value}} : w_{ij} = \tan \phi_{ij}/2 + \tan \phi_{ij+1}/2$$

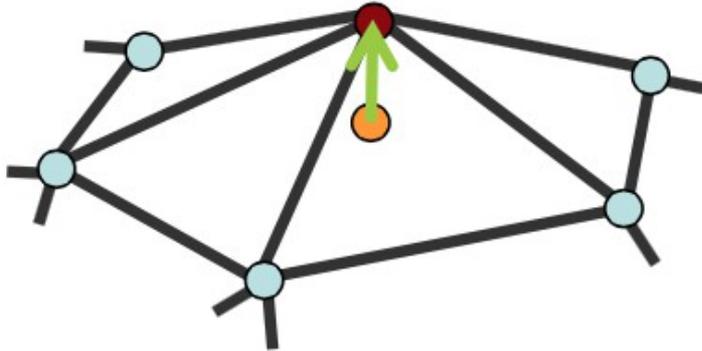
- Integrate over Voronoi region of vertex

$$\delta_{\text{cotangent}} : w_{ij} = \cot \alpha_{ij} + \cot \beta_{ij}$$



Laplacian Mesh

- Vertex positions are represented by Laplacian coordinates $(\delta_x, \delta_y, \delta_z)$



$$\delta_i = \sum_{j \in N(i)} w_{ij} (\mathbf{v}_i - \mathbf{v}_j)$$

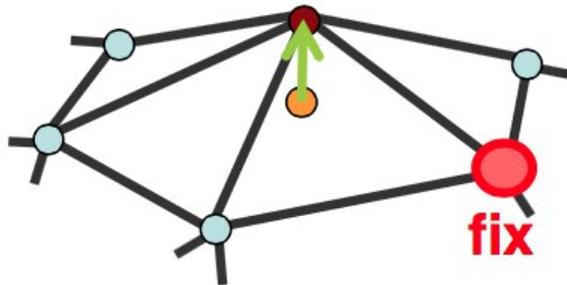
$$\mathbf{L} \mathbf{v}_x = \delta_x$$

$$\mathbf{L} \mathbf{v}_y = \delta_y$$

$$\mathbf{L} \mathbf{v}_z = \delta_z$$

Basic properties

- $\text{rank}(L) = n - c$ ($n - 1$ for connected meshes)
- We can reconstruct the xyz geometry from δ upto translation

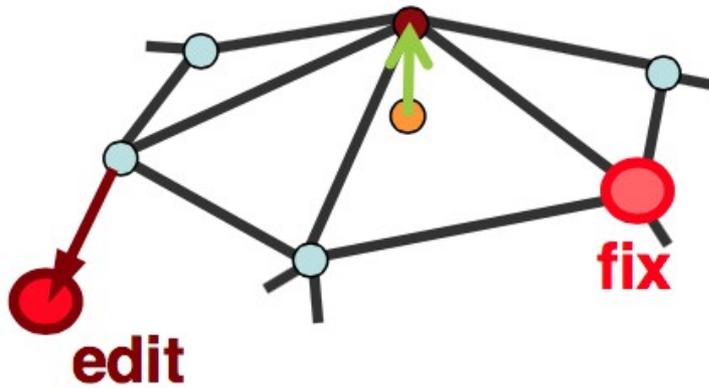


$$\begin{array}{c} \mathbf{L} \\ \hline 1 \end{array} \begin{array}{c} \mathbf{v}_x \end{array} = \begin{array}{c} \delta_x \\ \hline \mathbf{c}_x \end{array}$$

$$\begin{array}{c} \mathbf{L} \\ \hline 1 \end{array} \begin{array}{c} \mathbf{v}_y \end{array} = \begin{array}{c} \delta_y \\ \hline \mathbf{c}_y \end{array}$$

$$\begin{array}{c} \mathbf{L} \\ \hline 1 \end{array} \begin{array}{c} \mathbf{v}_z \end{array} = \begin{array}{c} \delta_z \\ \hline \mathbf{c}_z \end{array}$$

