Fundamentals of Computer Graphics

Mesh processing I

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Adjacency matrices

The mesh connectivity can be encoded in adjacency matrices

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, $|E|=e$, $|F|=m$ for a mesh $M=(V,E,F)$

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where $a_{ij} = 1$ if vertex v_i is connected to v_j (that is, $e_{ij} \in E$)

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- Each row and column has at least one 1 (that is, $\sum_{ij} a_{ij} = e$)

Adjacency matrices: Vertex-to-triangle

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- Each row has at least one 1 (each vertex belongs to some triangle)
- Each column sums up to 3 (each triangle has exactly 3 vertices)

Consider the product:

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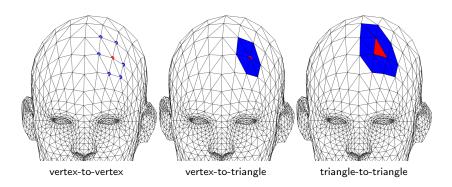
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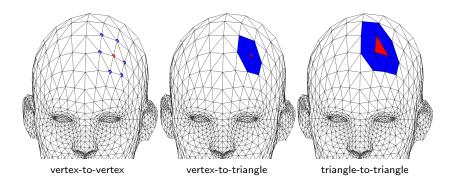
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Examples: Adjacency



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In general we have $m \approx 2n$, and with n in the order of several thousands these adjacency matrices can be very large (quadratic in n)

It is advisable to use sparse data structures to store them

The k-th power of ${\bf A}$ corresponds to composing ${\bf A}$ with itself $k \geq 1$ times For example, for k=2:

$$\mathbf{A}^{2} = \mathbf{A}\mathbf{A} = \begin{pmatrix} 0 & 1 & 0 & \cdots & 1 \\ \vdots & \cdots & \cdots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \cdots & \vdots \\ 1 & 0 & 1 & \cdots & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 & \cdots & 1 \\ \vdots & \cdots & \cdots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \cdots & \vdots \\ 1 & 0 & 1 & \cdots & 0 \end{pmatrix}$$

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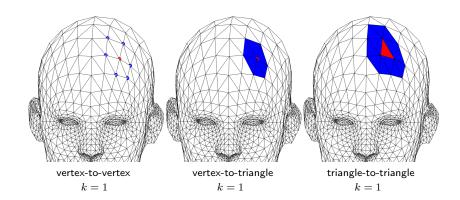
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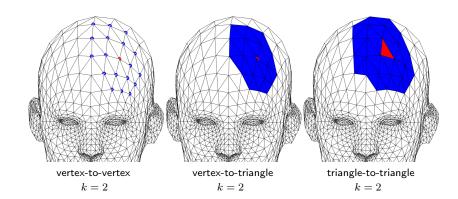
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Manipulating adjacency is useful in many tasks relying upon local context

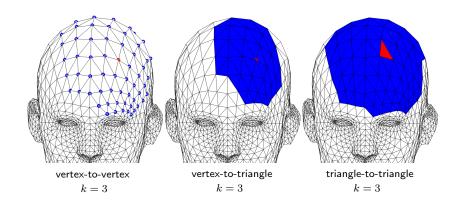
Examples: Powers



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Adjacency matrices as operators

We can see adjacency matrices as operators when applied to functions

For example, g = Af yields a vertex-based function g defined as:

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And similarly for triangle-based functions

Adjacency matrices: Point clouds

Adjacency is a general notion that can be extended to point clouds

(Such notions of adjacency can of course be used on meshes as well)

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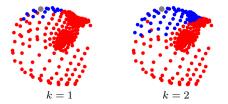


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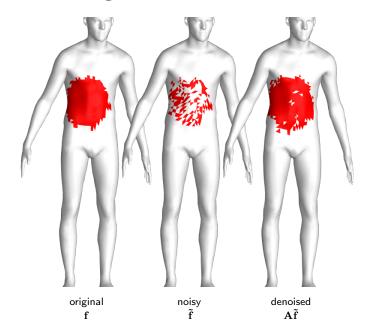
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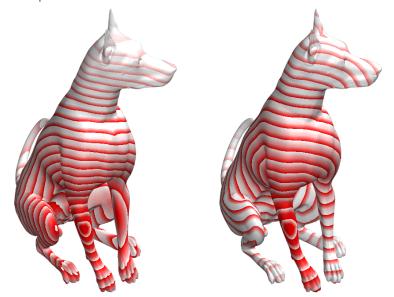
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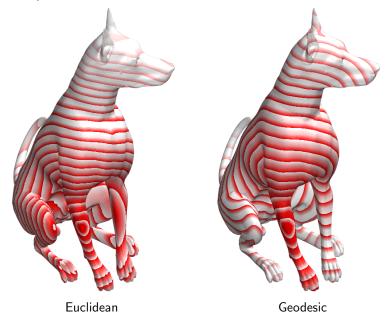
Example: Hole filling

