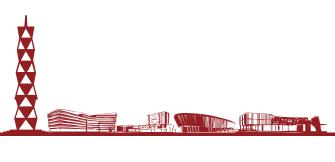




## **CS271 Computer Graphics II**

Lecture 6

Mesh Simplification



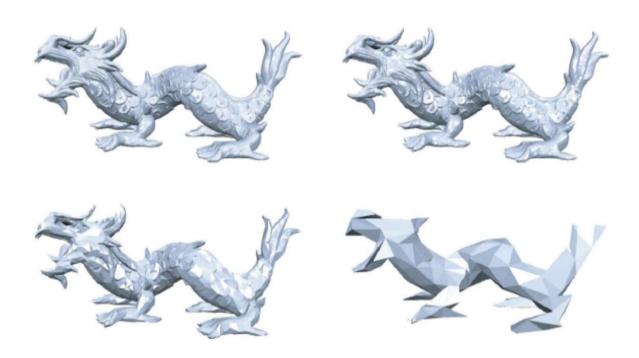


### **Definition**



#### Simplification, aka decimation, approximation, downsampling

- Transform a given polygonal mesh into another mesh with fewer faces, edges, and vertices
- The simplification or approximation procedure is usually controlled by user-defined quality criteria



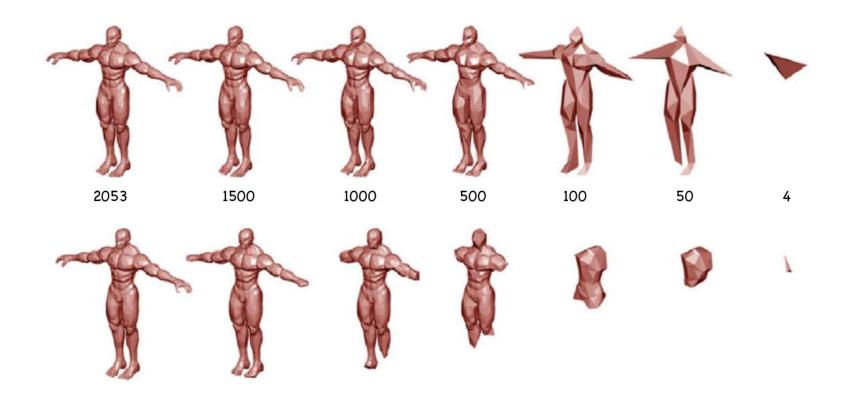






## Curvature-preserved vs. Curvature-removed Criteria





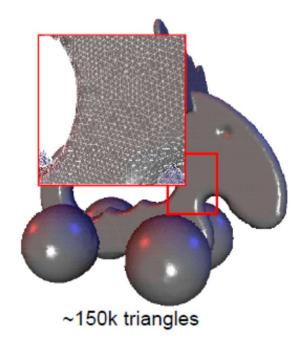


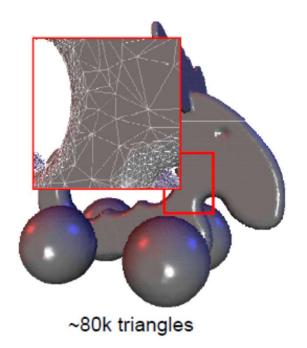






• Oversampled 3D scan data





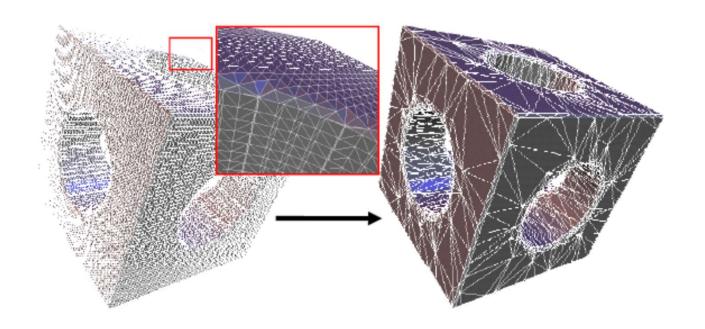








• Over-tessellation: e.g., iso-surface extraction











#### Multi-resolution hierarchies for

- Efficient geometry processing
- Level-of-detail (LOD) rendering













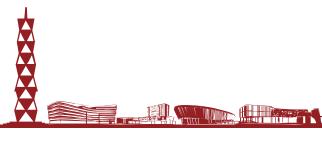




Adaption to hardware capabilities









## **Mesh Simplification**



#### Adjust the complexity of a geometry data set

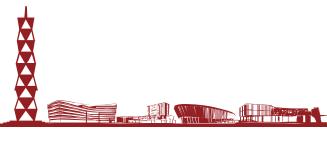
- Since many decimation schemes work iteratively, i.e., they decimate a mesh by removing one vertex at a time, they usually can be inverted
- Hierarchical method

#### **Problem Statement**

- Given:M = (V, F)
- Find: M' = (V', F') such that
- |V'| = n < |V| and ||M M'|| is minimal, or
- $\|M-M'\| < \varepsilon$  and  $\|V'\|$  is minimal

#### Respect additional fairness criteria

Normal deviation, triangle shape, scalar attributes, etc.





