

# INF562 – Géométrie Algorithmique et Applications

## Curve and surface reconstruction

Steve Oudot



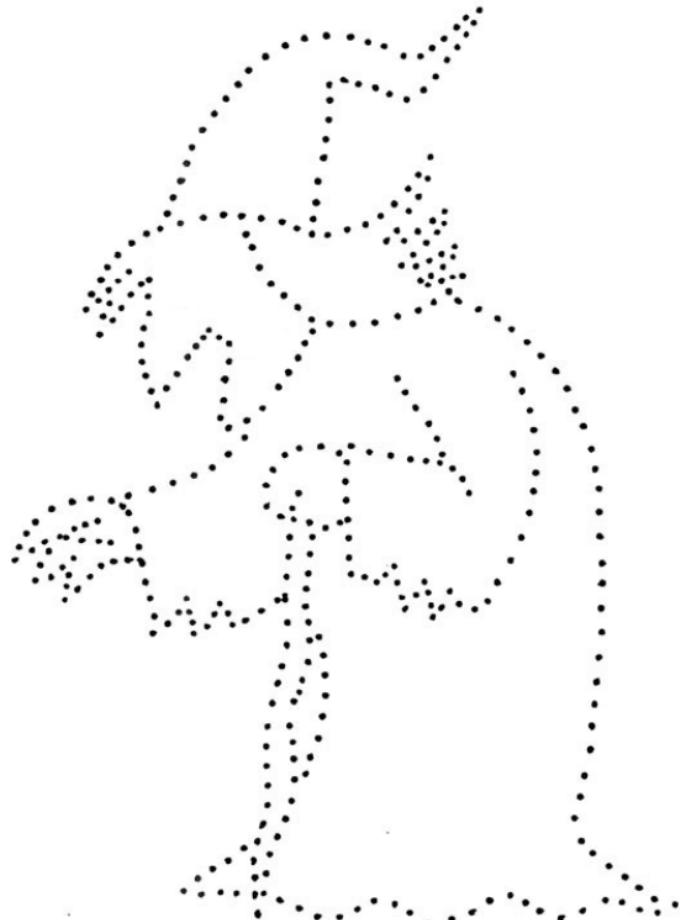
INSTITUT NATIONAL  
DE RECHERCHE  
EN INFORMATIQUE  
ET EN AUTOMATIQUE  
centre de recherche

 **INRIA**  
**SACLAY - ÎLE-DE-FRANCE**

# Reconstruction Paradigm

**Q** What do you see?

Why?



# Reconstruction Paradigm

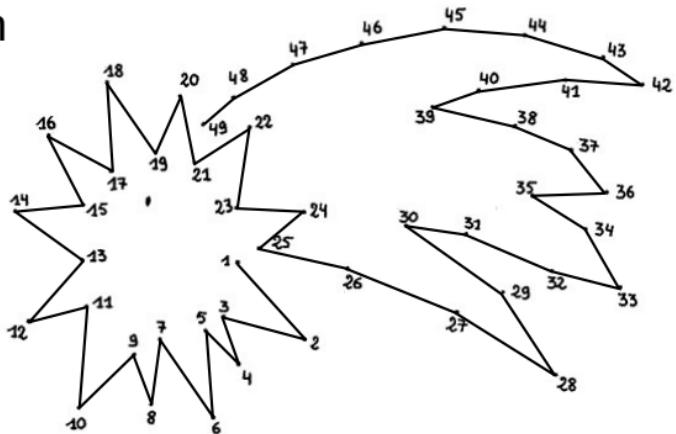
**Q** What do you see?

Why?

# Reconstruction Paradigm

Q What do you see?

Why?

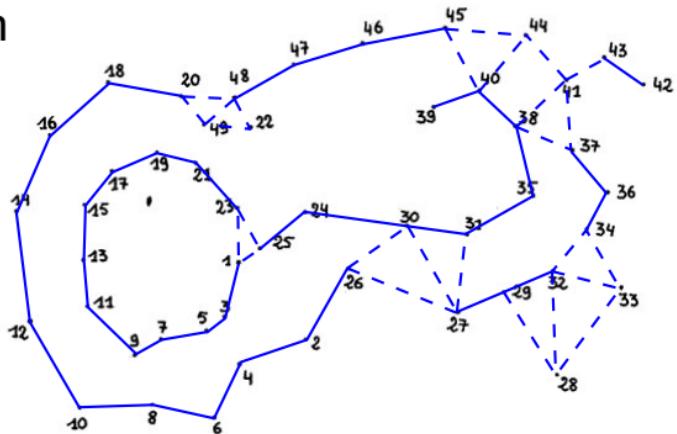


# Reconstruction Paradigm

**Q** What do you see?

## Why?

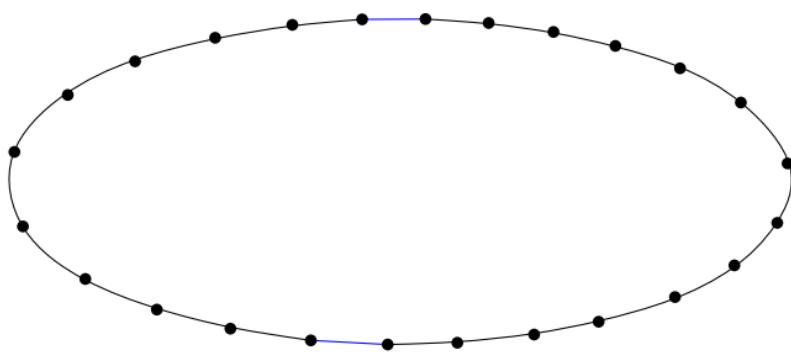
without the numbers...



## Reconstruction Paradigm (Cont'd)

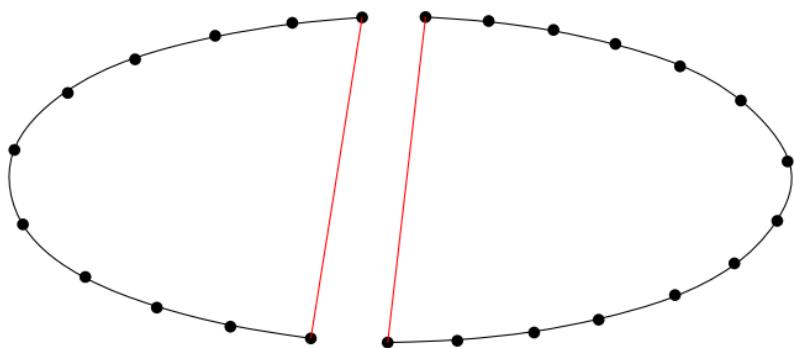


## Reconstruction Paradigm (Cont'd)



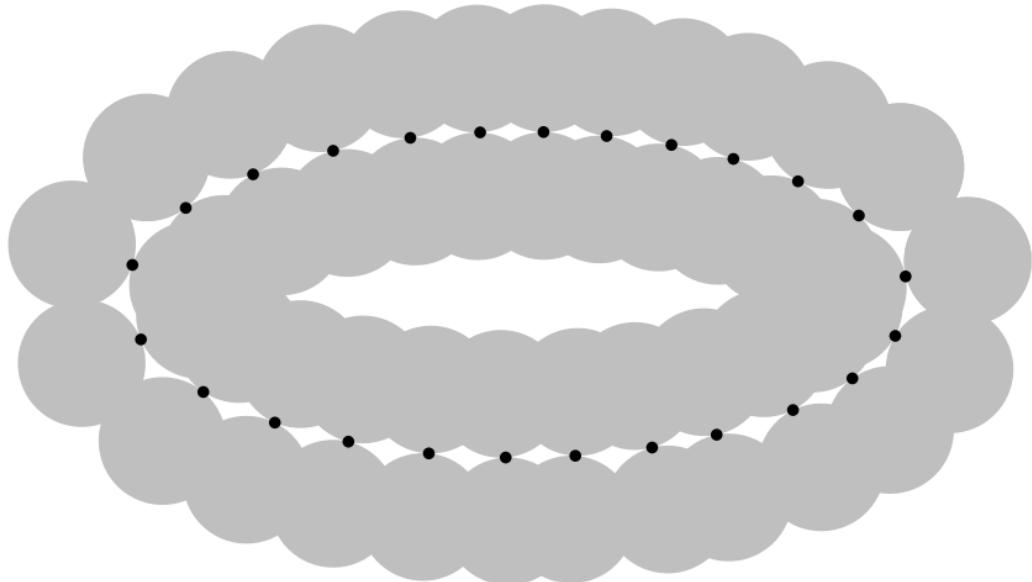
**Q** Given a point cloud, build a *faithful* (implicit, PL, ...) approximation of the shape underlying the data.

