

# Simplification of 3D Meshes

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## Outline

- Motivation
- Taxonomy of simplification methods
- Hoppe et al, Mesh optimization
- Hoppe, Progressive meshes

## Motivation

- High detailed meshes becoming widely available in computer vision, scientific visualization, terrain data from satellite... etc.
- Need to store, transmit, analyze, edit and display them efficiently.



> 1M triangles!

## Goals

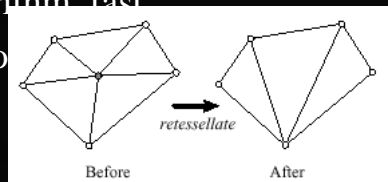
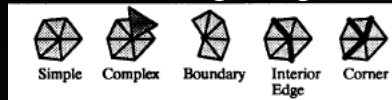
- Reduce number of polygons
  - Faster rendering
  - Less storage
  - Simpler manipulation
- Other qualities
  - General (non-manifold)
  - Efficiency
  - Preserve attributes other than geometry

## Taxonomy of methods

- Manifold Simplification
  - Vertex decimation
  - Wavelet
  - Edge collapse
- Non-manifold Simplification
  - Vertex clustering

## Mesh Decimation [Schroeder et al 92]

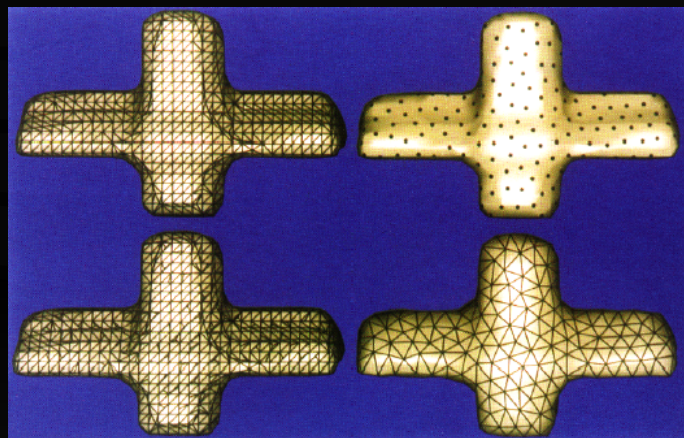
- Multiple passes
- For each pass remove all vertices that matches certain criteria and retriangulate
- Criteria: distance to plane/edge
- Advantage: simple algorithm, fast
- Disadvantage: big memo



## Re-Tiling [Turk 92]

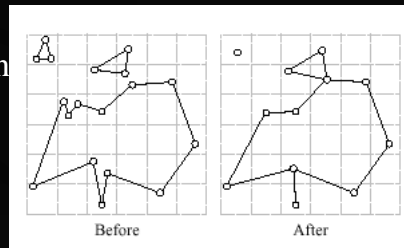
- Steps:
  - Generate random points on surface
  - Iterative repulsion spread the points out uniformly
  - Add new set of points to the surface and *mutual tessellate*
  - Remove old vertices one by one yielding a new triangulation
- Variant to put points adaptively depending on curvature
- Advantage : maintain topology
- Disadvantage : complex algorithm, blur sharp features.

## Re-Tiling (con't)



## 3D Grid Method [Rossignac-Borrel 93]

- Steps:
  - Subdivide the bounding volume into regular grid
  - Merge all vertices within each cell together into a new vertex
- Advantage: very general, fast
  - Form triangles according
- Disadvantage : low quality, non-adaptive

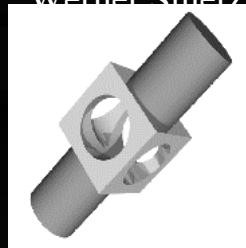


## Optimization [Hoppe et al 93]

- Optimization based simplification. Minimize energy function.
- Repeat semi-random changes to topology and optimize the geometry.
- Work on manifold
- Advantages: High quality, less sensitive to noise
- Disadvantages: slow

# Mesh Optimization

Hugues Hoppe, Tony DeRose,  
Tom Duchamp, John McDonald,  
Werner Stuetzle, SIGGRAPH 93