## Mesh Simplification

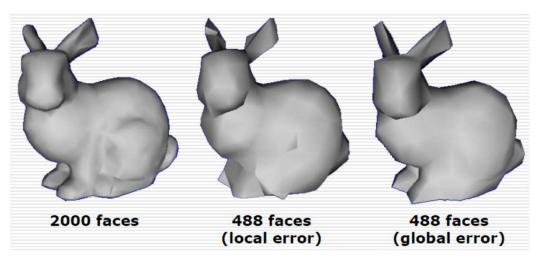


#### Start with the original fine mesh

- Simply progressively
  E.g., collapse edges, vertex clustering
- Aim to keep original appearance

  Normal deviation, triangle shape, scalar attributes

  Error control

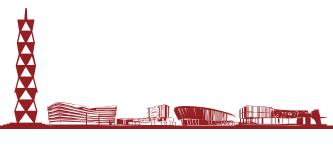








# **Local Operations**

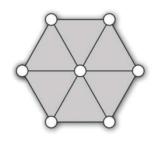




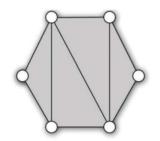
### **Local Simplification Operator**



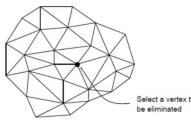
#### Vertex Removal



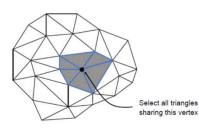




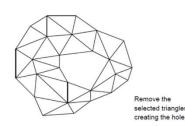
- Deletion & Insertion
- Reversible



Select a vertex to be eliminated



Select all triangles sharing the vertex



Remove the selected triangles, creating the hole



Fill the hole with new triangles



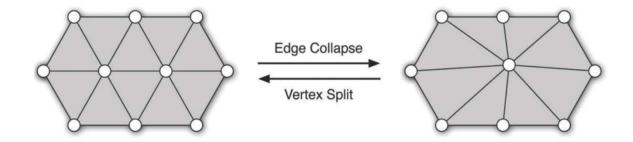




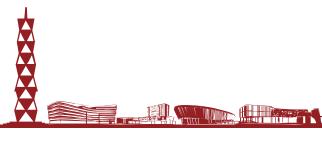
### **Local Simplification Operator**



Edge Collapse



- Merge two adjacent vertices
- Simple to implement
- Well-suited for implementing geomorphing

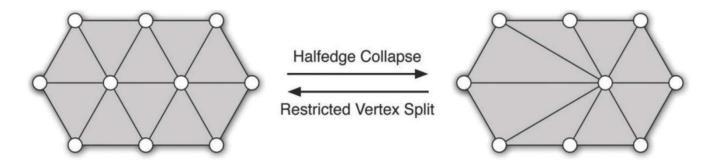




### **Local Simplification Operator**



Half-edge Collapse



• Collapse edge into one end point

Special case of vertex removal Special case of edge collapse

- After collapse: n(E) 3, n(V) 1, n(F) 2
- According to Euler Formula: unchanged
- Half-edge collapsing would not change the genus of a mesh
- Should determine whether collapse is ok (may introduce non-manifold structure)

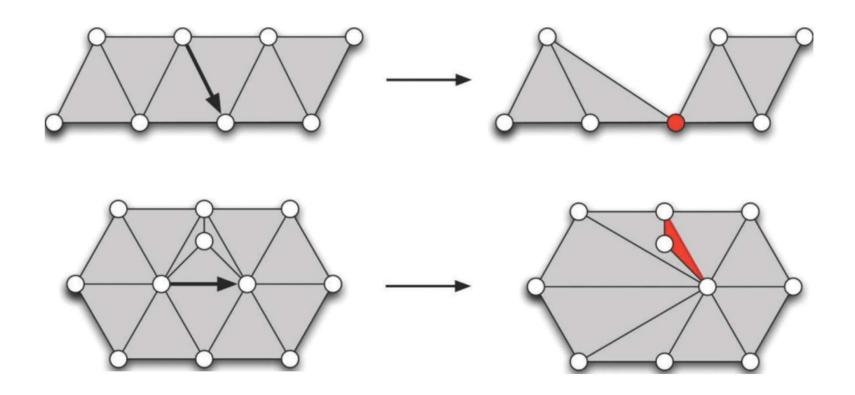






## Topologically Illegal (half-)edge Collapses











## Simplification via Edge Collapse



#### One popular scheme: iteratively collapse edges

Greedy algorithm from a general overview:

- Assign each edge a cost
- Collapse edge with least cost
- Repeat until target number of elements is reached
- Particularly effective cost function: Quadric Error Metric

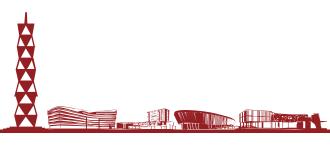








# **Quadric Error Metric**



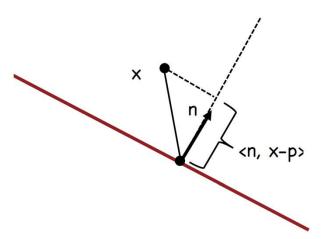


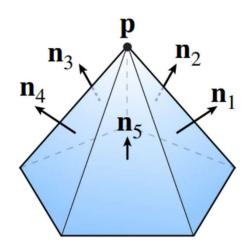
## **Quadric Error Metric (QEM)**



#### Approximate distance to a collection of triangles

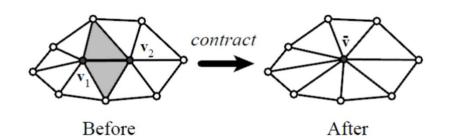
- Q: Distance to plane w/ normal n passing through point p?
- A:  $dist(x) = \langle n, x \rangle \langle n, p \rangle = \langle n, x \rangle$
- Quadric error is then sum of squared point-to-plane distances





$$Q(x) := \sum_{i=1}^{k} \langle n_i, x - p \rangle^2$$

$$Q^e = Q_1^v + Q_2^v$$









### **Quadric Error – Homogeneous Coordinates**



Suppose in coordinates we have

- A query point x = (x, y, z)
- A normal n = (a, b, c)
- An offset **d** := <**n**, **p**>

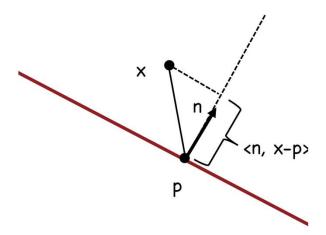
In homogeneous coordinates, let

- u := (x, y, z, 1)
- v := (a, b, c, d)



- Squared distance is  $\langle u, v \rangle 2 = u^T(vv^T)u =: u^TKu$
- Matrix  $\mathbf{K} = \mathbf{v}\mathbf{v}^{\mathsf{T}}$  encodes squared distance to plane

Key idea: sum of matrices K distance to union of planes  $\mathbf{u}^{\mathsf{T}}\mathbf{K}_{1}\mathbf{u} + \mathbf{u}^{\mathsf{T}}\mathbf{K}_{2}\mathbf{u} = \mathbf{u}^{\mathsf{T}}(\mathbf{K}_{1} + \mathbf{K}_{2})\mathbf{u}$ 



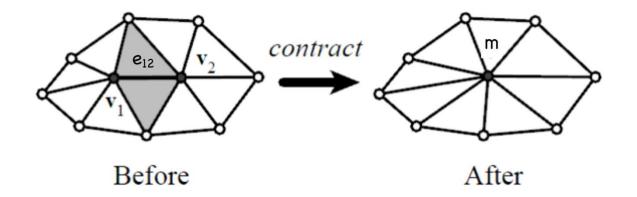
$$K = \begin{bmatrix} a^2 & ab & ac & ad \\ ab & b^2 & bc & bd \\ ac & bc & c^2 & cd \\ ad & bd & cd & d^2 \end{bmatrix}$$



### **Quadric Error of Edge Collapse**



- How much does it cost to collapse an edge e12?
   Idea: compute midpoint m, measure error Q(m) = m<sup>T</sup>(K<sub>1</sub>+K<sub>2</sub>)m
- Error becomes "score" for e<sub>12</sub>, determining priority



- Better idea: find point x that minimize error!
- But how to minimize quadric error?