## Reconstruction Paradigm (Cont'd)



Reconstruction problem is ill-posed by nature.

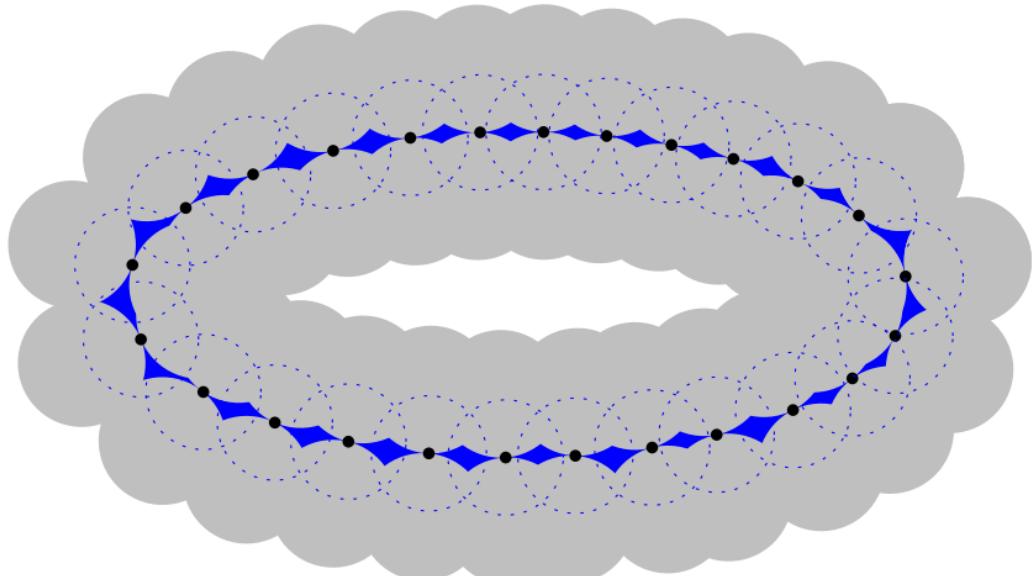
## Reconstruction Paradigm (Cont'd)



Reconstruction problem is ill-posed by nature.

→ make assumptions on the underlying shape, *e.g.*: fix dimension, topological type, regularity (differentiability), Hausdorff distance to input...

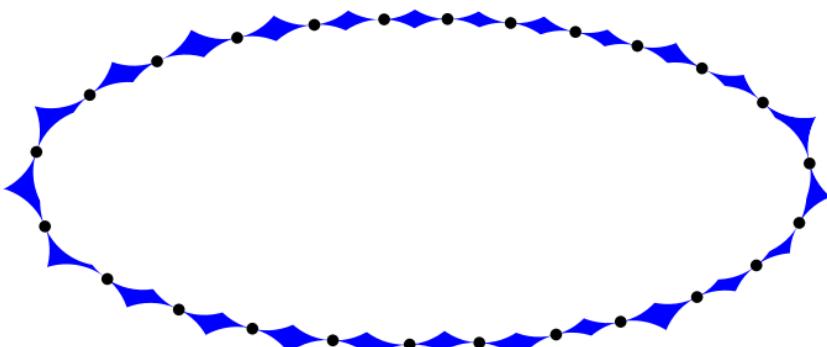
## Reconstruction Paradigm (Cont'd)



Reconstruction problem is ill-posed by nature.

→ make assumptions on the underlying shape, *e.g.*: fix dimension, topological type, regularity (differentiability), Hausdorff distance to input...

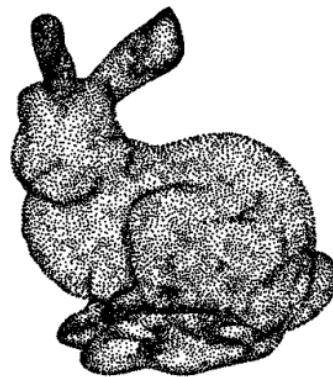
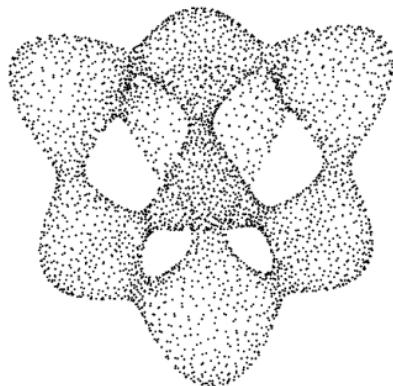
## Reconstruction Paradigm (Cont'd)



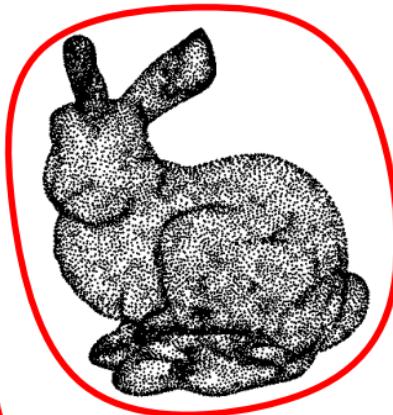
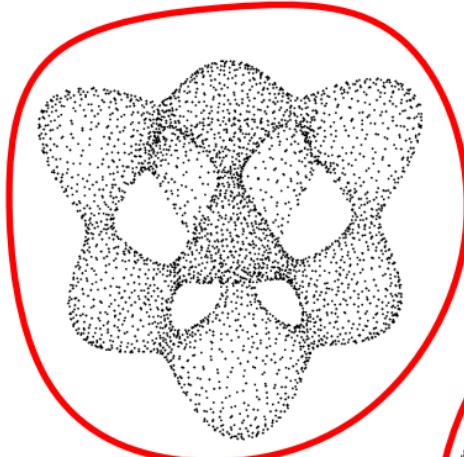
Reconstruction problem is ill-posed by nature.

- make assumptions on the underlying shape, *e.g.*: fix dimension, topological type, regularity (differentiability), Hausdorff distance to input...
- for a suitable choice of hypotheses, the solution becomes unique **up to a set of local regular deformations** (solution never unique!)

## Other (weaker) forms of reconstruction

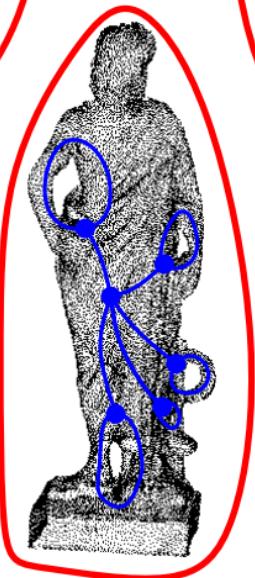
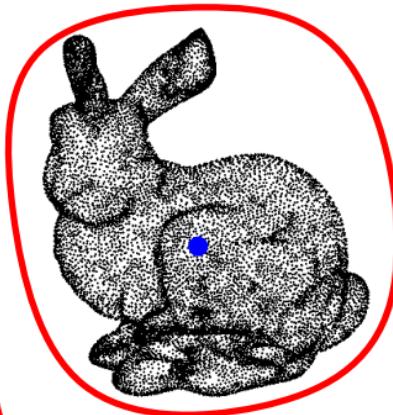
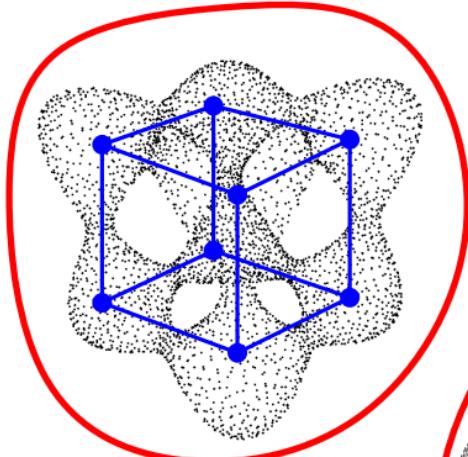