Reconstruction

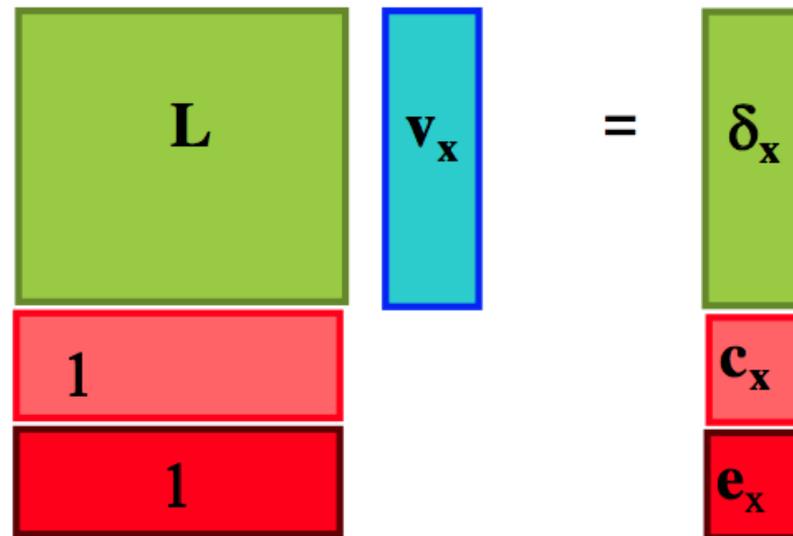


$$\begin{array}{c} \mathbf{L} \\ \hline 1 \\ \hline 1 \end{array} \mathbf{v}_x = \begin{array}{c} \delta_x \\ \hline \mathbf{c}_x \\ \hline \mathbf{e}_x \end{array}$$

$$\begin{array}{c} \mathbf{L} \\ \hline 1 \\ \hline 1 \end{array} \mathbf{v}_y = \begin{array}{c} \delta_y \\ \hline \mathbf{c}_y \\ \hline \mathbf{e}_y \end{array}$$

$$\begin{array}{c} \mathbf{L} \\ \hline 1 \\ \hline 1 \end{array} \mathbf{v}_z = \begin{array}{c} \delta_z \\ \hline \mathbf{c}_z \\ \hline \mathbf{e}_z \end{array}$$

Reconstruction



$$\tilde{\mathbf{x}} = \arg \min_{\mathbf{x}} \left(\left\| L\mathbf{x} - \delta_{\mathbf{x}} \right\|^2 + \sum_{s=1}^k \left| x_k - c_k \right|^2 \right)$$

Reconstruction

$$\begin{array}{|c|} \hline \mathbf{L} \\ \hline 1 \\ \hline 1 \\ \hline \end{array} \begin{array}{|c|} \hline \mathbf{v}_x \\ \hline \end{array} = \begin{array}{|c|} \hline \delta_x \\ \hline \mathbf{c}_x \\ \hline \mathbf{e}_x \\ \hline \end{array}$$

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

Normal Equations:

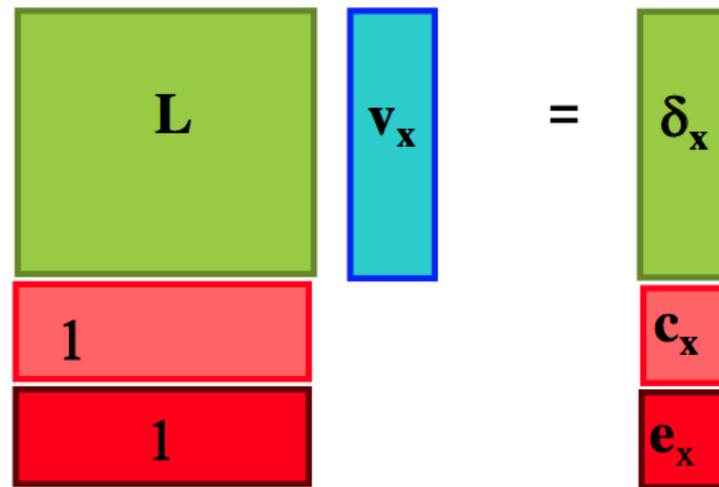
$$\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$$