# Optimization

- Similar problem to simplification
- Starting with sample data points from the surface and a initial mesh, find a simpler mesh
- Minimize energy function that describes the conciseness and accuracy of the mesh

## Mesh Representation

- A mesh is represented by a pair  $(K, V)$ 
  - $K$  is a simplicial complex representing the connectivity of the vertices/edges/faces.
  - $V = \{v_1, \dots, v_m\}$  is a set of  $m$  vertex positions defining the shape of the mesh in  $\mathbf{R}^3$
- If mesh is not self-intersecting, every point can be represented by a barycentric vector

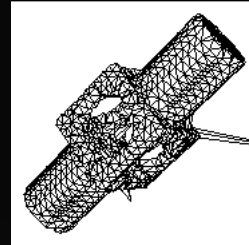
## Energy Function

- $E(K, V) = E_{\text{dist}}(K, V) + E_{\text{rep}}(K) + E_{\text{spring}}(K, V)$

- $E_{\text{dist}}(K, V) = \sum_{i=1}^n d^2(x_i, \varphi_V(|K|))$

- $E_{\text{rep}}(K) = c_{\text{rep}} m$

- $E_{\text{spring}}(K, V) = \sum_{\{j,k\} \in K} \kappa \|v_j - v_k\|^2$



Without  $E_{\text{spring}}$ , spikes possible for region with no data points

## Minimizing the Energy Function

- A bigger  $c_{\text{rep}}$   $\rightarrow$  sparser representation.
- Outer loop optimize  $\phi$  over  $K$  (Discrete)
- Inner loop optimize over  $V$  for a fixed  $K$  (Continuous)

## Optimization for fixed $K$

- = Minimize  $\sum_{i=1}^n d^2(\mathbf{x}_i, \phi_V(|K|)) + \sum_{\{j,k\} \in K} \kappa \|\mathbf{v}_j - \mathbf{v}_k\|^2$
- For each  $\mathbf{x}_i$ ,  $\text{distance}_i = \min_{\mathbf{b}_i \in |K|} \|\mathbf{x}_i - \phi_V(\mathbf{b}_i)\|^2$
- New objective function :  
–  $E(K, V, B) = \sum_{i=1}^n \min_{\mathbf{b}_i \in |K|} \|\mathbf{x}_i - \phi_V(\mathbf{b}_i)\|^2 + \sum_{\{j,k\} \in K} \kappa \|\mathbf{v}_j - \mathbf{v}_k\|^2$



## Optimization over fixed $K$

- Two subproblems :
  - For fixed vertices  $V$ , find optimal barycentric coordinate vectors  $B$  by projection
  - For fixed  $B$ , find optimal  $V$  by solving a linear least squares problem
- Find optimal solutions to both subproblems, so  $E(K, V, B)$  must converge.

## Projection Subproblem

- Find optimal  $B$  by projecting  $x_i$  onto the mesh.
- To accelerate, build a spatial partitioning data structure so that for each point only consider nearby subset of faces
- Assume a point's projection lies in the neighborhood of its projection in previous iteration (perform well in practice)

## Linear Least Squares Subproblem

- Minimize for  $x, y, z$  coordinates,  $\|A\mathbf{v} - \mathbf{d}\|^2$ 
    - $\mathbf{v}$   $m$ -vector
    - $A$   $(n+e) \times m$  matrix
    - $\mathbf{d}$   $(n+e)$ -vector
- For  $m$  mesh vertices,  $n$  data points,  $e$  edges
- First  $n$  rows of  $\mathbf{d}$  contains the  $n$  data points
  - Next  $e$  rows of  $\mathbf{d}$  are zeroes
  - $\mathbf{v}$  contains the  $m$  mesh vertices