## Other (weaker) forms of reconstruction



clustering

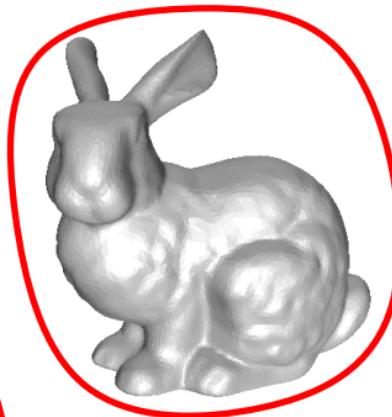
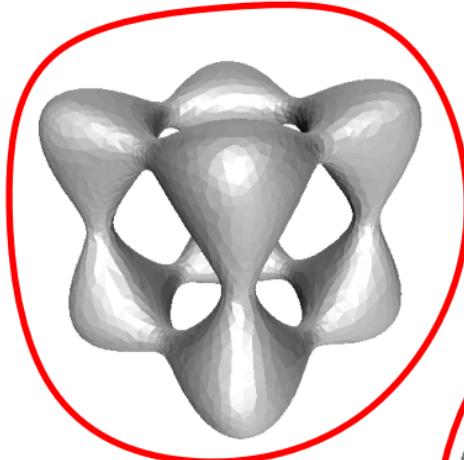
## Other (weaker) forms of reconstruction



clustering

topological inference

## Other (weaker) forms of reconstruction



clustering

topological inference

reconstruction

Where do the data come from?

## 3D scans

### Sources

LASER

stereo vision

mechanical sensor

### Applications

Reverse engineering

Prototyping

Quality control



Stanford Michelangelo Project

Where does the data come from?

## Medical Imaging

### Sources

- MRI scan
- echograph

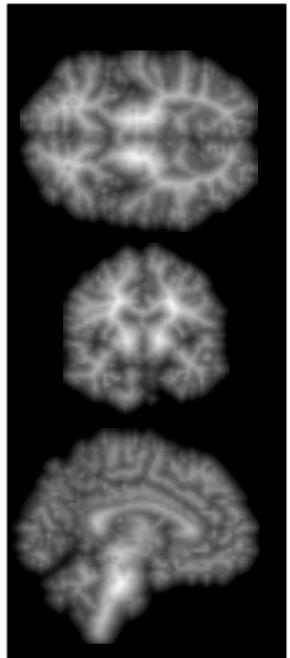
...

### Applications

- Diagnostic

- Endoscopy simulation

- Chirurgical intervention planning



Where does the data come from?

## Geography, Geology

### Sources

satellite/aerial images

ground probing

seismograph

### Applications

Maps making / Terrain modeling

Prospection (tunnels, oil)



Where does the data come from?

## Higher-Dimensions

### Sources

Data bases  
Simulations

### Applications

Machine Learning

Path planning

Pattern recognition

Image processing

...



$S^7$



# Various reconstruction techniques

## Delaunay-based

- Crust / Power Crust
- Cocone
- Gabriel /  $\alpha$ -shape /  $\beta$ -skeleton
- flow complex

## Implicitization

- Local polynomial fitting
- Natural Neighbors (Voronoi-based)
- Radial Basis Functions

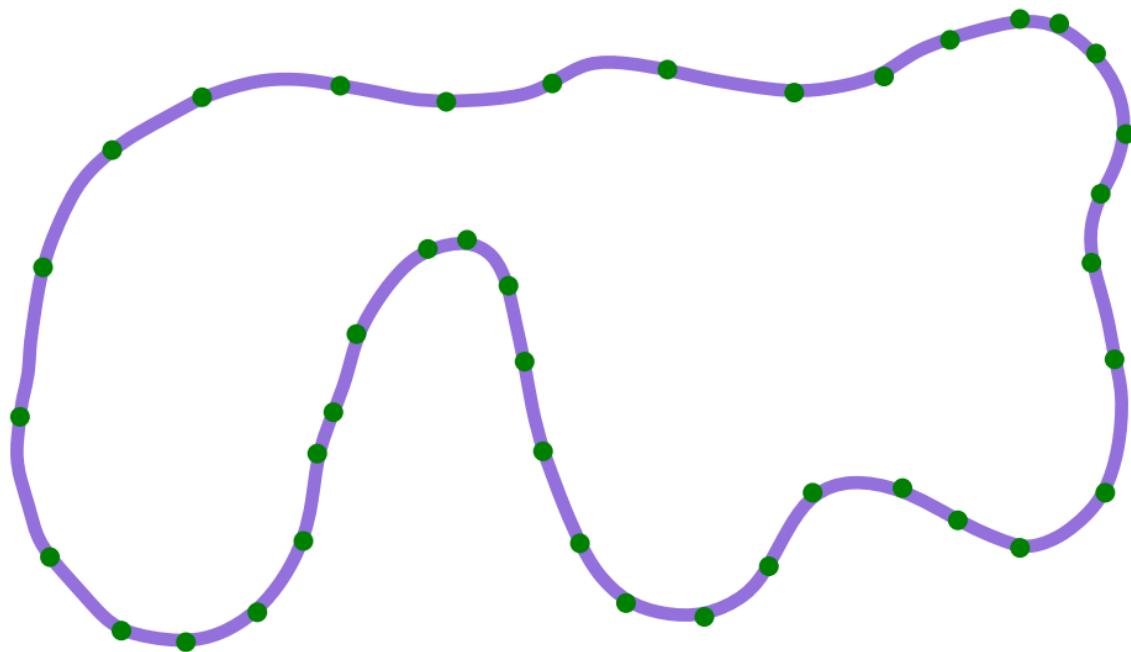
## Projection operators

- Moving Least Squares
- Extremal surfaces

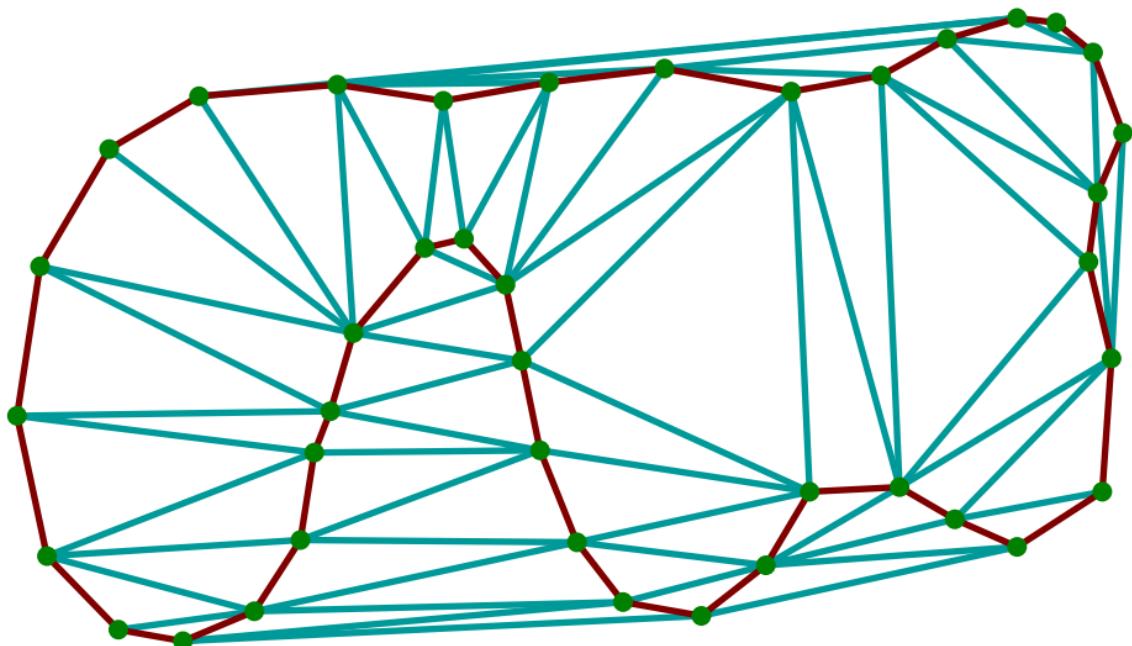
## For arbitrary dimensions and co-dimensions

