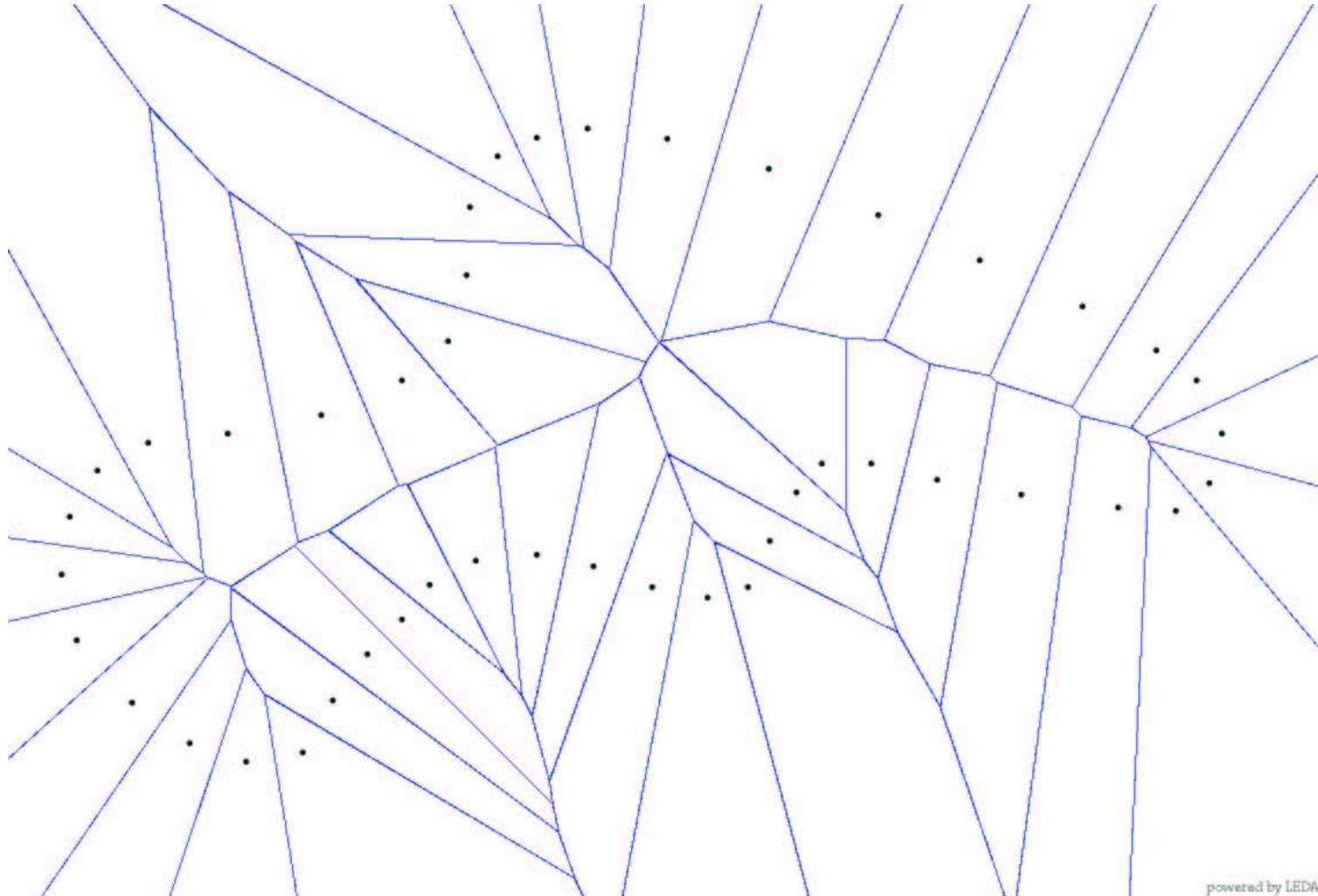

Estimating Geometry and Topology from Voronoi Diagrams

Tamal K. Dey

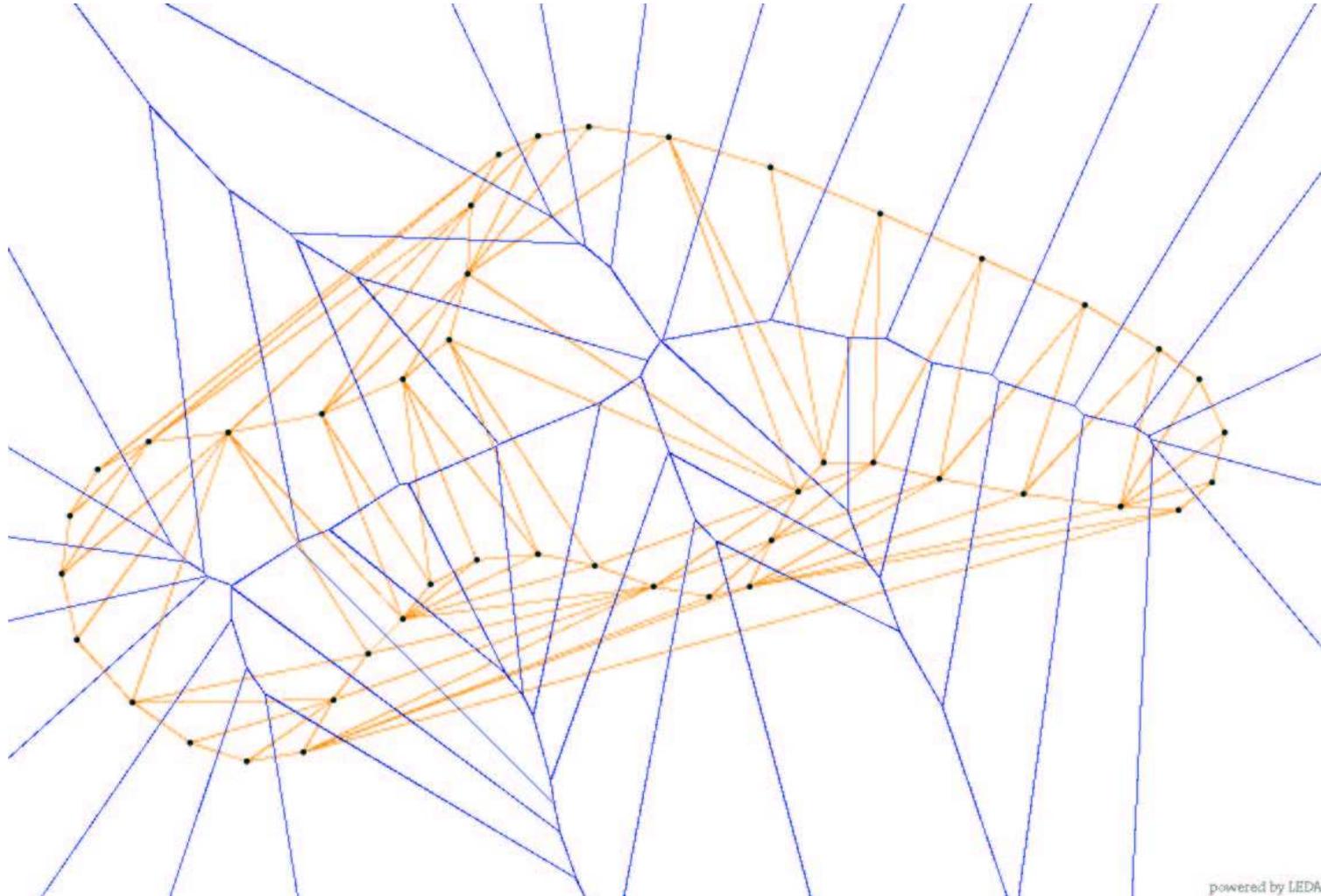
The Ohio State University



Voronoi diagrams



Voronoi diagrams



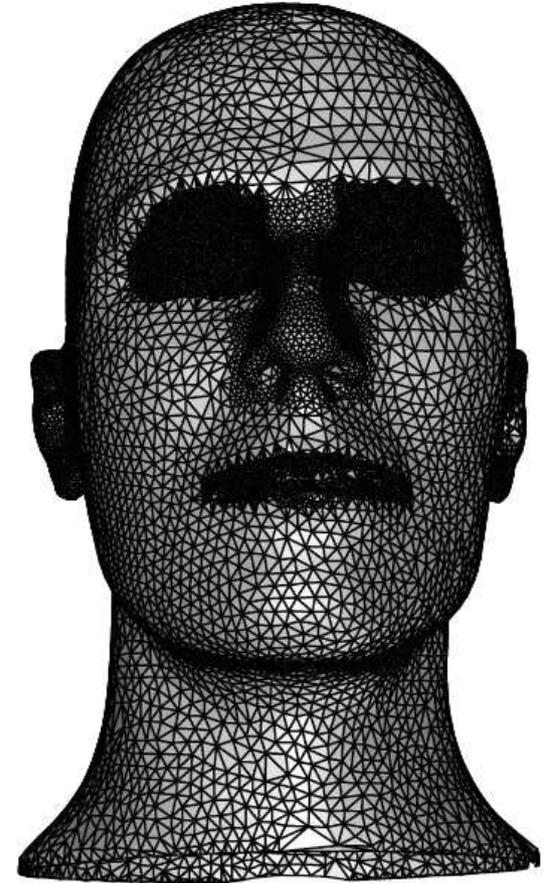
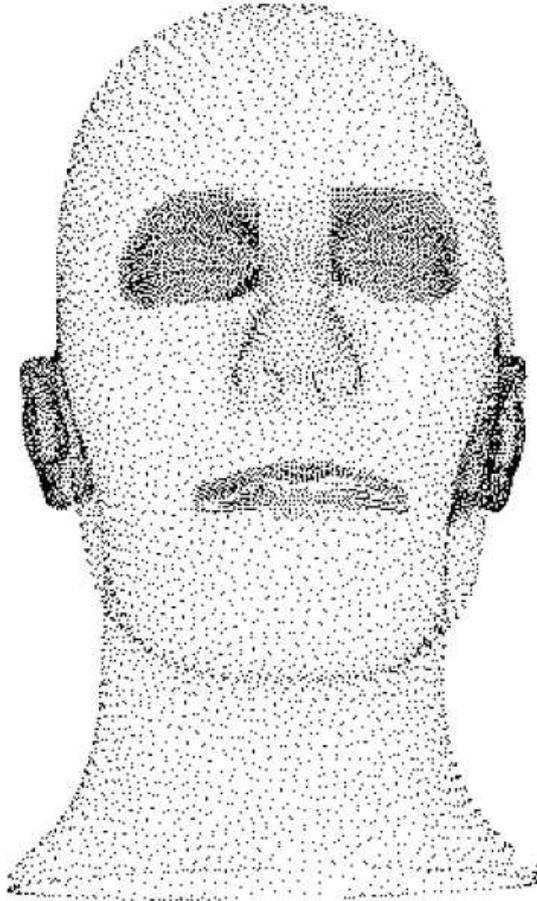
Problem

Estimating geometry: Given P presumably sampled from a k -dimensional manifold $\mathcal{M} \subset \mathbb{R}^d$ estimate geometric attributes such as normals, curvatures of \mathcal{M} from $V_{\text{or}} P$.

Estimating topology: (i) capture the major topological features (persistent topology) of \mathcal{M} from $V_{\text{or}} P$ (ii) capture the exact topology of \mathcal{M} from $V_{\text{or}} P$.



Three dimensions



Medial Axis and Local Feature Size

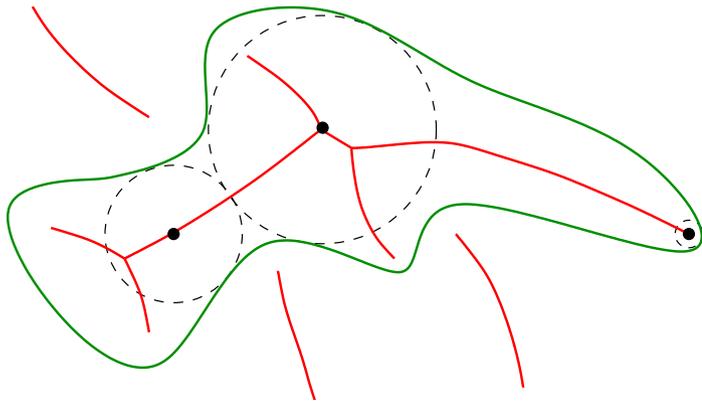
MEDIAL AXIS:

Set of centers of maximal empty balls.