## Linear Least Squares (con't)

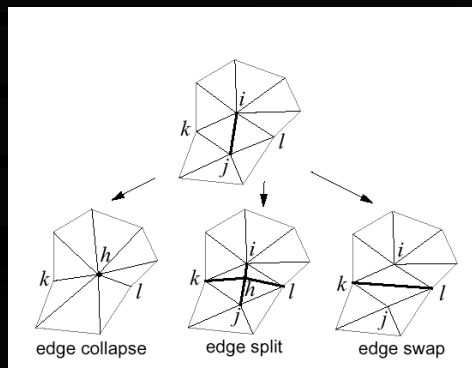
$$\|A\mathbf{v}_j - \mathbf{d}_j\|^2$$

- First  $n$  rows of  $A$  contains the barycentric coordinates computed in projection (at most 3 non-zero entries)
  - Next  $e$  rows represent the spring energy: each contains an  $\sqrt{\kappa}$  and  $-\sqrt{\kappa}$  entries in columns corresponding to indices of edge's endpoints (exactly 2 non-zero entries)
- $A$  is sparse. Use “conjugate gradient method” to solve in  $O(n+m)$  time

## Optimization over K

- Take a legal move and accept if it gives lower energy. Terminate if a number of trials failed to give a lower energy.
- Legal move : application of one of these that leaves the topological type of K unchanged
- Three elementary transformations:  
edge collapse/split/swap

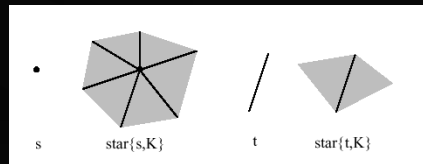
## Legal move



- Split – Always legal
- Collapse : if and only if...
  - For all vertices  $\{k\}$  adjacent to both  $\{i\}$  and  $\{j\}$ ,  $\{i, j, k\}$  is a face
  - If  $\{i\}$  and  $\{j\}$  are both boundary vertices,  $\{i, j\}$  is a boundary edge
  - K has more than 4 vertices if  $\{i\}$  and  $\{j\}$  both are not boundary vertices, or K has more than 3 vertices if either  $\{i\}$  or  $\{j\}$  are boundary vertices
- Swap – if and only if  $\{k, l\} \notin K$
- Proof in Hoppe et al, TR 93-01-01 University of Washington

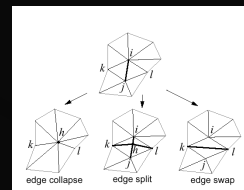
## Evaluation of a legal move

- Instead of re-computing the global energy after a legal move, only compute the change in energy for the local submesh
- For  $s \in K$ , define  $\text{star}(s;K) = \{s' \in K : s \text{ non-empty subset of } s'\}$



## Edge Collapse

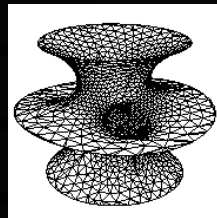
- To evaluate collapsing of an edge  $\{i,j\}$ , take the submesh to be  $\text{star}(\{i\};K) \cup \text{star}(\{j\};K)$ . Optimize over the new vertex  $h$  while holding all other constant
- Attempt optimizations starting at  $\mathbf{v}_i$ ,  $\mathbf{v}_j$ , and  $\frac{1}{2}(\mathbf{v}_i + \mathbf{v}_j)$ . Accept the best one.
- Instead of checking for self-intersection after collapse which is expensive, use a threshold value for the maximum dihedral angle of edges in  $\text{star}(\{h\};K')$



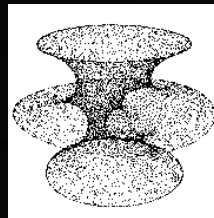
## Edge split/swap

- Edge split →
  - Same procedure as collapse, using the submesh to be  $\text{star}(\{i,j\}; K)$ . Initial value of  $\mathbf{v}_h$  be  $\frac{1}{2}(\mathbf{v}_i + \mathbf{v}_j)$ .
- Edge swap-
  - For a swap that replace an edge  $\{i,j\}$  with  $\{k,l\}$ , choose the best of the two optimizations, one with submesh  $\text{star}(\{k\}; K')$  varying vertex  $\mathbf{v}_k$ , another with submesh  $\text{star}(\{l\}; K')$  varying vertex  $\mathbf{v}_l$ .