- Unions of balls / nerves
- Witness Complex

# What Delaunay has to do with reconstruction

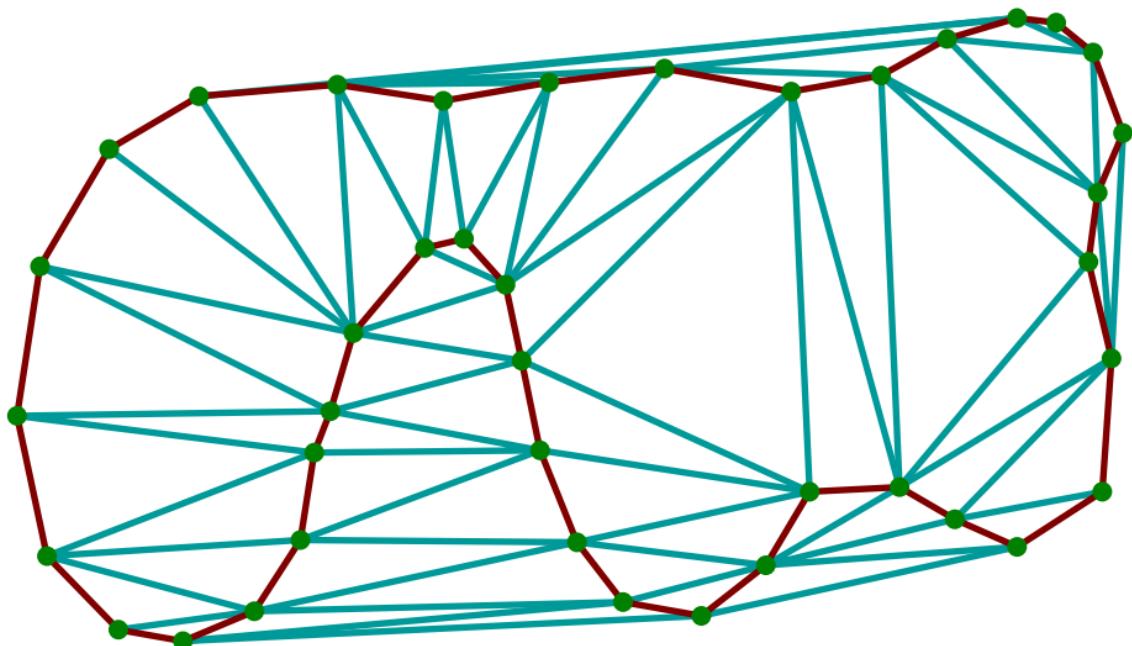


# What Delaunay has to do with reconstruction



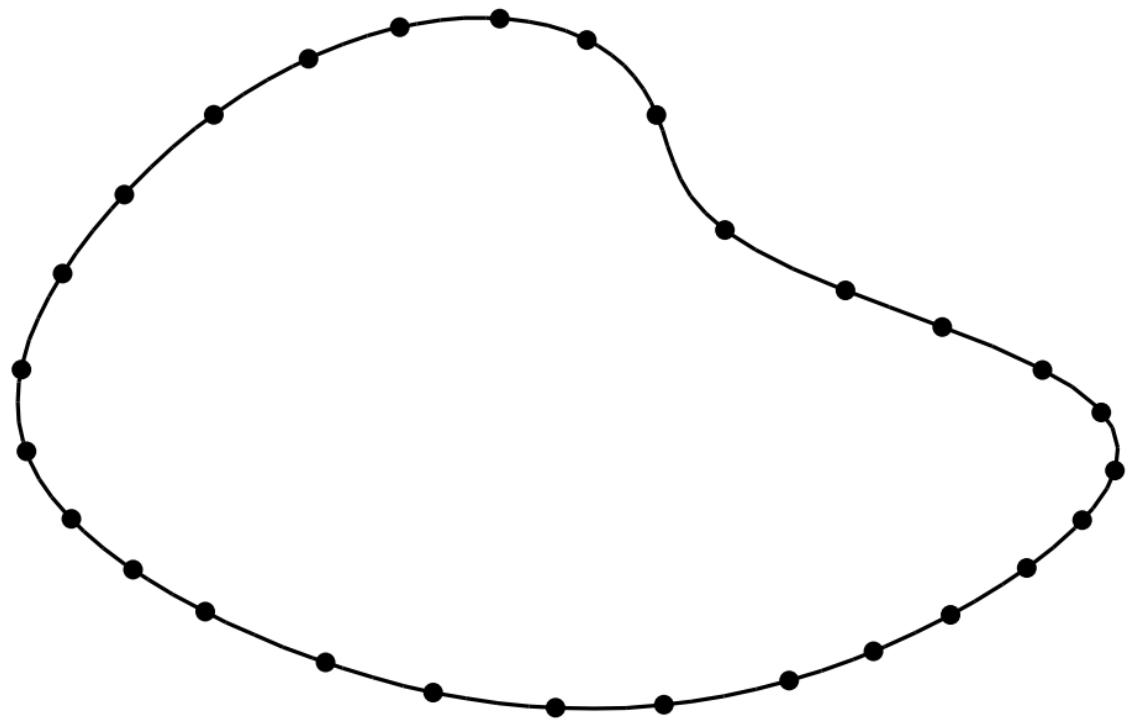
- a faithful approximation of the curve appears as a subcomplex of the Delaunay
- this should hold whenever the point cloud is sufficiently densely sampled along the curve