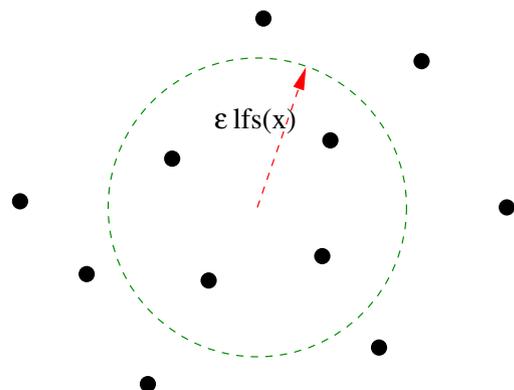
LOCAL FEATURE SIZE:

For $x \in \mathcal{M}$, $f(x)$ is the distance to the medial axis.

$$f(x) \leq f(y) + \|xy\| \quad \text{1-Lipschitz}$$



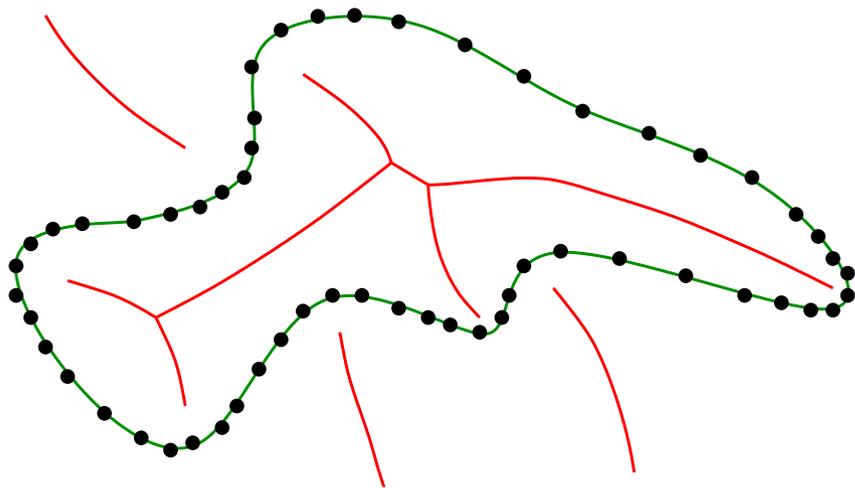
Good Sampling



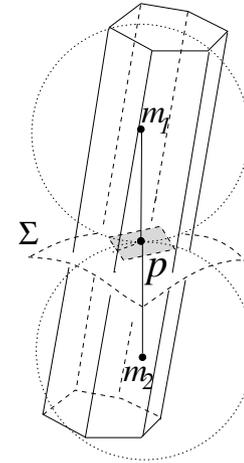
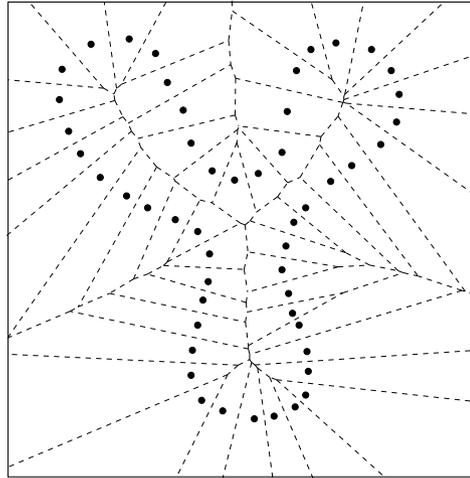
ϵ -SAMPLING [AMENTA-BERN-

EPPSTEIN 97]: $P \subset \mathcal{M}$ such
that

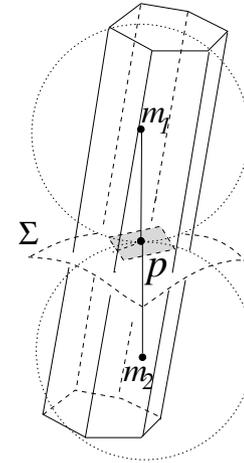
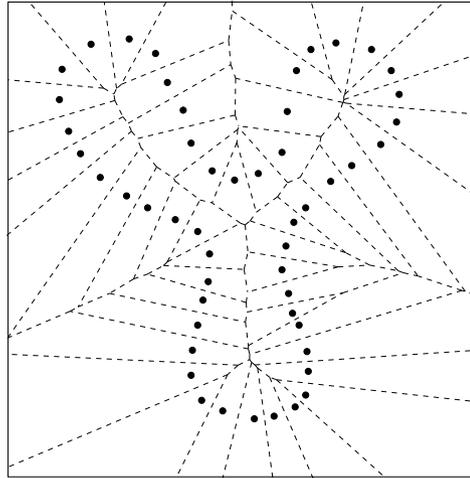
$$\forall x \in \mathcal{M}, \quad B(x, \epsilon \cdot f(x)) \cap P \neq \emptyset.$$



Normal estimation



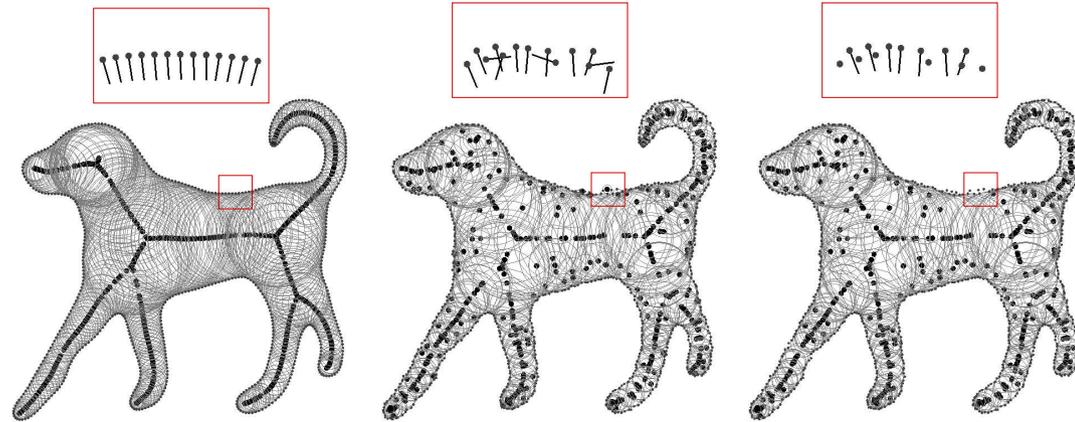
Normal estimation



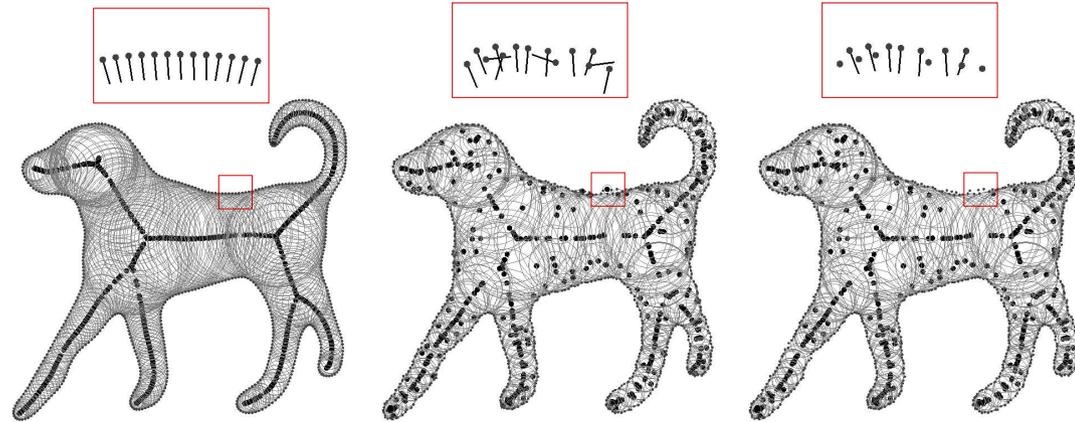
Normal Lemma [[Amenta-Bern 98](#)] : For $\varepsilon < 1$, the angle (acute) between the normal \mathbf{n}_p at p and the **pole vector** \mathbf{v}_p is at most

$$2 \arcsin \frac{\varepsilon}{1 - \varepsilon}.$$

Noise

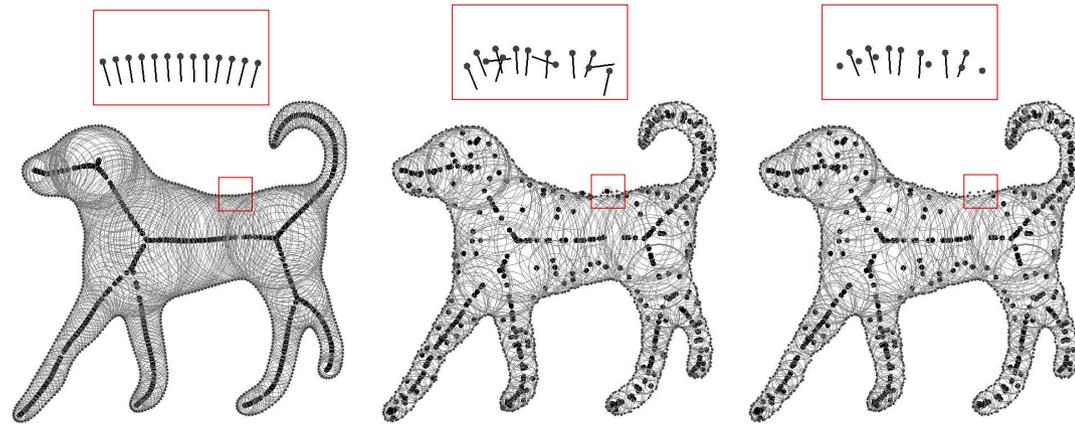


Noise

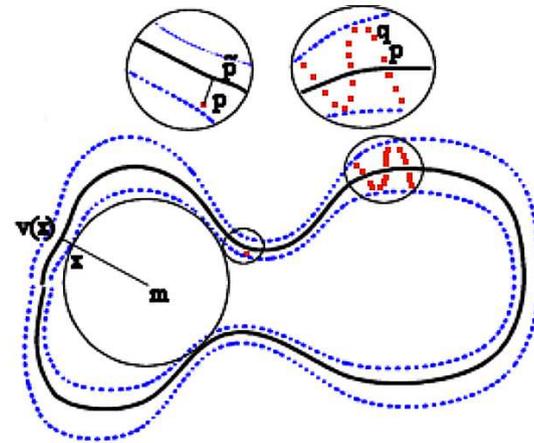


- \tilde{p} , \tilde{P} are orthogonal projections of p and P on \mathcal{M} .

