Geometric Modeling and Visualization http://www.cs.utexas.edu/~bajaj/cs384R08/



Lecture 6 Structure Reconstruction: Shape Segmentation



Computational Visualization Center Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Patient Specific Heart Models



Model Reconstruction from Regularized Voxel Centers (Point Clouds)



Problem Description – Surface and Medial Axis construction



Scanning

Point Sample with closeup

Reconstruction with closeup

Surface Reconstruction Problem:

Given a set of points P sampled from S, create an approximation of S and an approximation of the medial axis of S.

Medial Axis Transform (MAT) Problem:

Given a set of points P sampled from S, create an approximation of M of the medial axis of S.



University of Texas at Austin

Medial Axis and Sampling Density

Medial Axis M of a shape S is defined as a set of points which has more than one nearest point on S.

2D Illustration

s a a b b



Distance Function h_s assigns every point *x* the nearest distance to *S*.

Approximation of h_S is done via h_P when S is known only via a finite set of points P on S.

$$h_P : \mathbb{R}^3 \to \mathbb{R}, \ x \mapsto \min_{p \in P} \|x - p\|$$

Local Feature Size: Distance of a point x on surface to the medial axis. Denoted as f(x).

<u> ε -sampling</u>: Every point on the surface has a sample point p within $\varepsilon f(x)$.



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences University of

University of Texas at Austin

Computational Aspects: Vor/Del of P

Voronoi diagram

$$V_p = \{ x \in \mathbb{R}^3 : \forall q \in P - \{p\}, \|x - p\| \le \|x - q\| \} \}$$

Delaunay triangulation

Dual of Voronoi diagram.



Restricted Voronoi-Delaunay





Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Theory

What is a correct reconstruction?

The reconstructed surface is homeomorphic to the original surface and within a small Hausdorff distance.

Theorem [ES1994]

If the Voronoi diagram of sample points satisfies *ball-property*, the restricted Delaunay triangulation is *homeomorphic* to the surface.

What relation does Medial Axis have with the object? [Lieutier2004]

The medial axis is *homotopy equivalent* to the surface.

Ball Property: A k-dimensional Voronoi face intersects the surface in k-1 dimensional ball





Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Cocone Family - Cocone, TightCocone, RobustCocone and Medial

Cocone:

Given a set of points P sampled from S, create an approximation of S. Additionally detects the undersampling present in the data. [Amenta, Choi, Dey, Leekha 2002 – IJCGA]

TightCocone:

Given a set of points P sampled from S, create a watertight approximation of S. [Dey, Goswami 2003 – ACM Solid Modeling]

RobustCocone:

Given a set of (noisy) points P sampled from or near S, create a watertight approximation of S. [Dev, Goswami 2004 – ACM SoCG]

Medial Axis Transform:

Given a set of points P sampled from S, approximate the medial axis of S. [Dey, Zhao 2003 – Algorithmica]



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Cocone with Undersampling Detection

<u>Pole:</u> Furthest Voronoi vertex in the Voronoi cell of a point p.

Vector v_p denotes the vector from p to the pole. This estimates the normal at p.

Cocone: The set $C_p = \{y \in V_p : \angle((y-p), \mathbf{v}_p) \geq \frac{3\pi}{8}\}$ is called the cocone of p.

Cocone estimates the tangent plane.

- 1. Compute Delaunay of P
- 2. Compute *pole* at every p. (Poles approximate normals)
- 3. Select set of Delaunay triangles incident at p that fall within Cocone of p. Call them *candidate* triangles.
- 4. Extract the surface from the *candidate* triangles by collecting the outer layer.

<u>Theorem</u>: Reconstructed surface is *homeomorphic* to *S* if *P* is an ε sample of *S* with ε < 0.06.

Results



Cocone

Pole



Algorithm:



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Undersampling - TightCocone



Marking : Any point whose neighborhood on the initial reconstructed surface is a disk, separates the incident tetrahedra into *IN* and *OUT*.

Goal is to propagate the marking consistently.



<u>Problem</u> : Medial Axis passes through sharp corners. Feature size = 0. Requires infinite sampling to satisfy sampling density condition.

<u>Peeling</u> : Starting from convex hull, ``walk" among the *OUT* tetrahedra to sculpt them away leaving the set of *IN* tetrahedra whose boundary is the surface.