# What Delaunay has to do with reconstruction

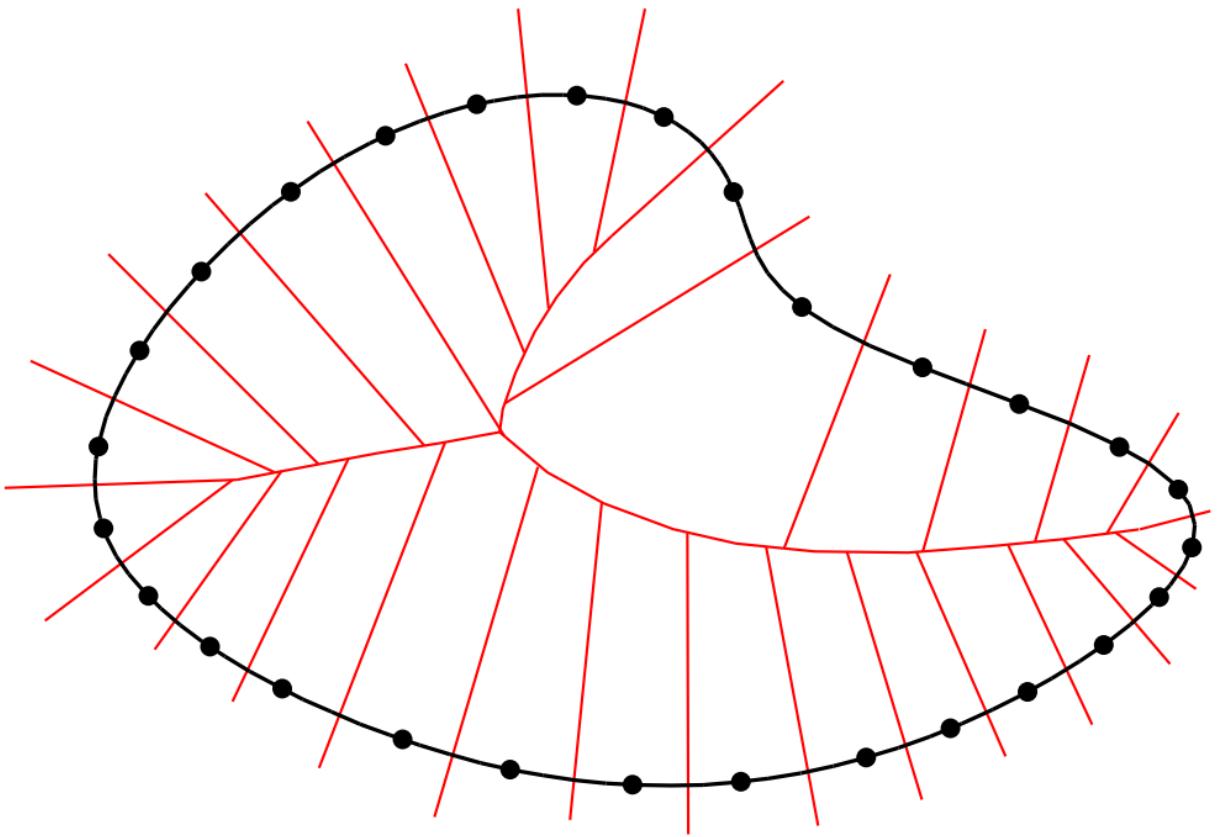


- a faithful approximation of the curve appears as a subcomplex of the Delaunay
  - this should hold whenever the point cloud is sufficiently densely sampled along the curve
- Q** What is this *good* subcomplex? Can it be defined in some canonical way?

# Restricted Delaunay triangulation

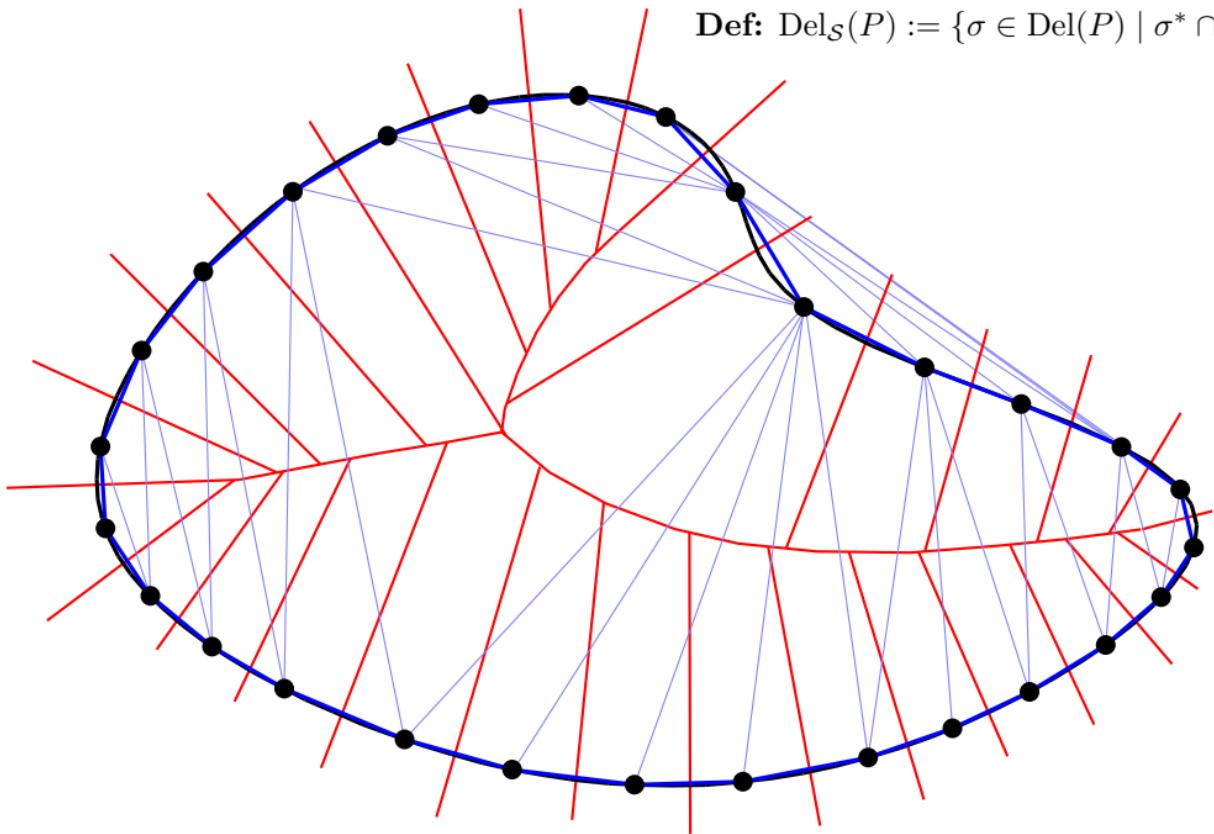


# Restricted Delaunay triangulation



# Restricted Delaunay triangulation

**Def:**  $\text{Del}_{\mathcal{S}}(P) := \{\sigma \in \text{Del}(P) \mid \sigma^* \cap \mathcal{S} \neq \emptyset\}$



# Approximation power of the restricted Delaunay

→ Our assumptions:

1. the underlying shape  $\mathcal{S}$  is a closed curve or surface with positive *reach*  $\varrho_{\mathcal{S}}$
2. the point cloud  $P$  is an  $\varepsilon$ -sample of  $\mathcal{S}$  with  $\varepsilon \in O(\varrho_{\mathcal{S}})$ .

# Approximation power of the restricted Delaunay

→ Our assumptions:

→ analogy with 1-d signal theory (Shannon's reconstruction theorem):

1. the underlying shape  $\mathcal{S}$  is a closed curve or surface with positive *reach*  $\varrho_{\mathcal{S}}$
- 1'. the underlying signal is a weighted sum of sinusoids
2. the point cloud  $P$  is an  $\varepsilon$ -sample of  $\mathcal{S}$  with  $\varepsilon \in O(\varrho_{\mathcal{S}})$ .
- 2'. the sampling has  $\geq 2$  samples per period (signal has bounded bandwidth)

# Approximation power of the restricted Delaunay

**Theorem:** [Amenta et al. 1998-99]

If  $\mathcal{S}$  is a curve or surface with positive *reach*, and if  $P$  is an  $\varepsilon$ -sample of  $\mathcal{S}$  with  $\varepsilon < \varrho_{\mathcal{S}}$  (curve) or  $\varepsilon < 0.1\varrho_{\mathcal{S}}$  (surface), then:

- $\text{Del}_{\mathcal{S}}(P)$  is homeomorphic to  $\mathcal{S}$ ,
- $d_H(\text{Del}_{\mathcal{S}}(P), \mathcal{S}) \in O(\varepsilon^2)$ ,
- $\forall f \in \text{Del}_{\mathcal{S}}(P), \forall v \in f, \angle n_f n_v \mathcal{S} \in O(\varepsilon)$ ,
- $\dots$  (similar areas, curvature estimation, etc.)