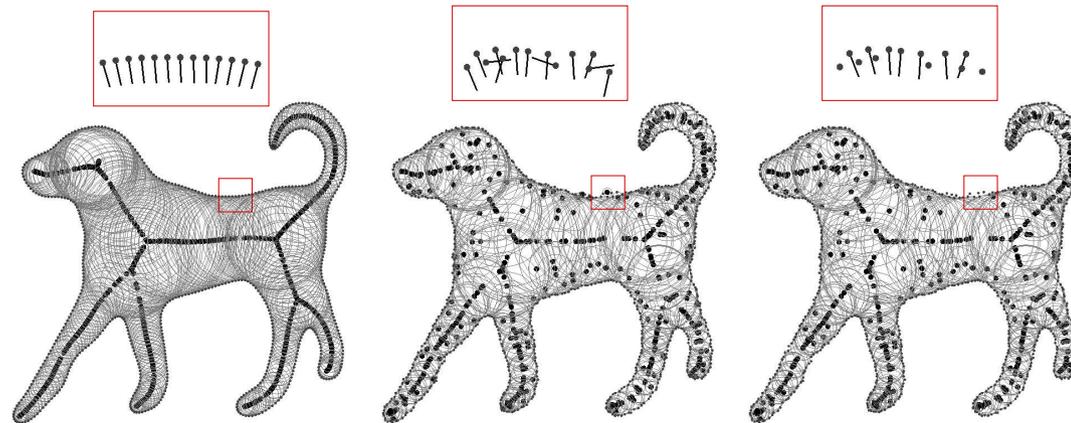
Noise



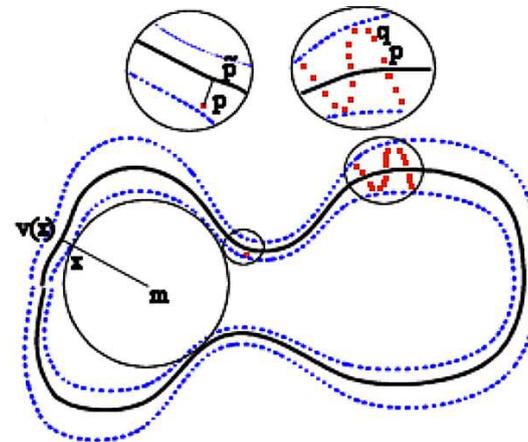
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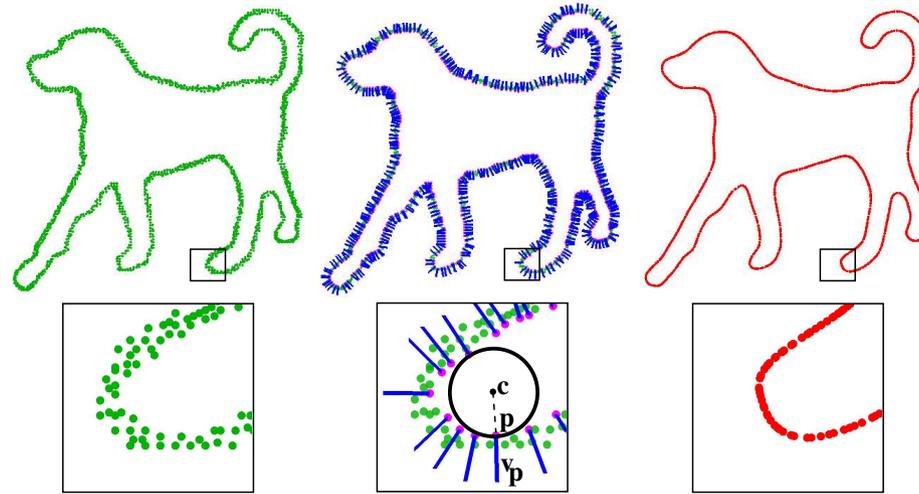
Noise



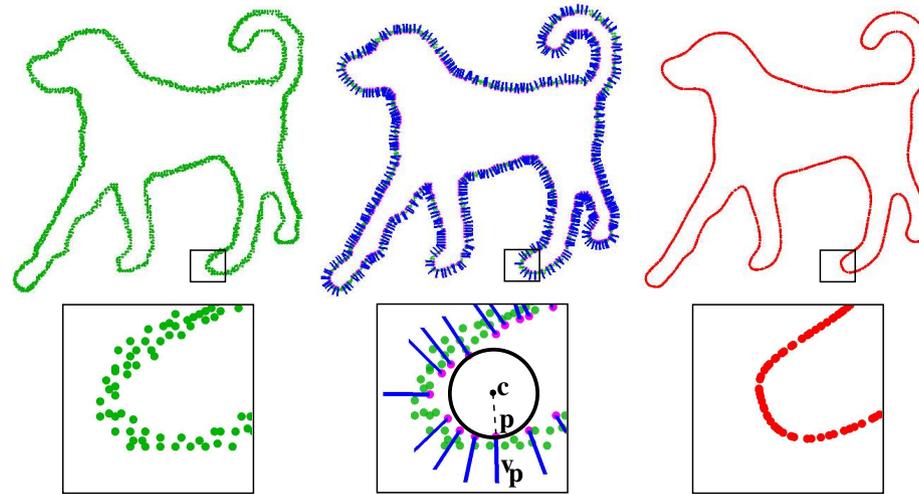
- \tilde{p} , \tilde{P} are orthogonal projections of p and P on \mathcal{M} .
- P is (ε, η) -sample of \mathcal{M} if
 - \tilde{P} is a ε -sample of \mathcal{M} ,
 - $d(p, \tilde{p}) \leq \eta f(\tilde{p})$.



Noise and normals



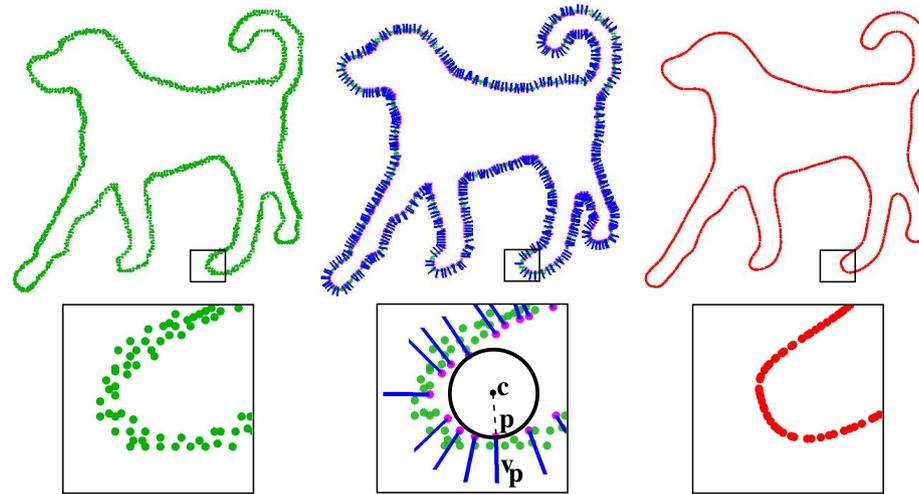
Noise and normals



Normal Lemma [[Dey-Sun 06](#)] : Let $p \in P$ with $d(p, \tilde{p}) \leq \eta f(\tilde{x})$ and $B_{c,r}$ be any Delaunay/Voronoi ball incident to p so that $r = \lambda f(\tilde{p})$. Then,

$$\angle cx, \mathbf{n}_{\tilde{x}} = O\left(\frac{\varepsilon}{\lambda} + \sqrt{\frac{\eta}{\lambda}}\right).$$

Noise and normals



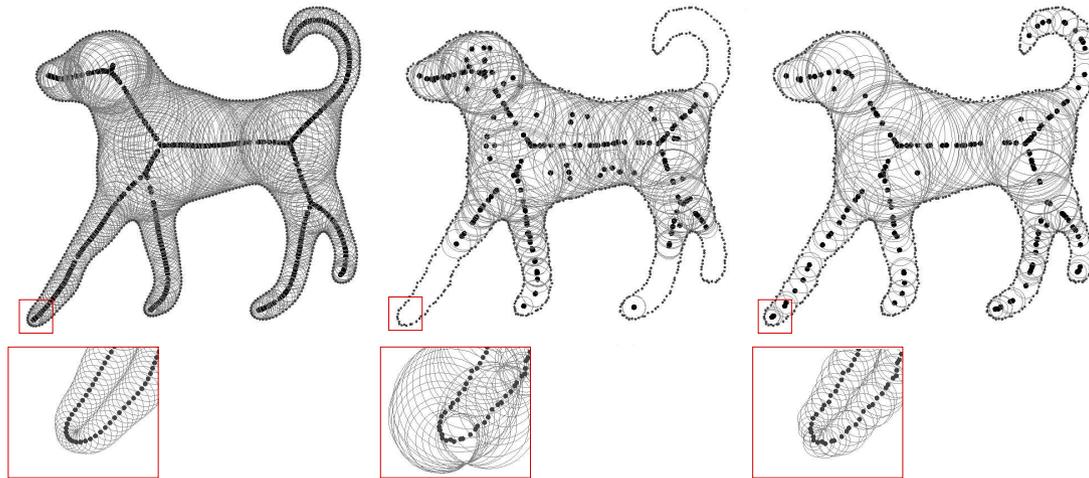
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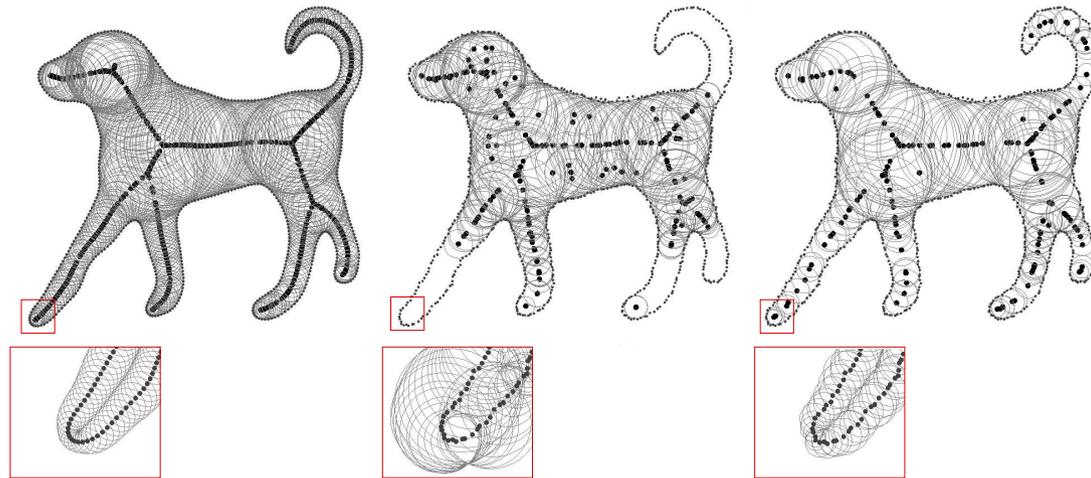
- Gives an [algorithm](#) to estimate normals.



Noise and features



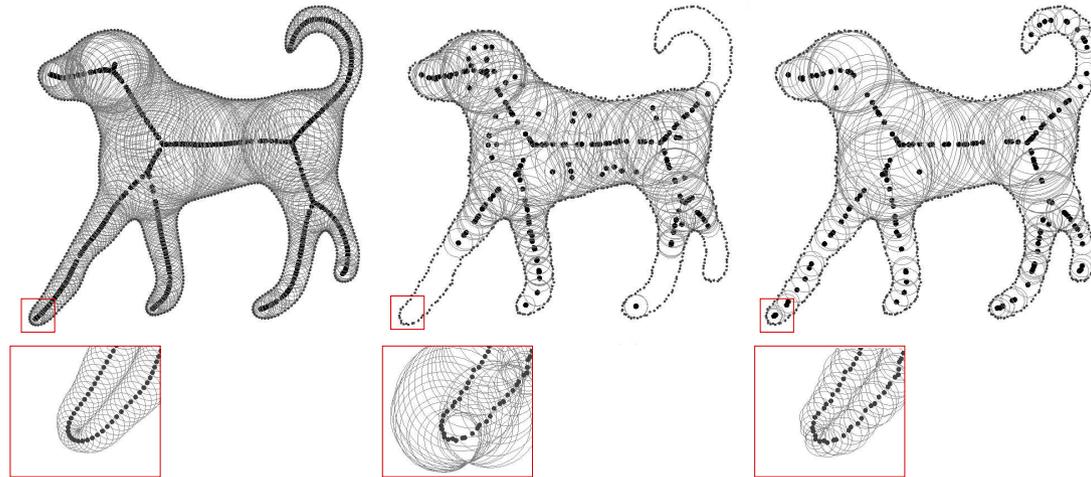
Noise and features



Medial Axis Approximation Lemma [Dey-Sun 06] : If P is a (ε, η) -sample of \mathcal{M} , then the **medial axis** of \mathcal{M} can be approximated with **Hausdorff distance** of $O(\varepsilon^{\frac{1}{4}} + \eta^{\frac{1}{4}})$ times the respective medial ball radii.