<u>Algorithm</u>

- 1. Compute initial surface using Cocone.
- 2. <u>Marking:</u> Around every point with manifold neighborhood, mark the tetrahedra *IN* and *OUT*.
- 3. <u>Peeling:</u> Starting from convex hull, *peel* all *OUT* tetrahedra.



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Results



Point Cloud Reconstruction Examples - BioMed



Department of Computer Sciences

University of Texas at Austin

Medial Axis Approximation

<u>Umbrella</u>: The plane normal to the pole vector intersects some of the Voronoi edges. Dual Del triangles to those edges form the Umbrella at p.

<u>Angle Condition (θ)</u>: Edge *pq* satisfies angle condition if it makes large angle with all the umbrella triangles at *p* and *q*.

<u>Ratio Condition (ρ)</u>: Length of *pq* is larger than (ρ times) the circumradius of the umbrella triangles at p and q.

<u>Algorithm</u>

(a`

- 1. Compute Vor-Del of P.
- 2. Collect all the Voronoi facets whose dual Delaunay edges satisfy both *angle* and *ratio* condition.

(b)

<u>Results</u>





Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin





Step 1: Vor/Del(P) computation (using CGAL).

Goswami, Dey, Bajaj, ICVGIP'2006

Step 2: Identification of Interior Medial Axis M.

Step 3:

- 3.a: Identification of Critical points of distance function from Vor/Del(P).
- 3.b: Selection of Critical points only on M.

[By DGRS2005, Critical points are either near S or near M]

Step 4: Classification of Medial Axis via

4.a: U₁ – <u>Unstable Manifold of index 1 saddle</u> point on M

4.b: $U_2 - \underline{\text{Unstable Manifold of index 2 saddle point on M}}$.

Step 5: Width Test to select the subsets of U₁ (β -sheets) and U₂ (α -helices).



January 2008

Critical Points, their Indices, and their Manifolds

Critical Point of a smooth function is a point where the gradient of the function vanishes.

Index of a critical point is the number of independent directions in which the function decreases.

In 3D, four types of critical points

- 1. Minima index 0
- 2. Saddle of index 1
- 3. Saddle of index 2
- 4. Maxima index 3



From EHNP SoCG'03

Integral curve : A path in the domain of the function on which at every point the tangent to the curve equals the gradient of the function.

Stable Manifold of a critical point is the union of all integral curves ending at the critical point.

Unstable Manifold of a critical point is the union of all integral curves starting at the critical point.



Vor/Del of P and Critical Points of h_P

Voronoi diagram

$$V_p = \{ x \in \mathbb{R}^3 : \forall q \in P - \{p\}, \|x - p\| \le \|x - q\| \} \}$$

Delaunay triangulation

Dual of Voronoi diagram.

Critical points of h_P can be computed from the Vor/Del(P). These are the intersection of Voronoi and their dual Delaunay objects.

> Minimum \ominus Saddle Maximum \oplus Siersma et al. Center for Computational Visualization http://www.ices.utexas.edu/CCV Institute of Computational and Engineering Sciences **Department of Computer Sciences**





University of Texas at Austin



Computation of U_1

• Start from the VF containing an index-1 saddle point.

• Generic step: Identify the VE s on the boundary which are active. Expand the unstable manifold by collecting their acceptor VF s.







Center for Computational Visualization http://www.ices.utexas.edu/CCV Institute of Computational and Engineering Sciences Department of Computer Sciences University of Texas at Austin





Step 1: Vor/Del(P) computation (using CGAL).

Step 2: Identification of Interior Medial Axis M

Step 3:

3.a: Identification of Critical points of h_P from Vor/Del(P).

3.b: Selection of Critical points only on M.

[By DGRS2005, Critical points are either near S or near M

Step 4: Decomposition of shape via S3 stable manifold of maxima on M

Step 5: Width Test to select the subsets of S3.



Center for Computational Visualization http://www.ices.utexas.edu/CCV Institute of Computational and Engineering Sciences Department of Computer Sciences University of Texas at Austin



The tertiary fold of 1AOR is a β -sandwich (two red sheets), which is surrounded by the differently colored helical segments.



The tertiary fold of 1TIM is .a α/β -barrel.