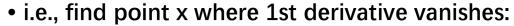
## Revisit: Minimizing a Quadratic Function



Suppose you have a function  $f(x) = ax^2 + bx + c$ 

- Q: What does the graph of this function look like?
- Q: How do we find the minimum?

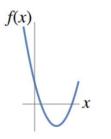


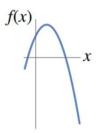


$$f'(x) = 0$$

$$2ax + b = 0$$

$$x = -b/2a$$











## Minimizing Quadratic Polynomial



Not much harder to minimize a quadratic polynomial in **n** variables

- Can always write in terms of a symmetric matrix A
- E.g., in 2D:  $f(x, y) = ax^2 + bxy + cy^2 + dx + ey + g$

$$\mathbf{x} = \begin{bmatrix} x \\ y \end{bmatrix} \quad A = \begin{bmatrix} a & b/2 \\ b/2 & c \end{bmatrix} \quad \mathbf{u} = \begin{bmatrix} d \\ e \end{bmatrix}$$

$$f(x, y) = x^{T}Ax + u^{T}x + g$$
  
(will have the same form for any n)

- Q: How do we find a critical point (min/max/saddle)?
- A: Set derivative to zero!

$$2A\mathbf{x} + \mathbf{u} = 0$$
$$\mathbf{x} = -1/2 \mathbf{A}^{-1}\mathbf{u}$$







## Positive Definite Quadratic Form

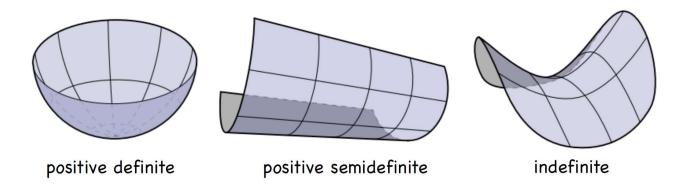


Just like our 1D parabola, critical point is not always a min!

- Q: In 2D, 3D, nD, when do we get a minimum?
- A: When matrix A is positive-definite:

$$X^TAX > 0 \quad \forall X$$

- 1D: Must have  $xax = ax^2 > 0$ , i.e., a is positive!
- 2D: Graph of function looks like a "bowl":









## Minimizing Quadric Error



Find "best" point for edge collapse by minimizing quadratic form

$$\min_{\mathbf{u} \in \mathbb{R}^4} \mathbf{u}^T K \mathbf{u}$$

- Already know 4th (homogeneous) coordinate for a point is 1
- So, break up our quadratic function into two pieces:

$$\begin{bmatrix} \mathbf{x}^{\mathsf{T}} & 1 \end{bmatrix} \begin{bmatrix} B & \mathbf{w} \\ \mathbf{w}^{\mathsf{T}} & d^2 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ 1 \end{bmatrix}$$
$$= \mathbf{x}^{\mathsf{T}} B \mathbf{x} + 2 \mathbf{w}^{\mathsf{T}} \mathbf{x} + d^2$$

- Now we have a quadratic polynomial in the unknown position  $\mathbf{x} \in \mathbb{R}^3$
- Can minimize as before:

$$2B\mathbf{x} + 2\mathbf{w} = 0$$

$$\mathbf{x} = -B^{-1}\mathbf{w}$$





## **QEM Simplification: Final Algorithm**



• Input: a mesh

• Output: a simplified mesh

#### Initialization:

Compute **K** for each triangle (squared distance to plane) Set  $\mathbf{K_i}$  at each vertex to sum of  $\mathbf{Ks}$  from incident triangles For each edge  $\mathbf{e_{ij}}$ :

Set  $K_{ij} = K_i + K_j$ 

Find point x minimizing error, set cost to  $K_{ii}(x)$ 

#### Until we reach target number of triangles:

Collapse edge eij with smallest cost to optimal pont  ${\bf x}$  Set quadric at new vertex to  ${\bf K}_{ij}$  Update cost of edges touching new vertex



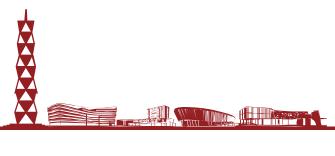
Full Resolution

60,000 triangles

1000 triangles



# **Variational Shape Approximation**

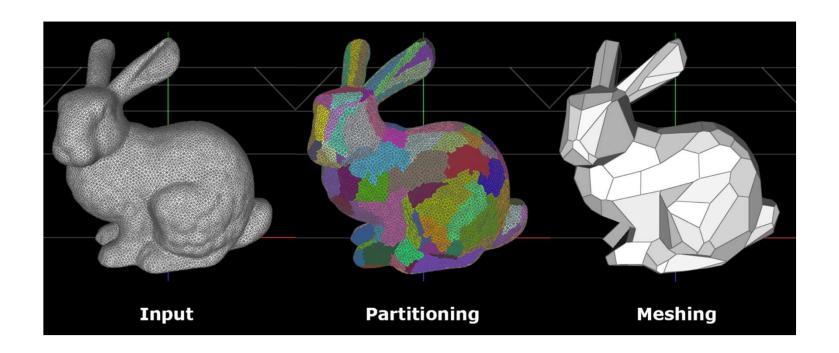




## Variational Shape Approximation (VSA)



VSA is highly sensitive to features and symmetries and produces anisotropic meshes of high approximation quality