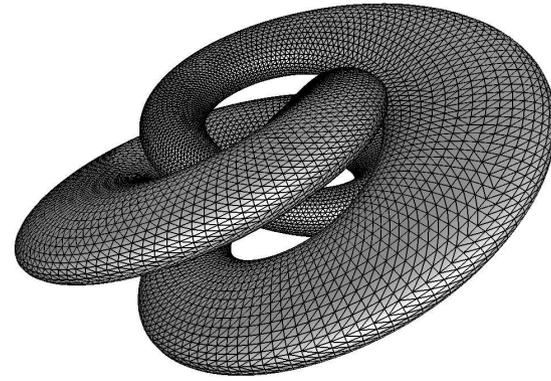
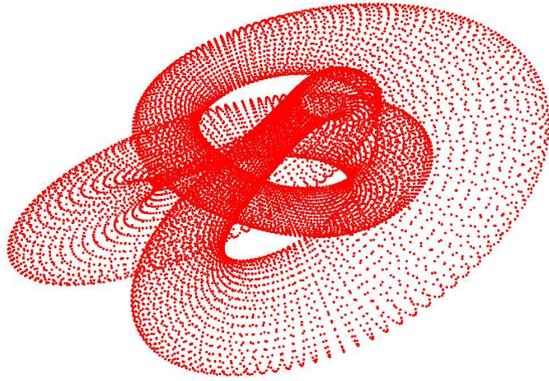
Noise and features



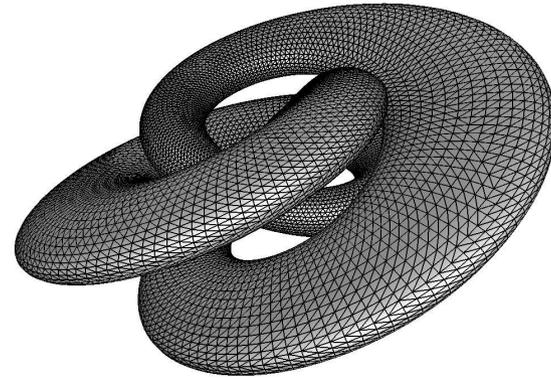
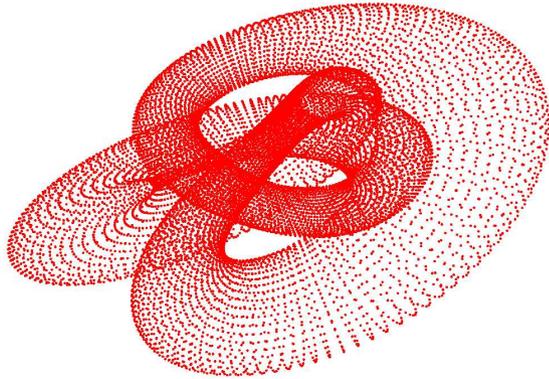
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- Gives an **algorithm** to estimate local feature size.

Topology

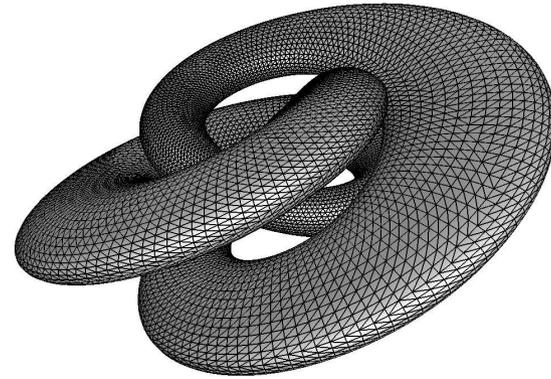
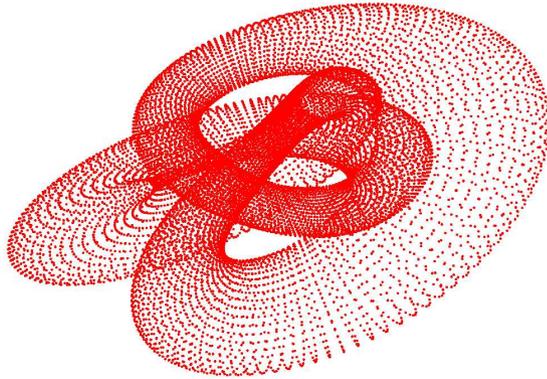


Topology



Homeomorphic/Isotopic reconstruction [[ACDL00](#)]: Let $P \subset \mathcal{M}$ be ε -sample. A Delaunay mesh $T \subset \text{Del } P$ can be computed so that

Topology



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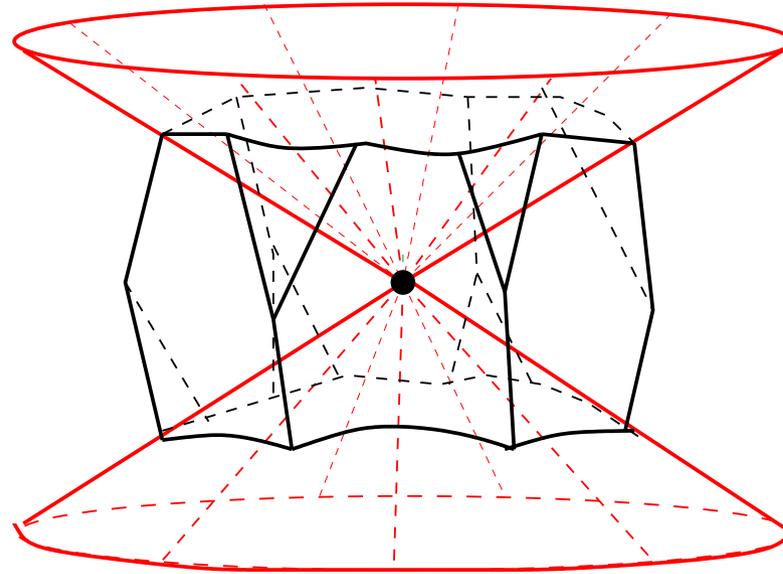
- there is an isotopy $h : |T| \times [0, 1] \rightarrow \mathbb{R}^3$ between $|T|$ and \mathcal{M} . Moreover, $h(|T|, 1)$ is the orthogonal projection map.
- the isotopy moves any point $x \in |T|$ only by $O(\varepsilon)f(\tilde{x})$ distance.
- triangles in T have normals within $O(\varepsilon)$ angle of the respective normals at the vertices.

Original Crust algorithm [[AB98](#)], Cocone algorithm [[ACDL00](#)], Natural neighbor algorithm [[BC00](#)] enjoy these properties.



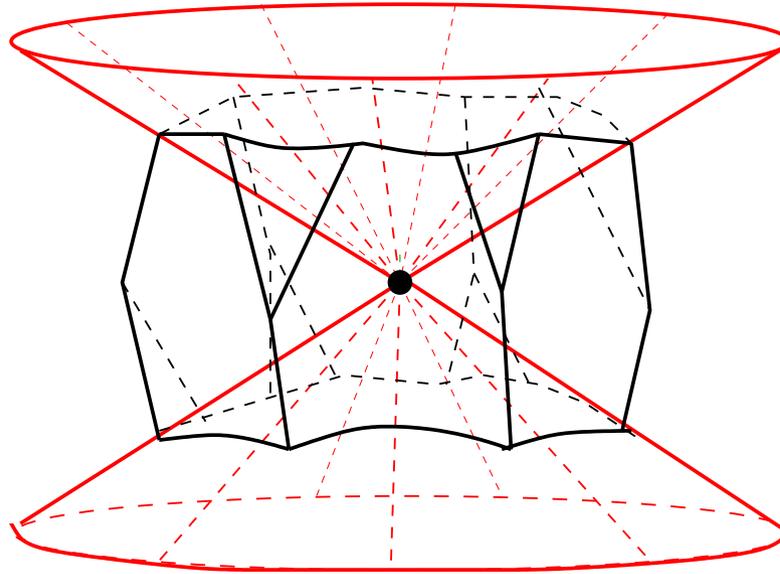
Cocone Algorithm

Amenta, Choi, Dey and Leekha



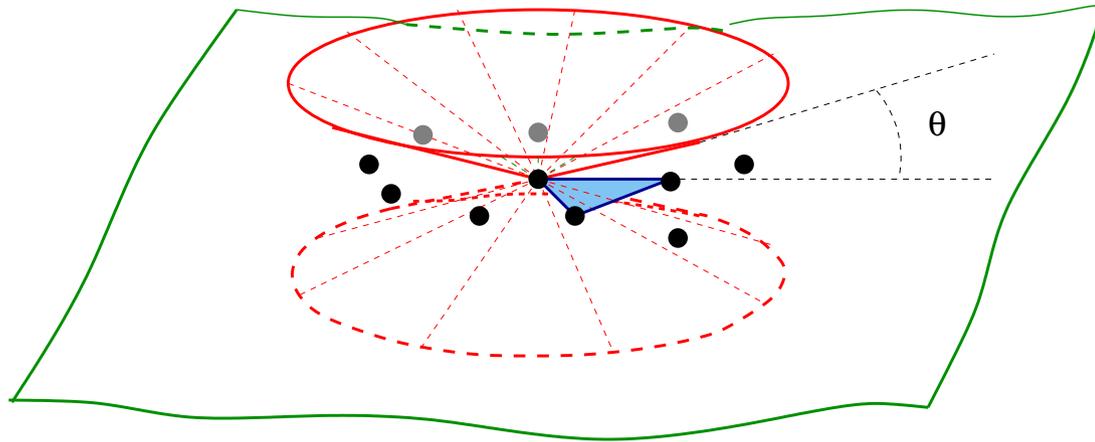
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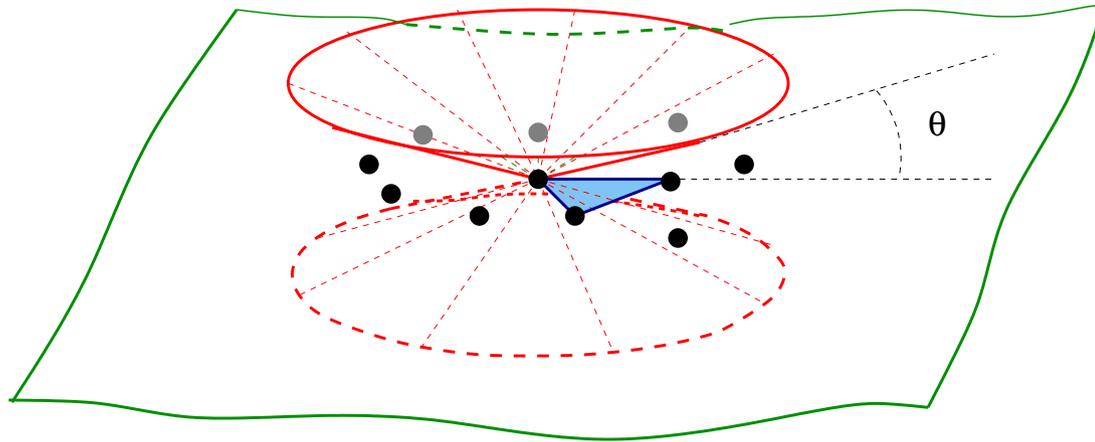


Cocone triangles for p : Delaunay triangles incident to p that are dual to Voronoi edges inside the cocone region.

Cocone Algorithm



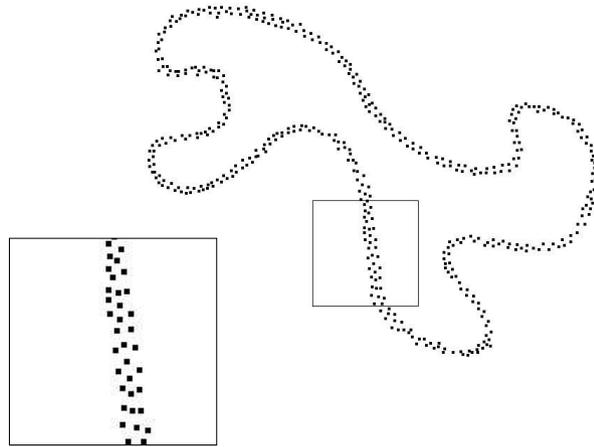
Cocone Algorithm



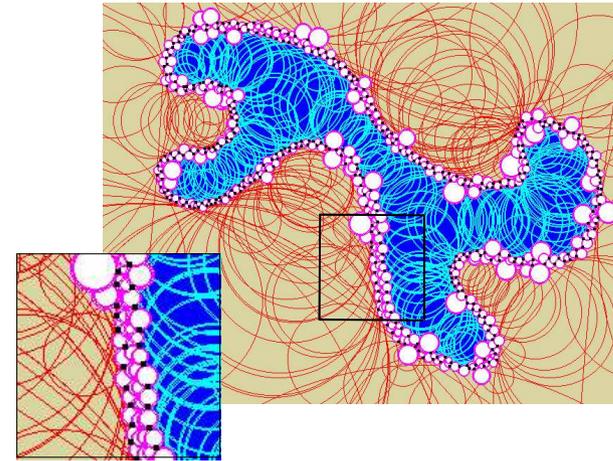
Cocone triangles :

- are nearly orthogonal to the estimated normal at p
- have empty spheres that are near equatorial.