The β -region in the middle is segmented as red while the helical segments surrounding it are colored differently.



Center for Computational Visualizationhttp://www.ices.utexas.edu/CCVInstitute of Computational and Engineering SciencesDepartment of Computer SciencesUniversity of Texas at Austin

Modeling Human Joint Dynamics and Stress



Department of Computational and Enginee

University of Texas at Austin



The conic curve interpolant is the zero of the bivariate quadratic polynomial interpolant over the triangle



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

A-spline segment over BB basis





Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Examples of Discriminating Curve Families







Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin



For a given discriminating family D(R, R₁, R₂), let f(x, y) be a bivariate polynomial . If the curve f(x, y) = 0 intersects with each curve in D(R, R₁, R₂) only once in the interior of R, we say the curve f = 0 is regular(or A-spline segment) with respect to D(R, R₁, R₂).

If $B_0(s)$, $B_1(s)$, ... has one sign change, then the curve is

- (a) D_1 regular curve.
- (b) D_2 regular curve.
- (c) D_3 regular curve.
- (d) D_4 regular curve.



University of Texas at Austin





A-patch Surface (C¹) Interpolant



 An implicit single-sheeted polynomial interpolant on a tetrahedron



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

A-Patches

• Given tetrahedron vertices $p_i=(x_i, y_i, z_i)$, i=1,2,3,4, α is barycentric coordinates of p=(x,y,z):



• function *f(p)* of degree *n* can be expressed in Bernstein-Bezier form :

$$f(p) = \sum_{|\lambda|=n} b_{\lambda} B_{\lambda}^{n}(\alpha), \ \lambda \in \mathbb{Z}_{+}^{4} \quad B_{\lambda}^{n}(\alpha) = \frac{n!}{\lambda_{1}!\lambda_{2}!\lambda_{3}!\lambda_{4}!} \alpha_{1}^{\lambda_{1}}\alpha_{2}^{\lambda_{2}}\alpha_{3}^{\lambda_{3}}\alpha_{4}^{\lambda_{4}}$$

• Algebraic surface patch(A-patch) within the tet is defined as f(p)=0.



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

October 2007

Algebraic Patches: Smooth Boundary Elements

• Implicit form of Isocontour : f(x,y,z) = w



A-patch Contouring













Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

October 2007

A-patches [Bajaj, Chen, Xu 1994]

- -Zero contour of a trivariate polynomial: f(x,y,z)=0
- -Single-sheeted patches
- -C¹ smoothing of arbitrary polyhedra
- Conversion to Trimmed NURBS







Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Computing functions on surface

- Surface: Weighted least squares approximation of data points and signeddistance samples
- -Function:Weighted least squares approximation of function values





A-Patch Approximation

~9200 points, 406 patches (degree 3), 1% error





Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Automatic CAD Model (features) Reconstruction from Point Clouds



Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Sculpturing



 α -solid is refined by iteratively removing tetrahedra adjacent to the boundary, based on two principles:

- remove if a data point is occluded
- remove if sum of dihedral angles decreases

University of Texas at Austin



Dihedral angles formed by boundary faces

Mesh reduction



• Mesh reduction technique for triangle meshes with multivariate data



- Based on incremental deletion of vertices and retriangulation
- Guaranteed, global error-bound
- Sharp feature recognition and preservation



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin



Singularities



Center for Computational Visualization Institute of Computational and Engineering Sciences Department of Computer Sciences