# Approximation power of the restricted Delaunay

**Theorem:** [Amenta et al. 1998-99]

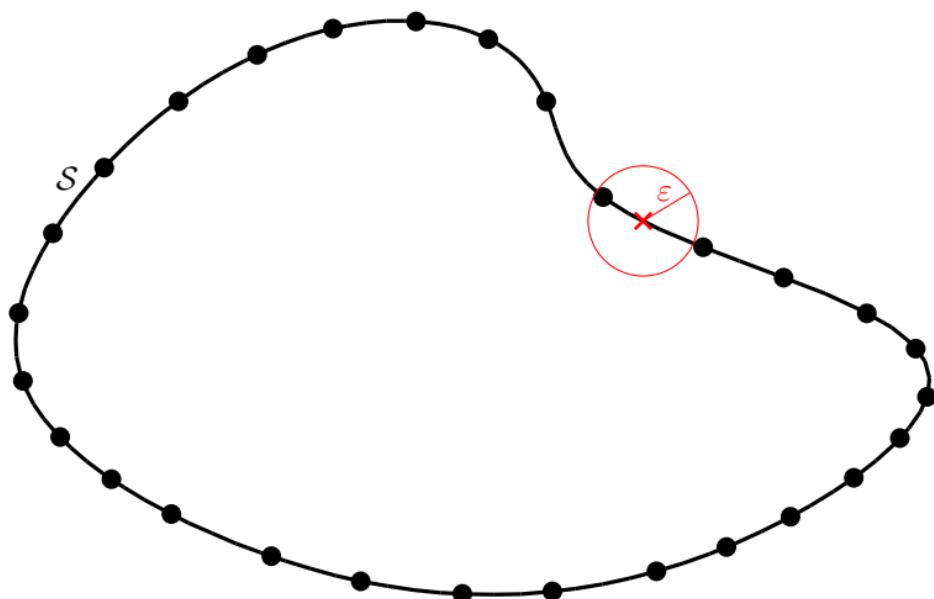
If  $\mathcal{S}$  is a curve or surface with positive *reach*, and if  $P$  is an  $\varepsilon$ -sample of  $\mathcal{S}$  with  $\varepsilon < \varrho_{\mathcal{S}}$  (curve) or  $\varepsilon < 0.1\varrho_{\mathcal{S}}$  (surface), then:

- $\text{Del}_{\mathcal{S}}(P)$  is homeomorphic to  $\mathcal{S}$ ,
- $d_H(\text{Del}_{\mathcal{S}}(P), \mathcal{S}) \in O(\varepsilon^2)$ ,
- $\forall f \in \text{Del}_{\mathcal{S}}(P), \forall v \in f, \angle n_f n_v \mathcal{S} \in O(\varepsilon)$ ,
- $\dots$  (similar areas, curvature estimation, etc.)

→ to be explicated:  $\varepsilon$ -sampling, reach

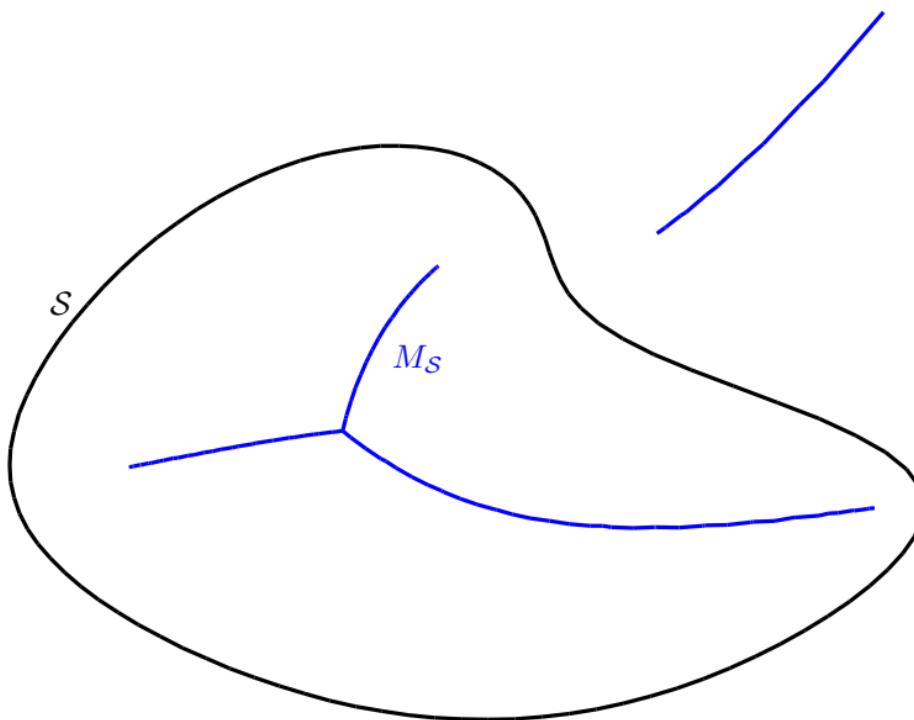
## $\varepsilon$ -samples

**Def:**  $P$  is an  $\varepsilon$ -sample of  $\mathcal{S}$  if  $\forall x \in \mathcal{S}, \min\{\|x - p\| \mid p \in P\} \leq \varepsilon$ .



# Shapes with positive reach [Federer 1958]

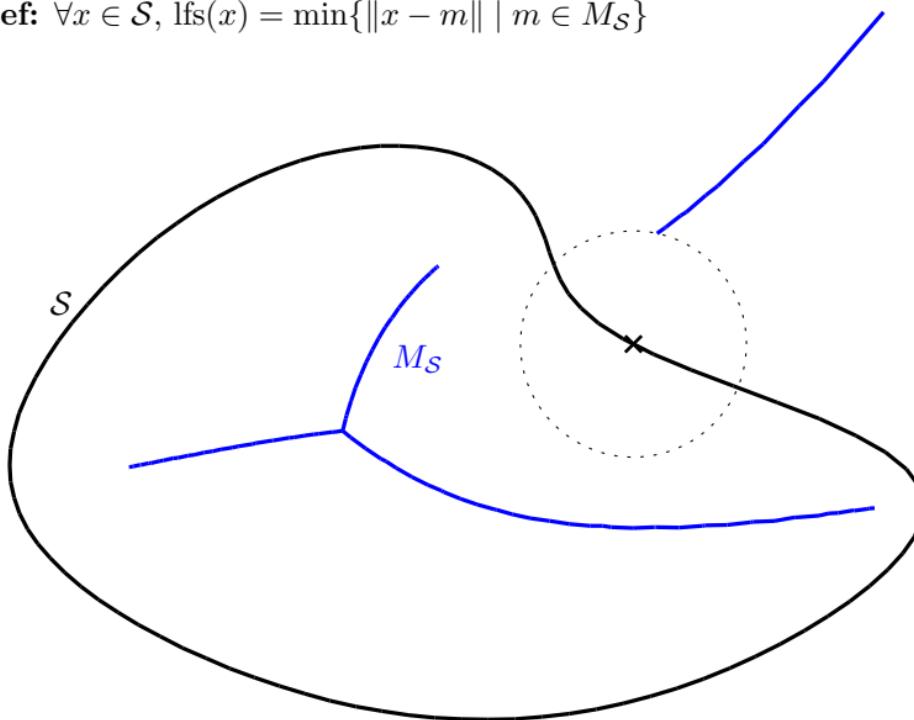
**Def:**  $M_S$  is the closure of the set of points of  $\mathbb{R}^d$  that have  $\geq 2$  nearest neighbors on  $S$ .



# Shapes with positive reach [Federer 1958]

**Def:**  $M_S$  is the closure of the set of points of  $\mathbb{R}^d$  that have  $\geq 2$  nearest neighbors on  $S$ .