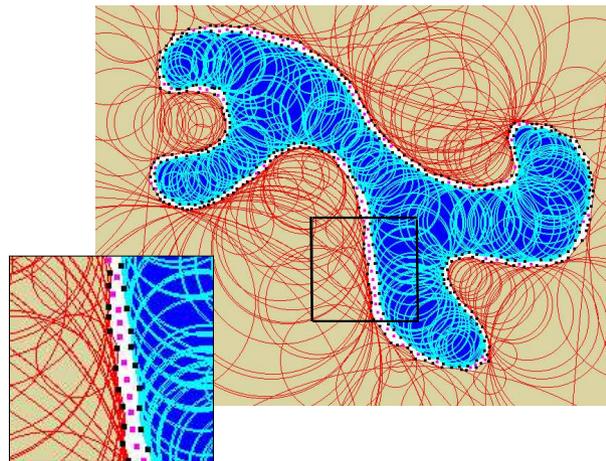
Noisy sample : Topology



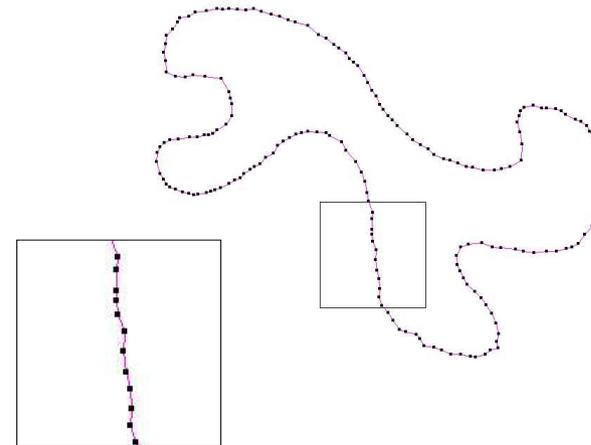
Input noisy sample



Step 1



Step 2



Step 3

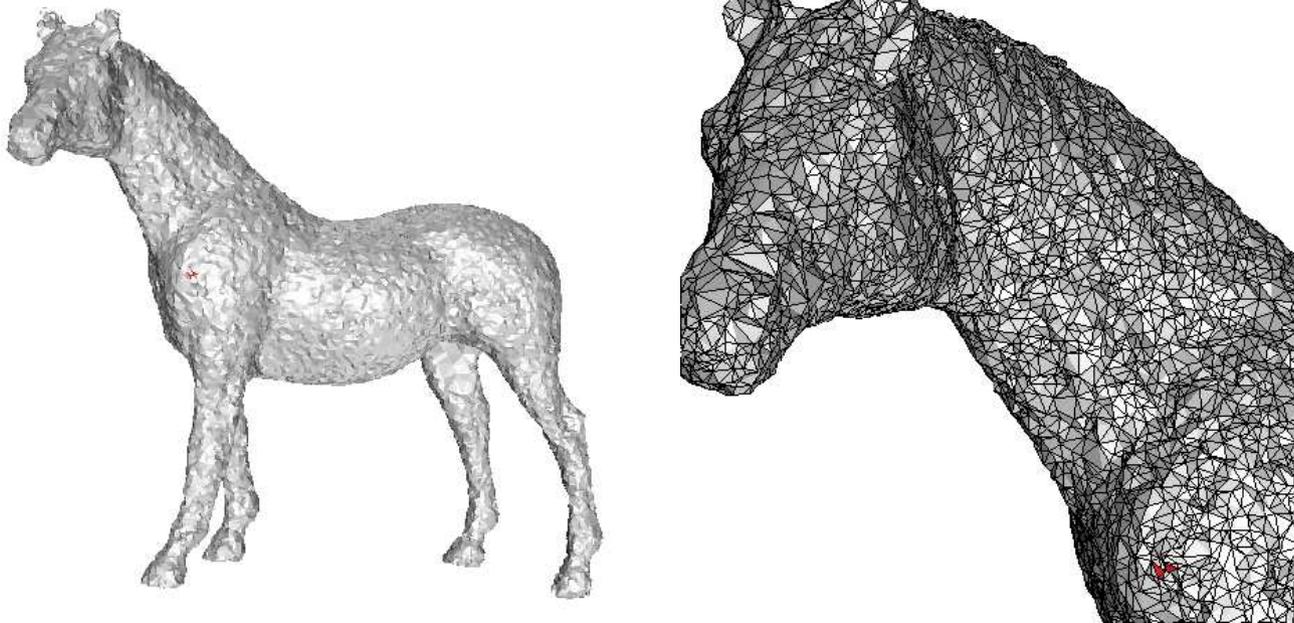
Noisy sample : topology

Homeomorphic reconstruction [[Dey-Goswami 04](#)] : Let $P \subset \mathcal{M}$ be $(\varepsilon, \varepsilon^2)$ -sample. For $\lambda > 0$, Let $B_\lambda = \{B_{c,r}\}$ be the set of (inner) Delaunay balls where $r > \lambda f(\tilde{c})$. There exists a $\lambda > 0$ so that $\text{bd} \bigcup B_\lambda \approx \mathcal{M}$.



Noisy sample : topology

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Higher dimensions

Assumptions: Sample P from a smooth, compact k -manifold $\mathcal{M} \subset \mathbb{R}^d$ without boundary. P is “sufficiently dense and uniform”.



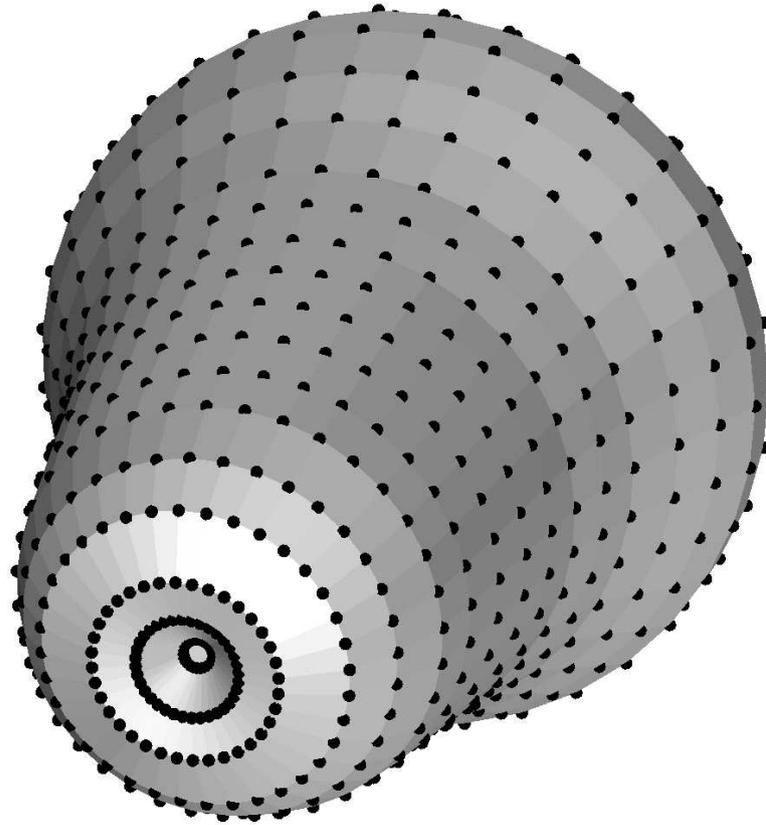
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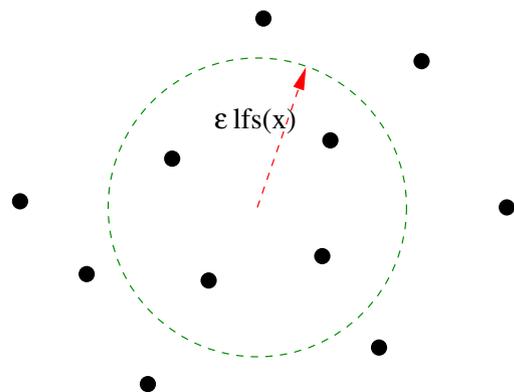
- Normal space estimation, dimension detection;
- Homeomorphic reconstruction



Good Sampling



Good Sampling

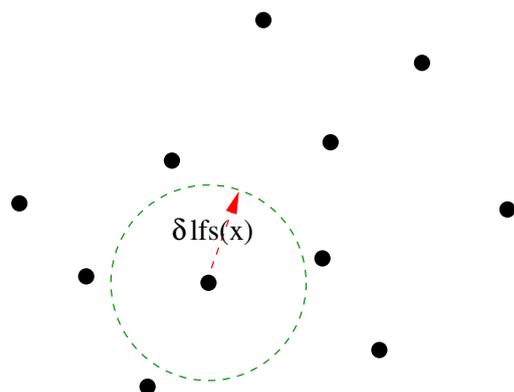


$$0 < \delta < \epsilon$$

(ϵ, δ) -SAMPLING: $P \subset \mathcal{M}$ such that

$$\forall x \in \mathcal{M}, \quad B(x, \epsilon \cdot f(x)) \cap P \neq \emptyset.$$

$$\forall p \in P, \quad B(p, \delta \cdot f(p)) \cap P = \{p\}.$$

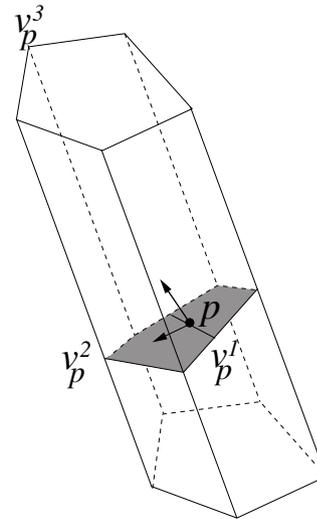
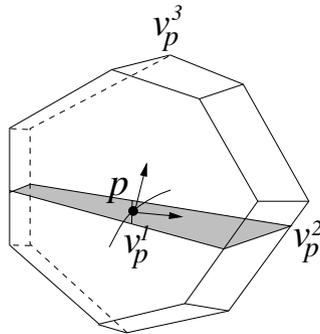


Dimension detection

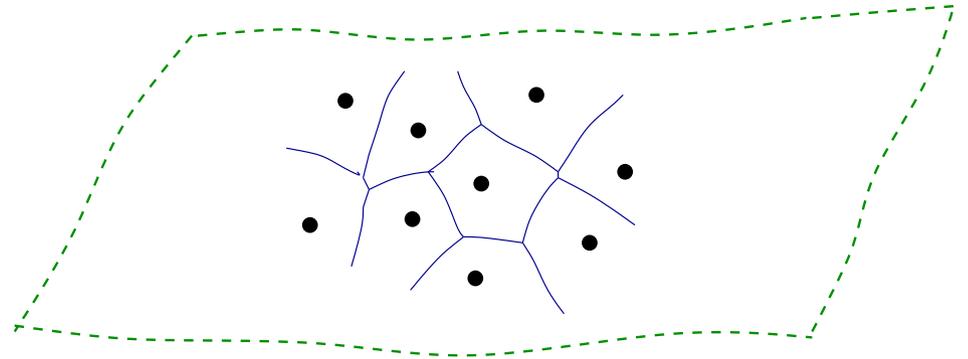
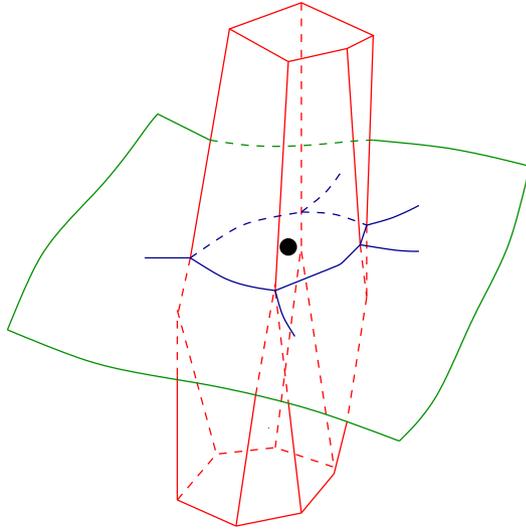
Dey-Giesen-Goswami-Zhao 2002

Define Voronoi subset V_p^i recursively for a point $p \in P$.