$$\mathbf{x} = \underbrace{(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}}$$

compute
once

Cool underlying idea

- Mesh vertices are defined by the minimizer of an objective function



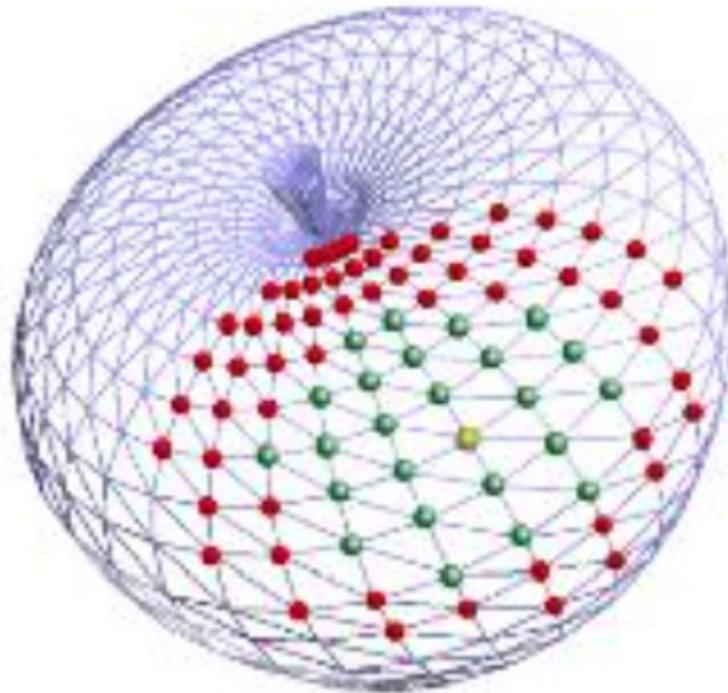
$$\tilde{\mathbf{x}} = \arg \min_{\mathbf{x}} \left(\left\| L\mathbf{x} - \delta_x \right\|^2 + \sum_{s=1}^k |x_k - c_k|^2 \right)$$

What we have so far

- Laplacian coordinates $\delta = L\mathbf{x}$
 - Local representation
 - Translation invariant
- Linear transition from δ to xyz
 - Can constrain one or more vertices
 - Least squares solution

Editing using differential coordinates

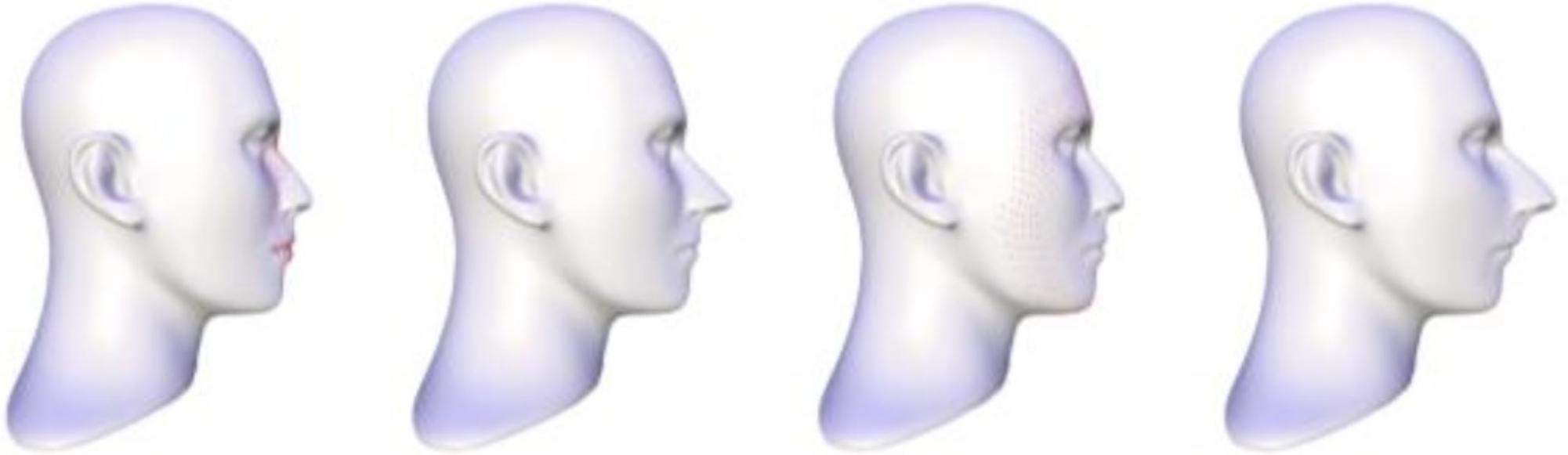
- The editing process from the user's point of view
 - First, set ROI, anchors and handle vertex
 - Move handle vertex to perform edit