University of Texas at Austin

Further Reading: Point-set Surface Reconstruction

- 1. Boissonnat, J. D., 1984, "Geometric Structures for Three Dimensional Shape Representation," ACM Transact. on Graphics, 3(4), pp 266–286.
- 2. Edelsbrunner, H., and Mucke, E. P., 1994, "Three-dimensional Alpha Shapes," ACM Trans. Graphics, 13, pp 43–72.
- 3. Bajaj, C., Bernardini, F., and Xu, G., 1995, "Automatic Reconstruction of Surfaces and Scalar Fields from 3D Scans," ACM SIGGRAPH, pp. 109–118.
- 4. Amenta, N., Bern, M., and Kamvysselis, M., 1998, "A New Voronoi-based Surface Reconstruction Algorithm," ACM SIGGRAPH, pp 415-421.
- 5. Amenta, N., Choi, S., Dey, T. K., and Leekha, N., 2002, "A Simple Algorithm for Homeomorphic Surface Reconstruction," Internat. J. Comput. Geom. & Applications, 12, pp. 125-141.
- 6. Boissonnat, J. D., and Cazals, F., 2000, "Smooth Surface Reconstruction via Natural Neighbor Interpolation of Distance Functions," Proc. 16th. Annu. Sympos. Comput. Geom., pp 223–232.
- 7. Amenta, N., Choi. S., and Kolluri, R. K., 2001, "The Power Crust," Proc. 6th Annu. Sympos. Solid Modeling Applications, pp. 249-260.
- 8. Dey T. K. and Goswami S., 2003, "TightCocone: A Watertight Surface Reconstructor," Proc. 8th Annu. Sympos. Solid Modeling Applications, pp. 127-134.
- 9. Dey T. K. and Goswami S., 2003, "Provable Surface Reconstruction from Noisy Samples," Proc. 20th Annu. Sympos. Comp. Geom., pp. 330-339.
- 10. Hoppe, H., DeRose, T., Duchamp, T., McDonald, J., and Stuetzle, W., 1992, "Surface Reconstruction from Unorganized Points," ACM SIGGRAPH 92, pp 71-78.
- 11. Curless, B., and Levoy, M., 1996, "A Volumetric Method for Building Complex Models from Range Images," ACM SIGGRAPH, pp 303-312.



University of Texas at Austin

Further Reading: Medial Axis Transform (MAT)

- H. Edelsbrunner and N. Shah. Triangulating topological spaces. Proc. 10th ACM Sympos. Comput. Geom., (1994), 285-292.
- T. K. Dey and W. Zhao. Approximate medial axis as a Voronoi subcomplex. Proc. 7th ACM Sympos. Solid Modeling Appl., 2002, 356--366.
- Nina Amenta and Ravi Krishna Kolluri, "The medial axis of a union of balls", Computational Geometry 2001" 20(1-2) pages "25-37",
- F. Chazal, A. Lieutier, The Lambda Medial Axis, Graphical Models, Volume 67, Issue 4, July 2005, Pages 304-331
- André Lieutier: Any open bounded subset of has the same homotopy type as its medial axis. Computer-Aided Design 36(11): 1029-1046 (2004)
- Joachim Giesen, Edgar A. Ramos, Bardia Sadri: Medial axis approximation and unstable flow complex. Symposium on Computational Geometry 2006: 327-336



Further Reading: A-splines & A-patches

- G. Xu, C. Bajaj, S. Evans
 C1 Modeling with Hybrid Multiple-sided A-patches
 Special issue on Surface and Volume Reconstructions in the International Journal of Foundations of Computer Science, 13, 2, 261-284.
- C. Bajaj, G. Xu, R. Holt, A. Netravali
 Hierarchical Multiresolution Reconstruction of Shell Surfaces Computer Aided Geometric Design, 19:2(2002), 89-112.
- G. Xu, C. Bajaj, H. Huang
 C1 Modeling with A-patches from Rational Trivariate Functions Computer Aided Geometric Design, 18:3(2001), 221-243.
- C. Bajaj, G. Xu
 Regular Algebraic Curve Segments (III) Applications in Interactive Design and Data Fitting Computer Aided Geometric Design, 18:3(2001), 149-173.
- 5. C. Bajaj, G. Xu
- 6. **Smooth Shell Construction with Mixed Prism Fat Surfaces** Brunett, G., Bieri,H., Farin, G. (eds.), Geometric Modeling Computing Supplement, 14, (2001), 19–35.
- G. Xu, C. Bajaj, W. Xue
 Regular algebraic curve segments (I)-Definitions and characteristics
 Computer Aided Geometric Design, 17:6(2000), 485-501.
- G. Xu, C. Bajaj, C. Chu
 Regular Algebraic Curve Segments (II) Interpolation and Approximation Computer Aided Geometric Design, 17:6(2000), 503-519
- 9. C. Bajaj, G. Xu
 A-Splines: Local Interpolation and Approximation Using Gk- Continuous Piecewise Real Algebraic Curves Computer Aided Geometric Design, 16:6(1999), 557-578.