- $\text{aff}V_p^k$ approximates T_p .
- k can be determined if P is (ε, δ) -sample for appropriate ε and δ .



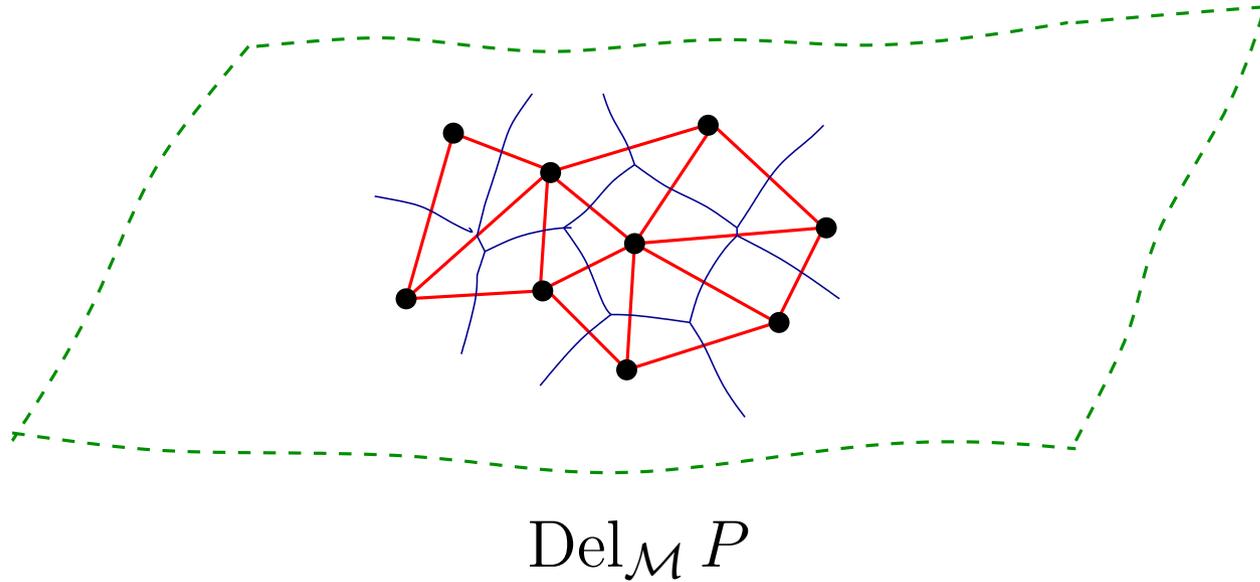
Restricted Voronoi Diagram



RESTRICTED VORONOI FACE: Intersection of Voronoi face with manifold

BALL PROPERTY: Each Voronoi face is topologically a ball.

Restricted Delaunay Triangulation



This is a good candidate to be a “correct reconstruction”.

Topology of RDT

TBP Theorem [Edelsbrunner-Shah 94] : If $V_{\text{or}} P$ has the topological ball property w.r.t. \mathcal{M} , then $\text{Del}_{\mathcal{M}} P$ has homeomorphic underlying space to \mathcal{M} .



Topology of RDT

TBP Theorem [Edelsbrunner-Shah 94] : If $\text{Vor } P$ has the topological ball property w.r.t. \mathcal{M} , then $\text{Del}_{\mathcal{M}} P$ has homeomorphic underlying space to \mathcal{M} .

RDT Theorem [AB98, CDES01] : If P is 0.2-sample for a surface $\mathcal{M} \subset \mathbb{R}^3$, then $\text{Vor } P$ satisfies TBP with respect to \mathcal{M} .



Difficulty I: No RDT Thm.

Negative result [Cheng-Dey-Ramos 05] : For a k -manifold $\mathcal{M} \subset \mathbb{R}^d$, $\text{Del}_{\mathcal{M}} P$ may not be homeomorphic to \mathcal{M} no matter how dense P is when $k > 2$ and $d > 3$.

- Due to this result, Witness complex [Carlsson-de Silva] may not be homeomorphic to \mathcal{M} as noted in [Boissonnat-Oudot-Guibas 07].



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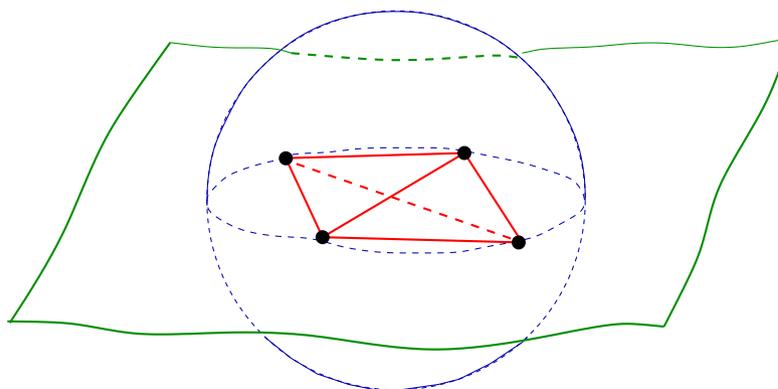


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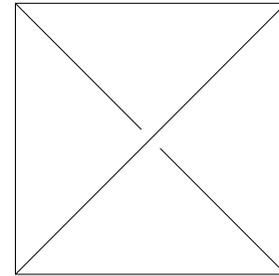
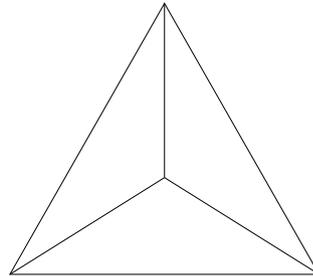
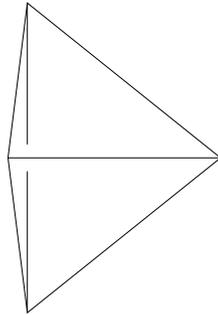
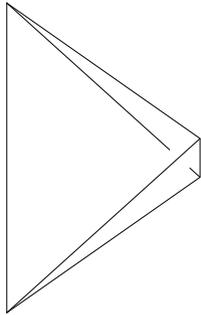
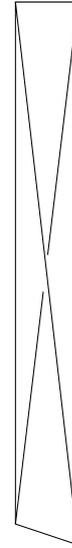
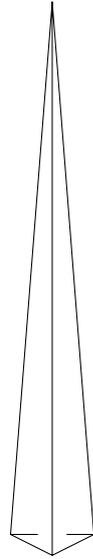
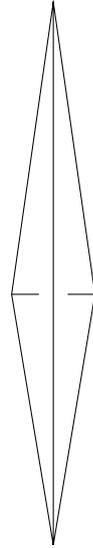


Difficulty I

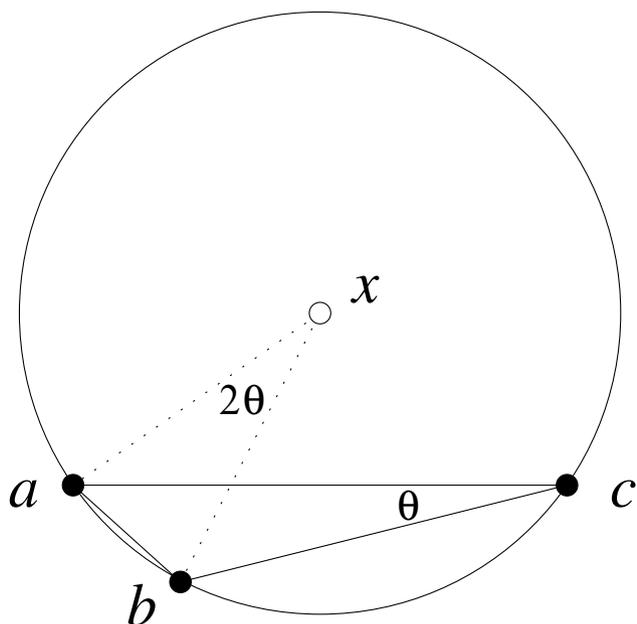
- RDT theorem used the fact that Voronoi faces intersect \mathcal{M} orthogonally.
- This is not true in high dimensions because of slivers whose dual faces may have large deviations from the normal space.



Simplex Shape



Simplex Shape

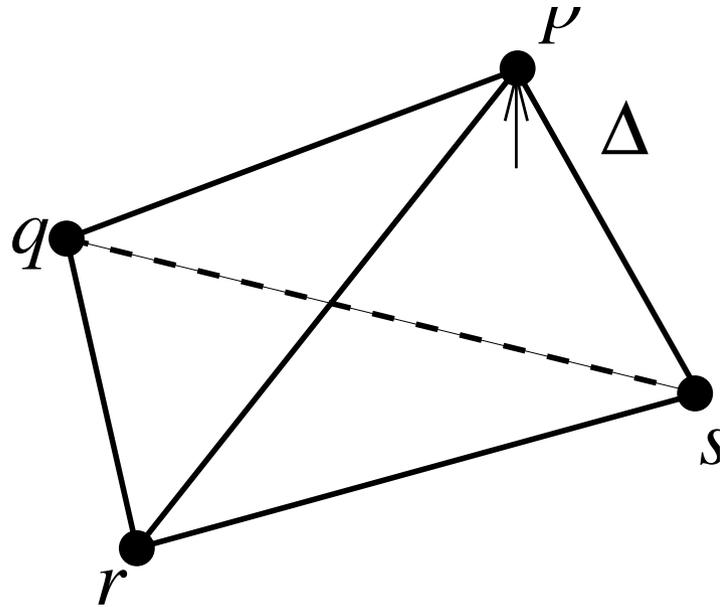


$$\sin \theta = \frac{L_{\tau}}{2R_{\tau}}$$

- R ... circumradius
- L_{τ} ... shortest edge length
- R_{τ} ... circumradius
- R_{τ}/L_{τ} ... circumradius-edge ratio



Sliver



A j -simplex, $j > 1$, τ is a sliver if none of its subsimplices are sliver and

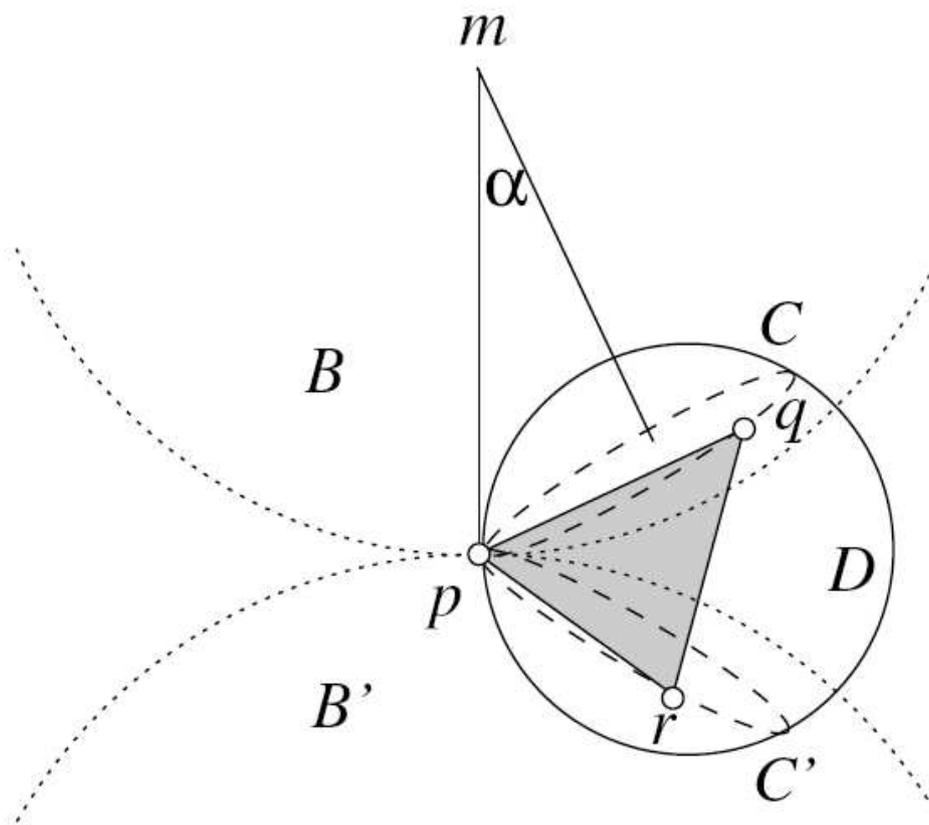
$$\text{vol}(\tau) \leq \sigma^j L_\tau^j$$

where L_τ is the shortest edge length, and σ is a parameter.