**Def:**  $\forall x \in S, \text{lfs}(x) = \min\{\|x - m\| \mid m \in M_S\}$

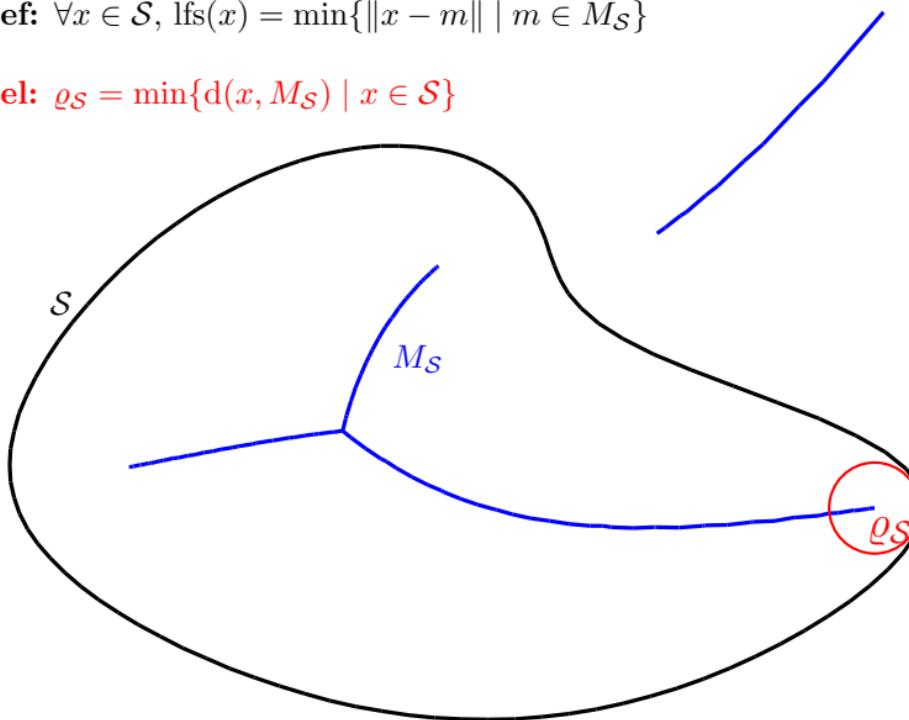


# Shapes with positive reach [Federer 1958]

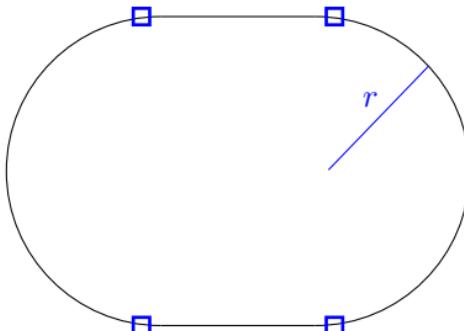
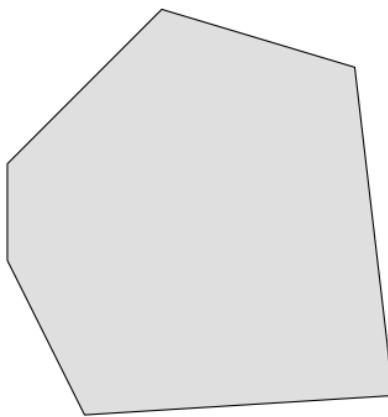
**Def:**  $M_S$  is the closure of the set of points of  $\mathbb{R}^d$  that have  $\geq 2$  nearest neighbors on  $S$ .

**Def:**  $\text{lfs}(x) = \min\{\|x - m\| \mid m \in M_S\}$

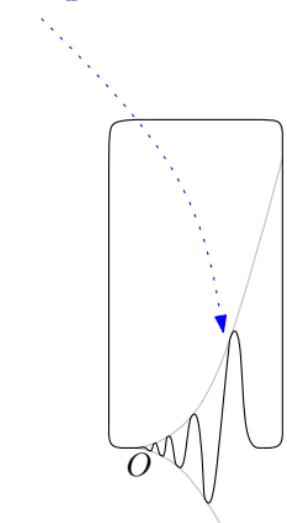
**Def:**  $\varrho_S = \min\{d(x, M_S) \mid x \in S\}$



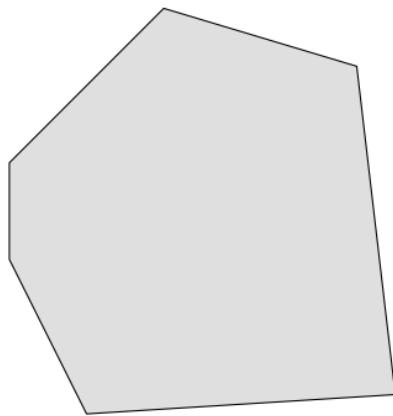
# Shapes with positive reach [Federer 1958]



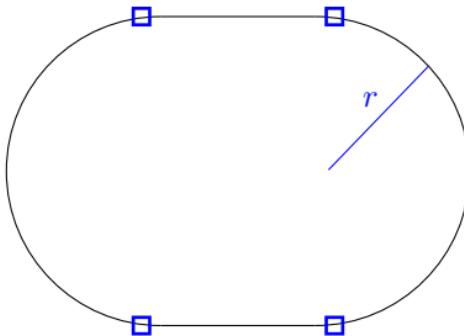
$$x \mapsto x^3 \sin \frac{1}{x}$$



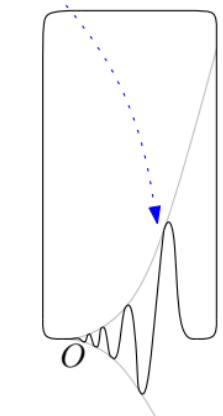
# Shapes with positive reach [Federer 1958]



$\varrho_S = +\infty$   
(convex)



$\varrho_S = r$   
 $C^{1,1}$  but not  $C^2$ )



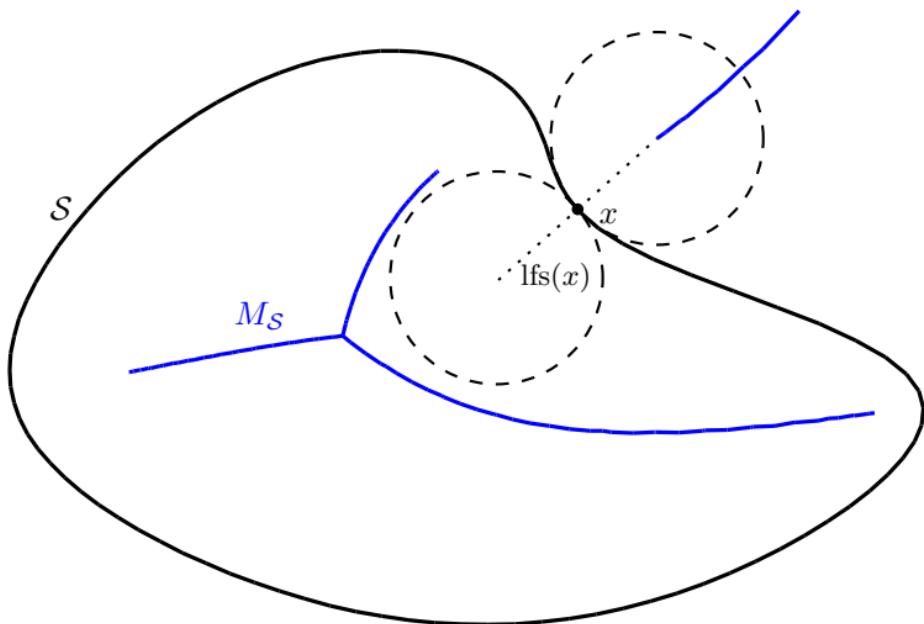
$\varrho_S = 0$   
 $(C^1$  but not  $C^{1,1})$

$$x \mapsto x^3 \sin \frac{1}{x}$$

## Shapes with positive reach (Cont'd)

→ Fundamental properties: (see [Federer 1958])

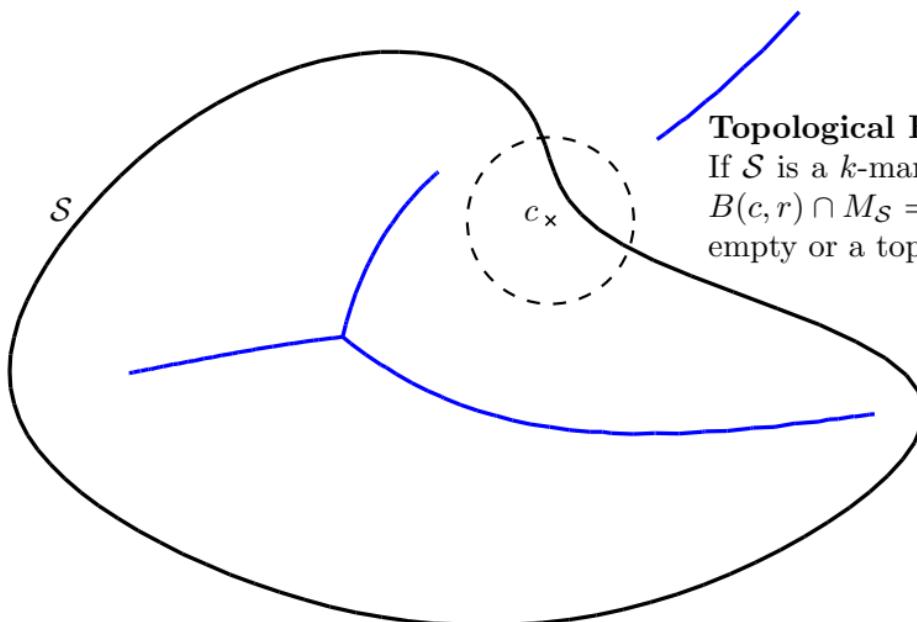
**Tangent Ball Lemma:**  $\forall x \in \mathcal{S}, \forall c \in n_x \mathcal{S}, \|x - c\| < \text{lfs}(x) \Rightarrow B(c, \|x - c\|) \cap \mathcal{S} = \emptyset.$



