## Multi-level Partition of Unity Implicit

#### SIGGRAPH 2003



### Outline

- Introduction
  - Problem
  - Issues

#### Algorithm

- Partition of Unity
- □ Adaptive Octree Subdivision
- Estimating Local Shape Functions
- Applications
- Performance
- Conclusion



#### Introduction - Problem

- Goal: Representing implicit solid as a function *f*
- Input: Points with Normals (typical output of range scanners)



f(x,y,z) > 0 inside f(x,y,z) < 0 outside f(x,y,z) = 0approximates points

Signed distance.

#### Introduction - Problem

Issues
 Local vs. Global

 (N/k of k x k) vs. (N x N)

 Sharp Features
 Error Control

[Turk and O'brien 2002.]

 $f(\mathbf{x}) = \sum w_j \phi(\mathbf{x} - \mathbf{c}_j) + P(\mathbf{x})$ 

j=1



Weighted average of local approximations





### Approach of this paper

 $f(\mathbf{x}) = \sum w_i(\mathbf{x}) Q_i(\mathbf{x}) / \sum w_i(\mathbf{x})$ 

- Piecewise quadratic local approximations Q<sub>i</sub>(x).
- Partition of unity  $\left\{ w_i(\mathbf{x}) / \sum w_i(\mathbf{x}) \right\}$ 
  - Used to blend local approximations
- Octree based multi-level structure
  - Adapted to geometrical complexity
  - Delivers an adaptive approximation of the distance-function
  - Allows a user to specify approximation accuracy





#### Octree and ball

#### Balls proportional to cell size, center at c<sub>i</sub>



#### Multi-level Partition of Unity

#### Adaptive octree subdivision



### Algorithm of MPU

- For cell i,
  - $\Box$  Fit  $Q_i$
  - $\Box$  Calculate  $\varepsilon_i$
  - $\Box$  If ( $\varepsilon_i > \varepsilon_0$ ) subdivide the cell and re-compute
- Blending (assembling) all leaf Q<sub>i</sub> 's using w<sub>i</sub>'s
  - $\Box f(\mathbf{x}) = \sum w_i(\mathbf{x}) Q_i(\mathbf{x}) / \sum w_i(\mathbf{x})$
- Remaining problem: choices of  $Q_i$  and  $w_i$

### Weight functions

- For approximation B-spline b(t)  $w_i(\mathbf{x}) = b\left(\frac{3|\mathbf{x} - \mathbf{c}_i|}{2R_i}\right)$
- For interpolation

Inverse-distance singular weights

$$w_i(\mathbf{x}) = \left[\frac{\left(R_i - |\mathbf{x} - \mathbf{c}_i|\right)_+}{R_i |\mathbf{x} - \mathbf{c}_i|}\right]^2, \text{ where } (a)_+ = \begin{cases} a & \text{if } a > 0\\ 0 & \text{otherwise} \end{cases}$$
  

$$\Rightarrow \text{infinity near } \mathbf{c}_i$$

### Local Shape Function (Q<sub>i</sub>)

- Second-order polynomial approx. by least square fitting
- Approximation type: according to the deviation of normals.



#### Local Shape Function (Q<sub>i</sub>)

Expand balls to include sufficient number of points.



#### **Sharp Features**

• Quadrics  $\rightarrow$  Impossible to represent sharp features





#### **Sharp Features**

Use piecewise smooth local approximations



#### **Sharp Features**

- Edge: most deviated n<sub>1</sub>, n<sub>2</sub>
   Corport bigbly deviated from m
- Corner: highly deviated from  $n_3 = n_1 \times n_2$



max/min Boolean -operations → piecewise smooth

local approximations

Clustering Normals  $\rightarrow$  Clustring Points

#### Accuracy Control

## For visualization purposes 0.01% accuracy is sufficient.



Original mesh (David head 1mm) Approximation by MPU with 0.01% accuracy

# Applications – Geometric operations





**Boolean operations** 

Space transformation

## Applications - offsettingIf *f* is a good approx. of signed distance.



#### **Applications - Blending**





## Applications - Morphing $f(\mathbf{x}) - (1-t) f_1(\mathbf{x}) + t f_1(\mathbf{x})$





 $f_1(\mathbf{x}) = 0$ 







 $f_2(\mathbf{x}) = 0$ 

#### Applications – Filling, Smoothing

No topological restrictions









#### Performance



#### Performance

#### with 0.08% accuracy [Carr et al. SIG01] MPU **RAM: 195 MB** ×100 **RAM: 306MB** Time: 99 sec. →Time: 170 min. (Pentium4 1.6 GHz) (Pentium3 550 MHz) $\times 3$

Reconstruction

#### Conclusion

- A new implicit representation for 3D scattered point data
  - Easy to implement
  - Fast reconstruction
  - Can handle a very large data
  - □ Can represent sharp features
  - □ Good for function-based modeling