Difficulty I: Slivers in 3D

In 3-d, if a Delaunay triangle has a circumradius $O(\varepsilon f(p))$ then its normal and the normal of \mathcal{M} at p form an angle $O(\varepsilon)$.



Difficulty I: Slivers' Normals

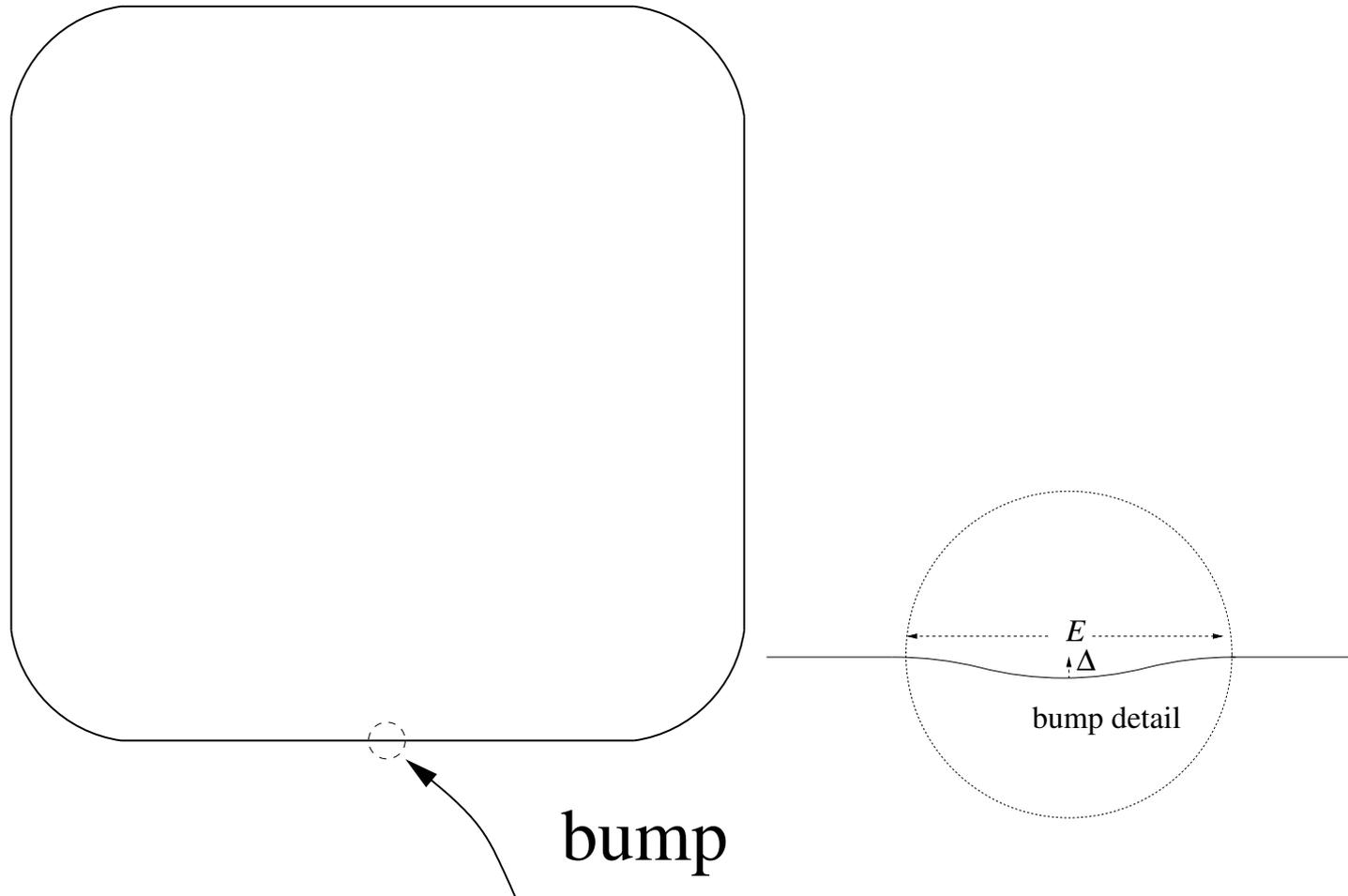
In 4-d, considering a 3-manifold, a Delaunay 3-simplex may have small circumradius, that is $O(\varepsilon f(p))$, but its normal may be very different from that of \mathcal{M} at p . **Slivers are the culprits:**

$$p = (0, 0, \Delta, 0); q = (1, 0, 0, 0), r = (1, 1, 0, 0), s = (0, 1, 0, 0)$$

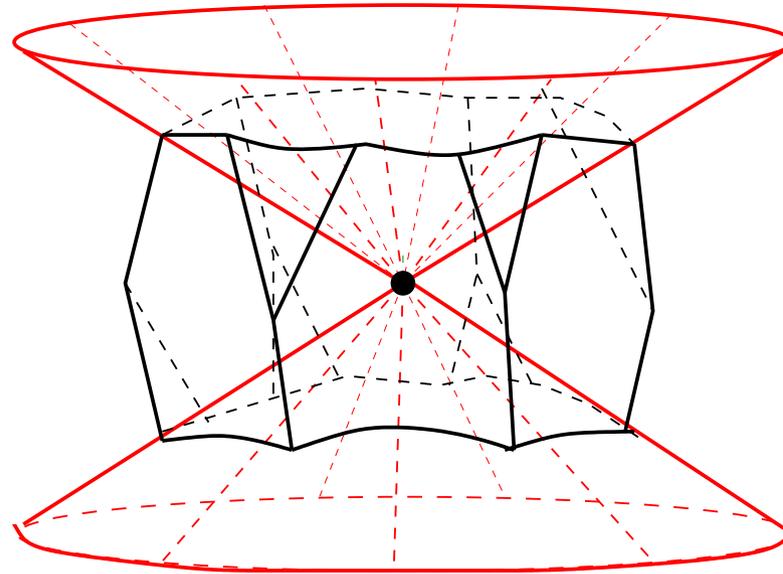
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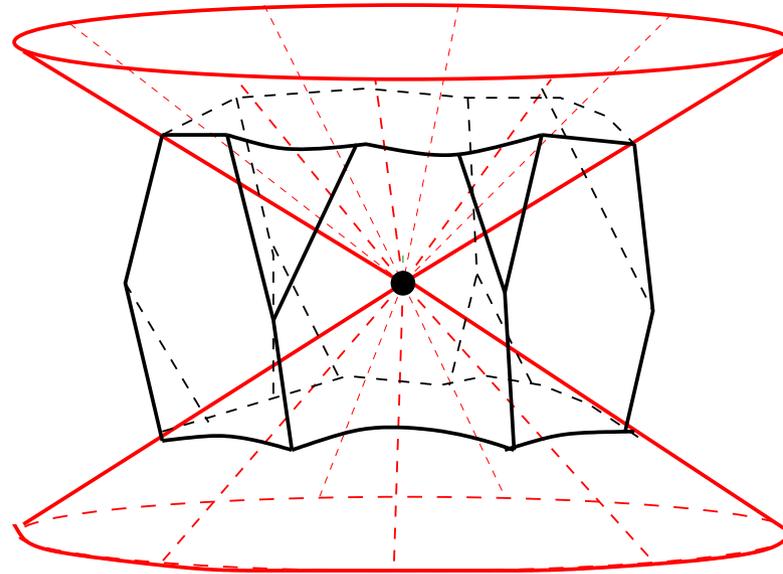
Bad Normal



Difficulty II



Difficulty II



- It is not possible to identify precisely the restricted Delaunay triangulation because of **slivers**: there are Voronoi faces close to the surface but not intersecting it.

Solution[Cheng-Dey-Ramos 05]

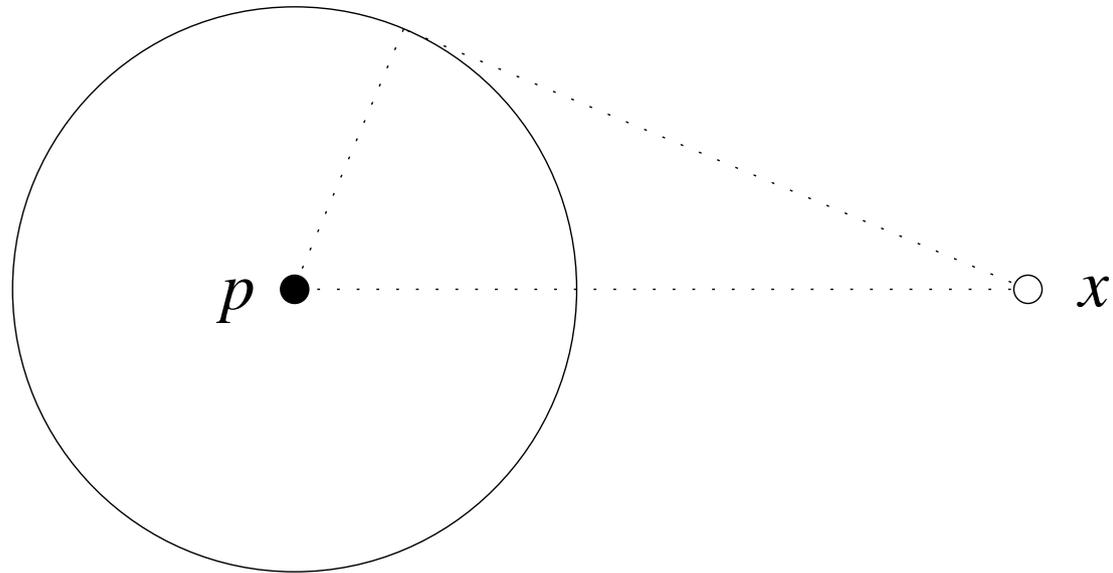
Get rid of the slivers:

Follow [sliver exudation](#) approach of [Cheng-Dey-Edelsbrunner-Facello-Teng 2000](#) in the context of meshing.