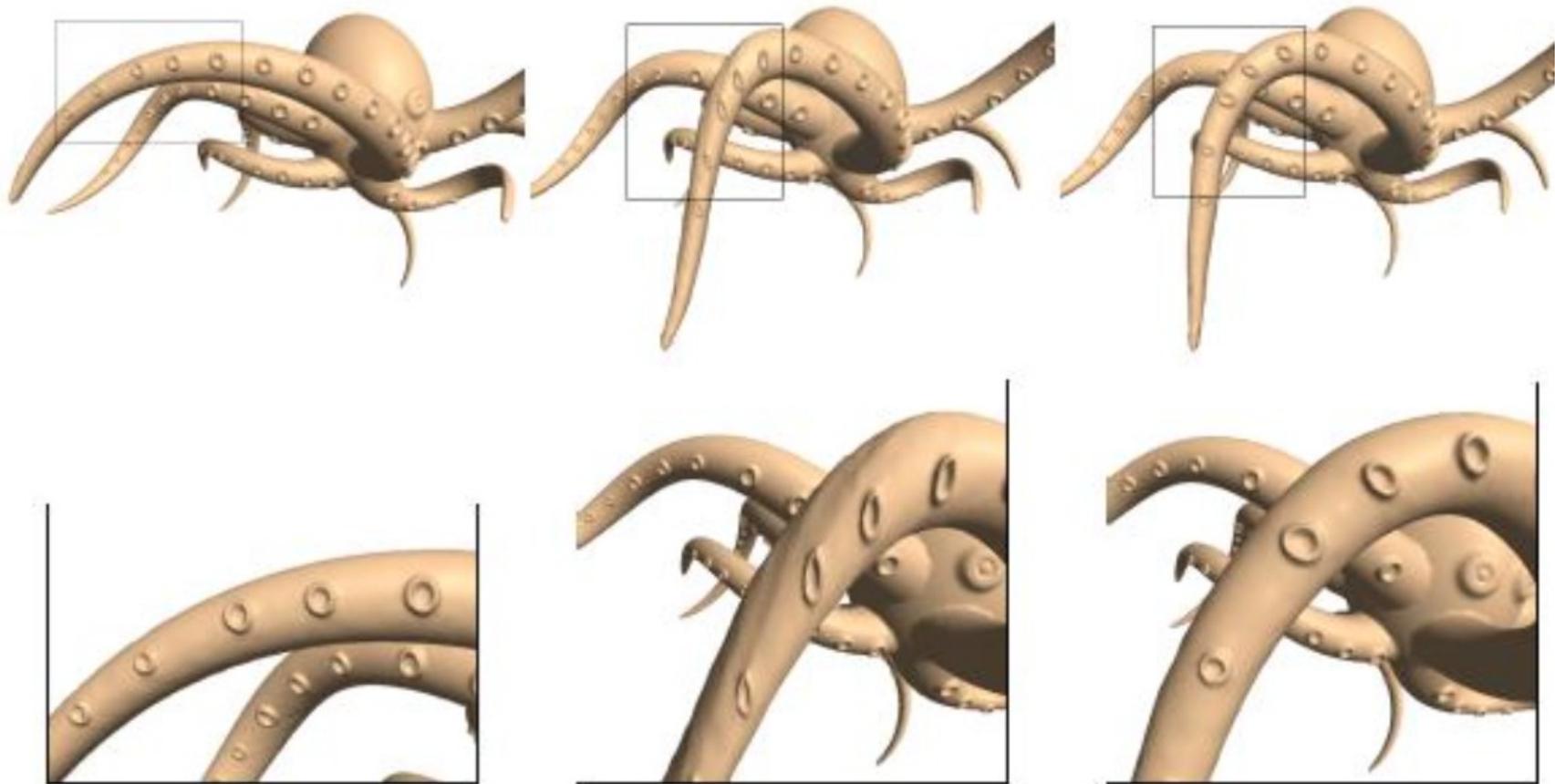
Editing using differential coordinates

- The user moves the handle and interactively the surface changes
- The stationary anchors are responsible for smooth transition of the edited part to the rest of the mesh