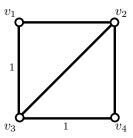


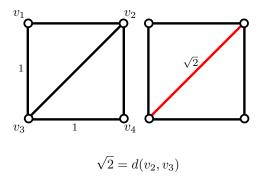
Shortest paths

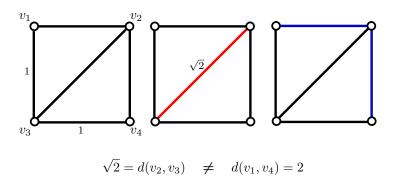


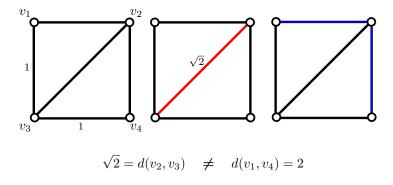
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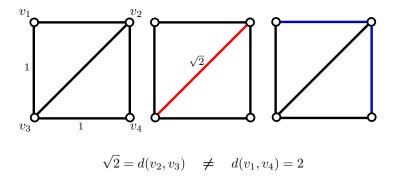






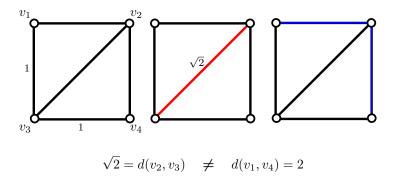


Shortest paths along edges provide upper bounds to exact geodesics



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Still useful with high resolution meshes or for local distances



Shortest paths along edges provide upper bounds to exact geodesics

- Still useful with high resolution meshes or for local distances
- Solved by Dijkstra's algorithm on the mesh graph

Given a mesh graph G=(V,E), consider this condition on vertex v_i :

$$\mathbf{v}_i - \frac{1}{d_i} \sum_{j:(i,j)\in E} \mathbf{v}_j = 0$$

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$$LV = 0$$

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ullet Trivial embedding ${f V}={f 0}$

Is there an embedding satisfying $\mathbf{L}\mathbf{V}=\mathbf{0}$?

- ullet Trivial embedding ${f V}={f 0}$
- ullet One-dimensional subspace $\mathrm{span}(\mathbf{1})$

Is there an interesting embedding satisfying LV = 0?

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Assume $m \geq 1$ anchor vertices $v_s \in \mathcal{A}$ with known 3D position

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Assume $m \geq 1$ anchor vertices $v_s \in \mathcal{A}$ with known 3D position

Then, consider the linear system:

$$\begin{pmatrix} \mathbf{L} \\ \mathbf{A} \end{pmatrix} \mathbf{V} = \mathbf{b}$$

where

$$a_{ij} = \left\{ \begin{array}{ll} 1 & \text{if } v_j \in \mathcal{A} \\ 0 & \text{otherwise} \end{array} \right., \quad b_k = \left\{ \begin{array}{ll} (0,0,0) & k \leq n \\ \mathbf{v}_{s_{k-n}} & n < k \leq n+m \end{array} \right.$$

Sorkine and Cohen-Or, "Least-squares meshes". Proc. SMI, 2004

$$egin{pmatrix} \mathbf{L} \\ \mathbf{A} \end{pmatrix} \mathbf{V} pprox \mathbf{b}$$

$$\min_{\mathbf{V} \in \mathbb{R}^{n \times 3}} \| \begin{pmatrix} \mathbf{L} \\ \mathbf{A} \end{pmatrix} \mathbf{V} - \mathbf{b} \|_2^2$$

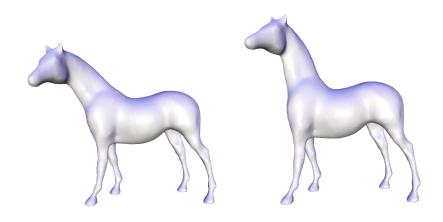
$$\min_{\mathbf{V} \in \mathbb{R}^{n \times 3}} \|\mathbf{L}\mathbf{V}\|_2^2 + \sum_{i=1}^{n+m} \|\mathbf{v}_i - \mathbf{b}_i\|_2^2$$

$$\min_{\mathbf{V} \in \mathbb{R}^{n \times 3}} \|\mathbf{L}\mathbf{V}\|_2^2 + \sum_{v_i \notin \mathcal{A}} \|\mathbf{v}_i\|_2^2 + \sum_{v_i \in \mathcal{A}} \|\mathbf{v}_i - \mathbf{b}_i\|_2^2$$

$$\min_{\mathbf{V} \in \mathbb{R}^{n \times 3}} \|\mathbf{L}\mathbf{V}\|_2^2 + \sum_{v_i \notin \mathcal{A}} \|\mathbf{v}_i\|_2^2 + \sum_{v_i \in \mathcal{A}} \|\mathbf{v}_i - \mathbf{b}_i\|_2^2$$



- Anchor constraints are not satisfied exactly
- At higher resolution, error distributes better among the constraints



Move the anchor positions to do shape modeling

Sorkine and Cohen-Or, "Least-squares meshes". Proc. SMI, 2004

Exercise: Least squares meshes

Implement the example in Figure 4 from:

Sorkine and Cohen-Or, "Least-squares meshes". Proc. SMI, 2004