Weighted Voronoi/Delaunay

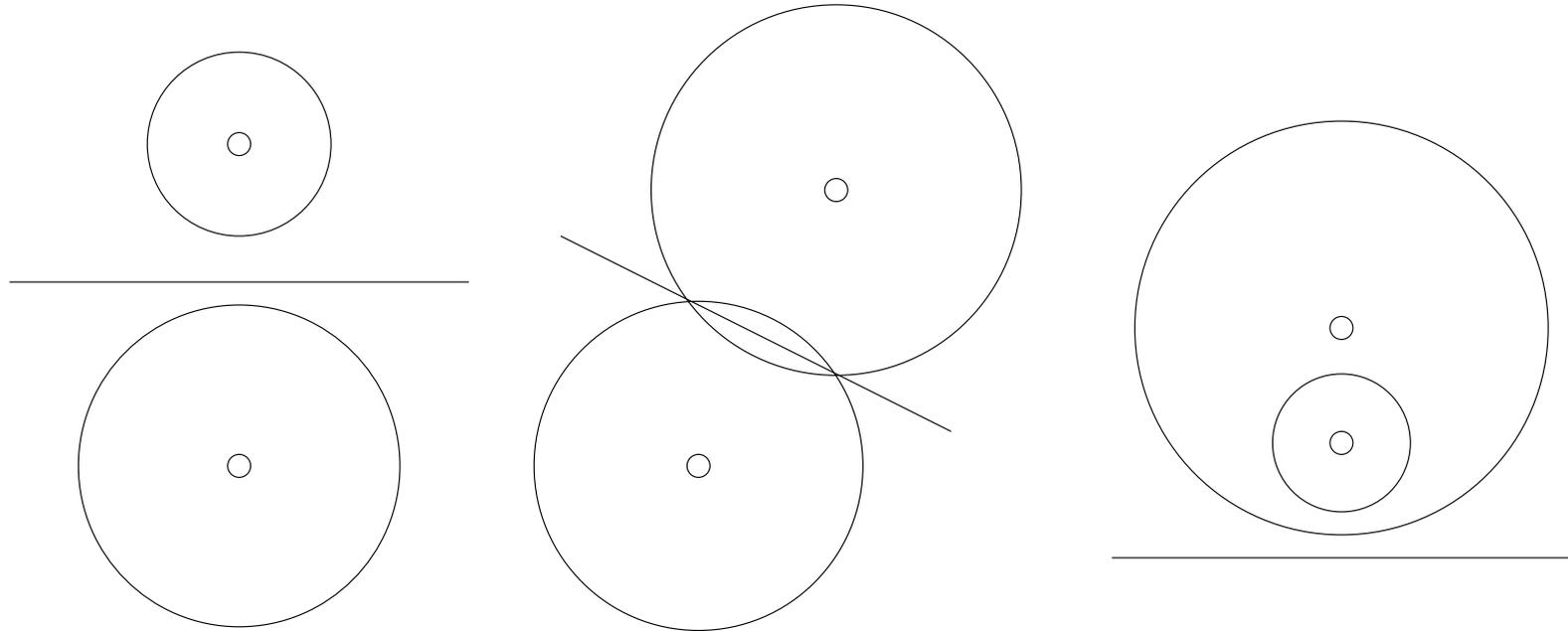
Weighted Points: (p, w_p) .



Weighted Distance: $\pi_{(p, w_p)}(x) = \|x - p\|^2 - w_p^2$.



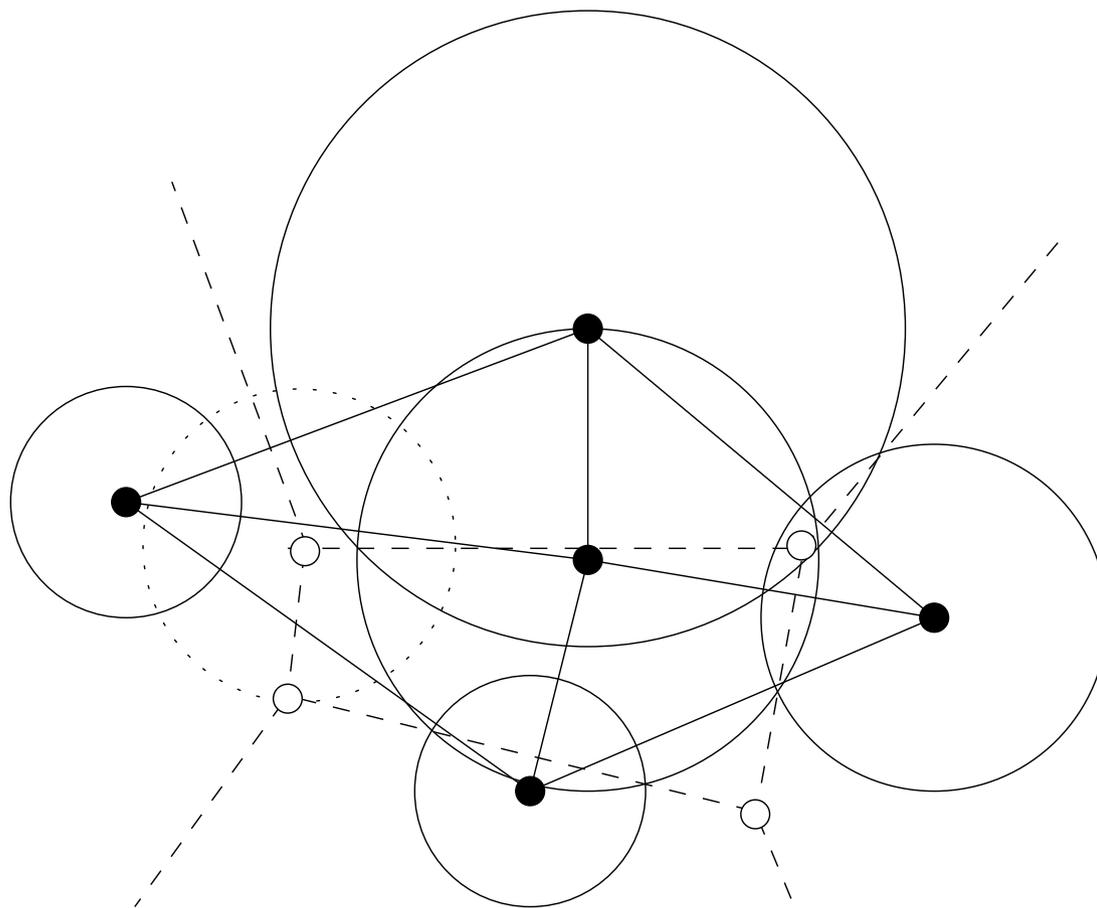
Weighted Voronoi/Delaunay



Bisectors of weighted points

Weight property $[\omega]$: For each $p \in P$, $w_p \leq \omega N(p)$, where $N(p)$ is the distance to closest neighbor. Limit $\omega \leq 1/4$.

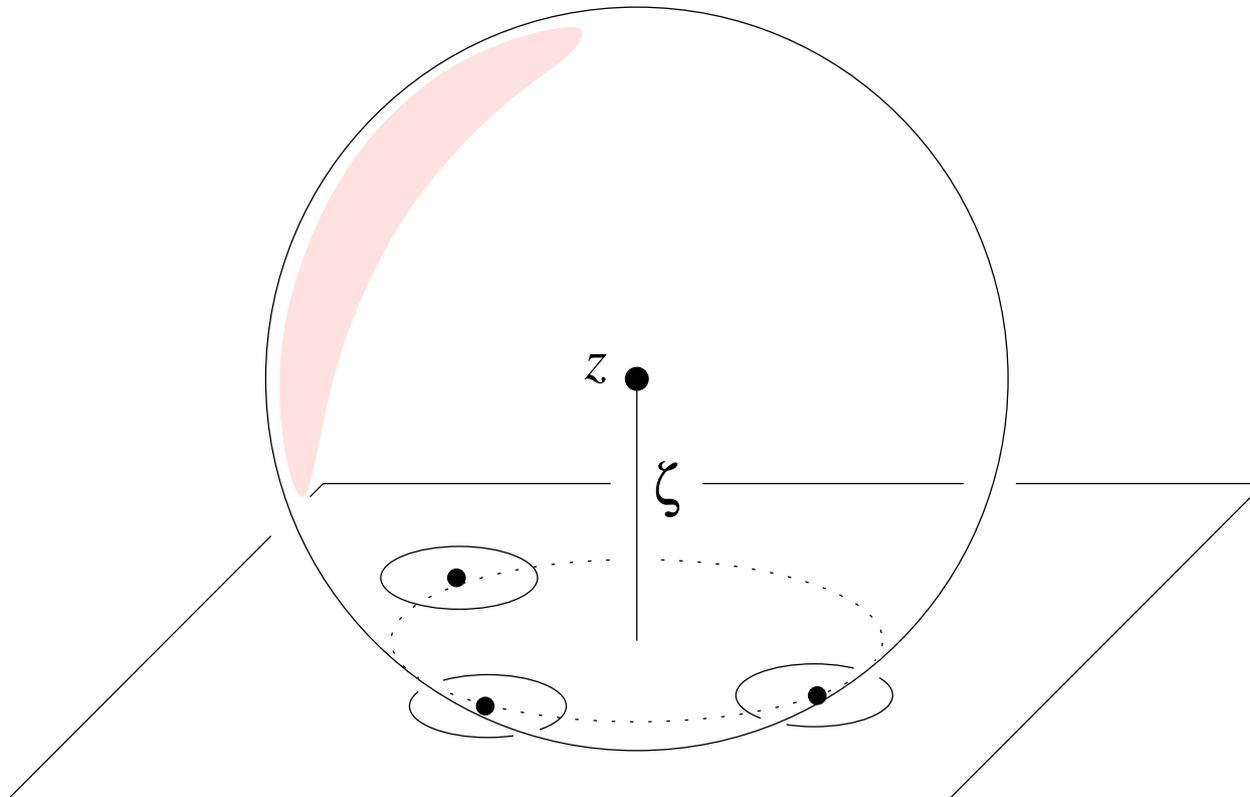
Weighted Voronoi/Delaunay



Weighted Delaunay/Voronoi complexes

Orthocenter Sensitivity

- An appropriate weight on each sample moves undesirable Voronoi faces away from the manifold



Sliver Exudation

- By a packing argument, the number of neighbors of p that can be connected to p in any weighted Delaunay triangulation with $[\omega]$ weight property is

$$\lambda = O(\nu^{2d})$$

where $\nu = \varepsilon/\delta$.

- The number of possible weighted Delaunay ($\leq k$)-simplices in which p is involved is at most

$$N_p = O(\lambda^k) = O(\nu^{2kd})$$



Sliver Exudation

- The total w_p^2 length of “bad” intervals is at most:

$$N_p \cdot c' \sigma \varepsilon^2 f^2(p).$$

- The total w_p^2 length is $\omega^2 \delta^2 f^2(p)$. So if we choose

$$\sigma < \frac{\omega^2}{c' N_p \nu^2}$$

then there is a radius w_p for p such that no cocone simplex is a sliver.

- For ε sufficiently small, if τ is a cocone $(k + 1)$ simplex, it must be a sliver.



Algorithm

- Construct Vor P and Del P .
- Determine the dimension k of \mathcal{M} .
- “Pump up” the sample point weights to remove all j -slivers, $j = 3, \dots, k + 1$, from all point cocones.
- Extract all cocone simplices as the resulting output.



Correctness

The algorithm outputs $\text{Del}_{\mathcal{M}}(\hat{P})$:

- (i) for σ sufficiently small, there is a weight assignment to the sample points so that no cocone j -simplex, $j \leq k + 1$, is a sliver;
- (ii) for ε sufficiently small, any cocone $(k + 1)$ -simplex must be a sliver.



Correctness

$\text{Del}_{\mathcal{M}}(\hat{P})$ is “close” to \mathcal{M} and homeomorphic to it:

- (i) the normal spaces of close points in \mathcal{M} are close: if p, q are at distance $O(\varepsilon f(p))$, then their normal spaces form an angle $O(\varepsilon)$;
- (ii) for any j -simplex with $j \leq k$, if its circumradius is $O(\varepsilon f(p))$ and neither τ nor any of the boundary simplices is a sliver, then the normal space of τ is close to the normal space of \mathcal{M} at p ;
- (iii) each cell of $\text{Vor}_{\mathcal{M}} \hat{P}$ is a topological ball
- (iii) implies that $\text{Del}_{\mathcal{M}} \hat{P}$ is homeomorphic to \mathcal{M} (Edelsbrunner and Shah)



Summary

- For sufficiently dense samples, the algorithm outputs a mesh that is *faithful* to the original manifold.
- The running time – under (ε, δ) -sampling – is $O(n \log n)$ (constant is exponential with the dimension).
- The ε for which the algorithm works is quite small (more than exponentially small in the dimension).



Concluding remarks

- Various connections to Voronoi diagrams and geometry/topology estimations
- Further works on manifold/compact reconstructions [Niyogi-Smale-Weinberger 06, Chazal-Lieutier 06, Chazal-Cohen Steiner-Lieutire 06, Boissonnat-Oudot-Guibas 07]
- Practical algorithm under realistic assumptions ...???



Thank You