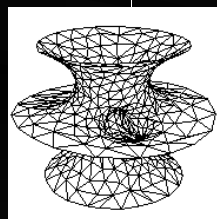
## Results



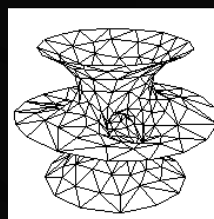
Initial mesh  
(2032 vertices)



Sample Points  
(6752 vertices)

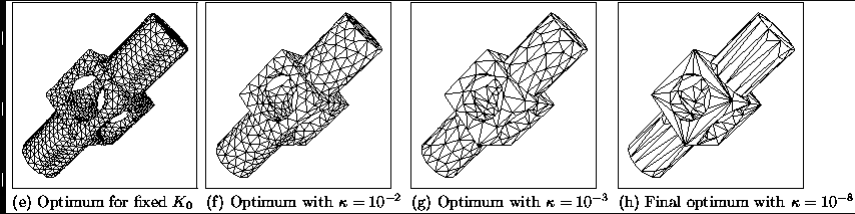


$c_{\text{rep}} = 10^{-5}$   
(487 vertices)

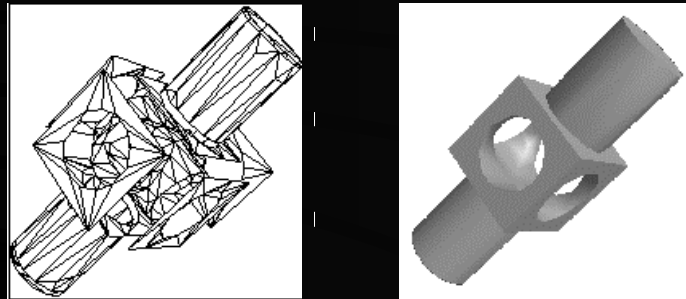


$c_{\text{rep}} = 10^{-4}$   
(239 vertices)

## Results

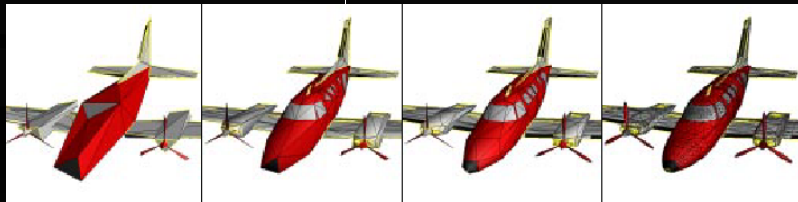


## Results - Segmentation



# Progressive Meshes

Hugues Hoppe  
SIGGRAPH 96



## Additional Goal

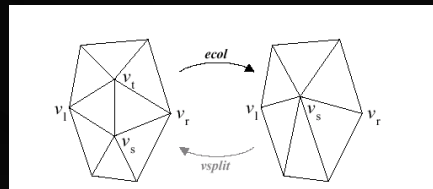
- Lossless, continuous-resolution, progressive representation
- More compact representation
- Apart from geometry, also preserve other attributes (colors, normals, materials, texture coordinates...)

# Mesh Representation

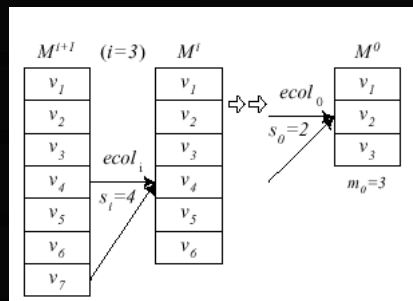
- Edge collapse only is sufficient!
- An initial mesh  $\hat{M} = M^n$  can be simplified into a coarser mesh  $M^0$  by a sequence of edge collapse.

$$(\hat{M} = M^n) \xrightarrow{ecol_{n-1}} \dots \xrightarrow{ecol_1} M^1 \xrightarrow{ecol_{n-1}} M^0$$