- This is done using increasing weight with geodesic distance in the soft spatial equations

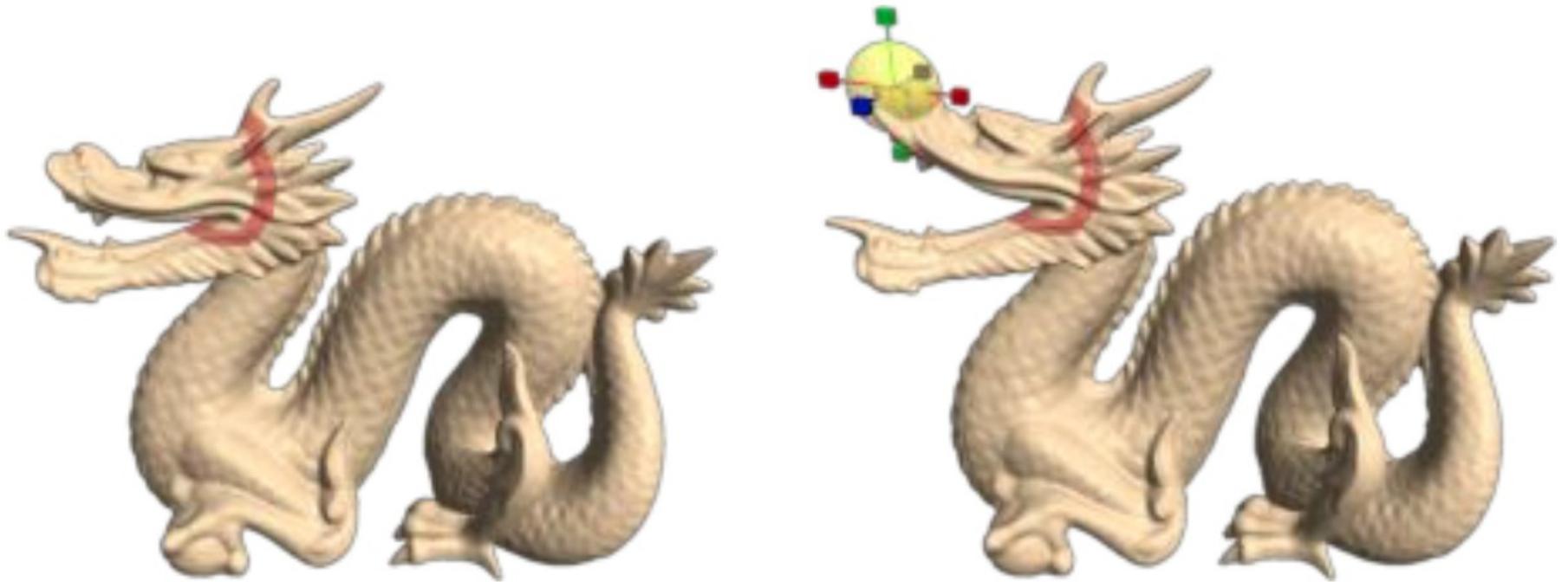
Mesh editing example



Mesh editing example



Mesh editing example



Mesh editing example