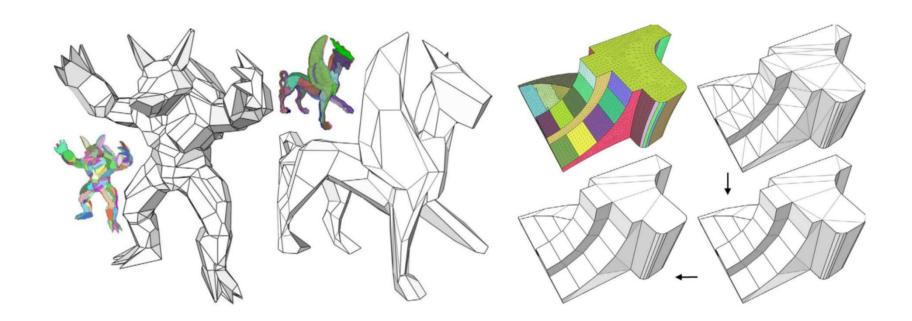




# Variational Shape Approximation (VSA)



- The input shape is approximated by a set of proxies
- ullet A plane in space through the point  $\mathbf{x_i}$  with normal direction  $\mathbf{n_i}$









## **Region Representation**



- M: a triangle mesh
- $\mathbf{R} = \{\mathbf{R}_1, ..., \mathbf{R}_k\}$ : a partition of  $\mathbf{M}$  into  $\mathbf{k}$  regions  $R_1 \cup ... \cup R_k = M$
- Proxies:  $P = \{P_1, ..., P_k\}, P_i = (x_i, n_i)$

### Distance metrics between $R_i$ and $P_i$

ullet The squared orthogonal distance of x from the plane  $P_i$ 

$$L^{2}(R_{i}, P_{i}) = \int_{x \in R_{i}} (n_{i}^{T}x - n_{i}x_{i})^{2} dA$$

• A measure of the normal field:

$$L^{2,1}(R_i, P_i) = \int_{x \in R_i} ||n(x) - n_i||^2 dA$$





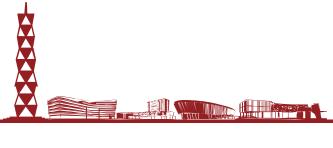
### Goal of VSA



Given a number k and an error metric  $E(L_2 \text{ or } L_{2,1})$ , find a set  $R = \{R_1, ..., R_k\}$  of regions and a set  $P = \{P_1, ..., P_k\}$  of proxies such that the global distortion

$$E(R, P) = \sum_{i=1}^{k} E(R_i, P_i)$$

is minimized





## Lloyd's Clustering Algorithm



 The algorithm iteratively alternates between a geometry partitioning phase and a proxy fitting phase

#### Geometry partitioning phase

- A set of regions that best fit a given set of proxies
- Modifies the set R of regions to achieve a lower approximation error while keeping the proxies P fixed

#### · Proxy fitting phase

- The partitioning is kept fixed, and the proxies are adjusted to minimize approximation error
- L2 metric: the best proxy is the least-squares fitting plane
- $L^{2,1}$  metric: the proxy normal  $n_i$  is just the area-weighted average of the triangle normals

#### Initialization

- Randomly picking k triangles as R
- The planes of k triangles are used to initialize P









## **More Paper**



- Liu Y J, Xu C X, Fan D, et al. **Efficient construction and simplification of Delaunay meshes**[J]. ACM Transactions on Graphics (TOG), 2015, 34(6): 1-13.
- Yi R, Liu Y J, He Y. **Delaunay mesh simplification with differential evolution**[J]. ACM Transactions on Graphics (TOG), 2018, 37(6): 1-12.
- Liang Y, He F, Zeng X. **3D mesh simplification with feature preservation based on whale optimization algorithm and differential evolution**[J]. Integrated Computer-Aided Engineering, 2020, 27(4): 417-435.