- The sequence is invertible



# Progressive Mesh



$$(\hat{M} = M^n) \xrightarrow{ecol_{n-1}} \dots \xrightarrow{ecol_1} M^1 \xrightarrow{ecol_{n-1}} M^0$$



## Energy Function

- Old  
 $E(K, V) = E_{\text{dist}}(K, V) + E_{\text{rep}}(K) + E_{\text{spring}}(K, V)$
- New  
 $E(M) = E_{\text{dist}}(M) + E_{\text{spring}}(M) + E_{\text{scalar}}(M) + E_{\text{discrete}}(M)$  where  $M = (K, V, D, S)$
- $E_{\text{rep}}$  goes away! And so does the parameter  $c_{\text{rep}}$

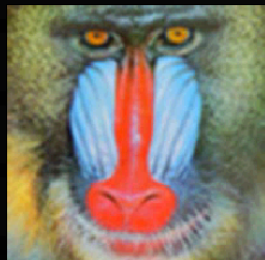
## Preserving geometry

- $E_{\text{dist}} + E_{\text{spring}}$
- Same as before, but only consider edge collapse in the outer loop.
- The possible legal collapses are placed in a priority queue with its estimated energy change  $\Delta E$
- For an edge collapse  $K \rightarrow K'$ ,  $\Delta E = E_{K'} - E_K$
- After each collapse, the cost of its neighborhood edges in the priority queue are updated.

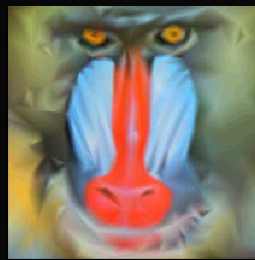
## Preserving Scalar attributes

- For  $d$  scalar attributes, we could have added  $d$  dimension to  $E_{\text{dist}}$ .
- For efficiency purpose, we separate it as a different term  $E_{\text{scalar}}(\underline{\mathbf{V}}) = (c_{\text{scalar}})^2 \sum_{i=1}^n \|\underline{\mathbf{x}}_i - \phi_{\underline{\mathbf{V}}}(\mathbf{b}_i)\|^2$
- First solve for the vertex positions. Then using the same  $\mathbf{b}_i$  to find vertex attributes minimizing  $E_{\text{scalar}}$  by linear least square!

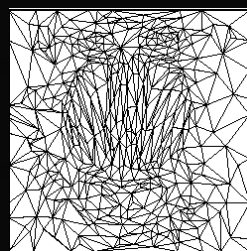
## Preserving Scalar Attributes



200x200 vertices

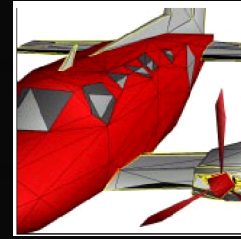


400 vertices



## Preserving discontinuity

- We want to preserve discontinuity because they often form noticeable features
- sample an additional set of points  $X_{disc}$  from sharp edges of initial mesh. Compute  $E_{disc}$  by projecting  $X_{disc}$  onto the corresponding sharp edges
- disallow/penalize collapse of boundary and discontinuity edges



Without  $E_{disc}$

## References

- Survey of Polygonal Surface Simplification Algorithms, Paul Heckbert and Michael Garland, tech. report, CS Dept., Carnegie Mellon U., to appear
- H. Hoppe, T. DeRose, T. Duchamp, J. McDonald, and W. Stuetzle, Mesh Optimization, SIGGRAPH 93, 19-26.
- Hugues Hoppe, Progressive Meshes, SIGGRAPH 96, 99-108.
- Hugues Hoppe, Efficient implementation of progressive meshes, Computers & Graphics, Vol. 22, No. 1, 1998, pages 27-36.
- Greg Turk, Re-Tiling Polygonal Surfaces, SIGGRAPH 92, 55-64.
- William J. Schroeder, Jonathan Zarge, and William Lorensen, Decimation of Triangle Meshes, SIGGRAPH 92, 65-70.
- J. Rossignac and P. Borrel, Multi-resolution 3D approximations for rendering complex scenes