## Shapes with positive reach (Cont'd)

→ Fundamental properties: (see [Federer 1958])

**Tangent Ball Lemma:**  $\forall x \in \mathcal{S}, \forall c \in n_x \mathcal{S}, \|x - c\| < \text{lfs}(x) \Rightarrow B(c, \|x - c\|) \cap \mathcal{S} = \emptyset.$



**Topological Ball Lemma:**

If  $\mathcal{S}$  is a  $k$ -manifold, then  $\forall B(c, r)$  s.t.  $B(c, r) \cap M_{\mathcal{S}} = \emptyset$ ,  $B(c, r) \cap \mathcal{S}$  is either empty or a topological  $k$ -ball.

# Approximation power of the restricted Delaunay

**Theorem:** [Amenta et al. 1998-99]

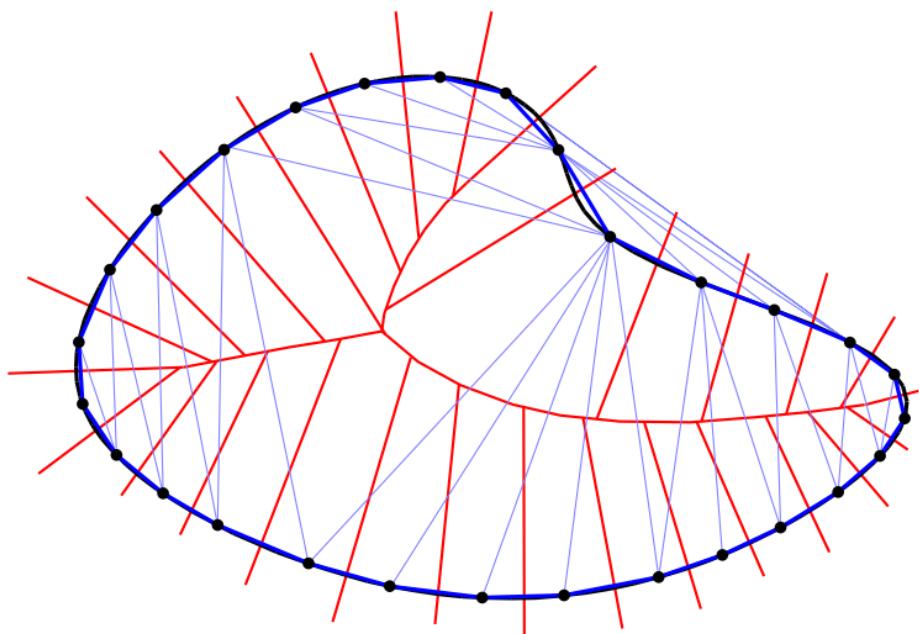
If  $\mathcal{S}$  is a curve or surface with positive *reach*, and if  $P$  is an  $\varepsilon$ -sample of  $\mathcal{S}$  with  $\varepsilon < \varrho_{\mathcal{S}}$  (curve) or  $\varepsilon < 0.1\varrho_{\mathcal{S}}$  (surface), then:

- $\text{Del}_{\mathcal{S}}(P)$  is homeomorphic to  $\mathcal{S}$ ,
- $d_H(\text{Del}_{\mathcal{S}}(P), \mathcal{S}) \in O(\varepsilon^2)$ ,
- $\forall f \in \text{Del}_{\mathcal{S}}(P), \forall v \in f, \angle n_f n_v \mathcal{S} \in O(\varepsilon)$ ,
- $\dots$  (similar areas, curvature estimation, etc.)

# Approximation power of the restricted Delaunay

Proof for curves:

→ show that every edge of  $\text{Del}_{\mathcal{S}}(P)$  connects consecutive points of  $P$  along  $\mathcal{S}$ , and vice-versa



# Approximation power of the restricted Delaunay

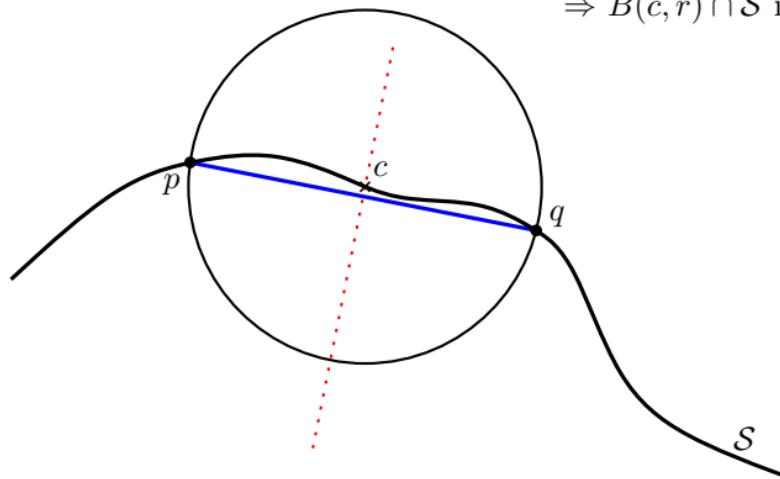
Proof for curves:

→ show that every edge of  $\text{Del}_{\mathcal{S}}(P)$  connects consecutive points of  $P$  along  $\mathcal{S}$ , and vice-versa

Let  $c \in pq^* \cap \mathcal{S}$ .

$$r = \|c - p\| = \|c - q\| = d(c, P) \leq \varepsilon < \varrho_{\mathcal{S}} \leq \text{lfs}(c)$$

$\Rightarrow B(c, r) \cap \mathcal{S}$  is a topological arc



# Approximation power of the restricted Delaunay

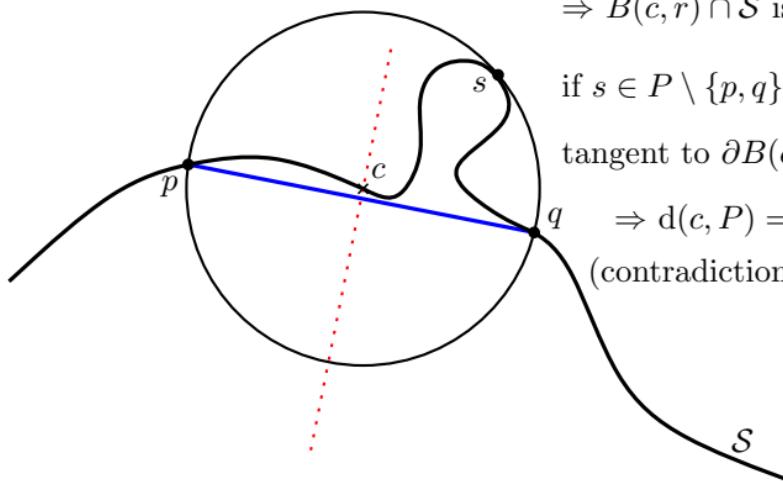
Proof for curves:

→ show that every edge of  $\text{Del}_{\mathcal{S}}(P)$  connects consecutive points of  $P$  along  $\mathcal{S}$ , and vice-versa

Let  $c \in pq^* \cap \mathcal{S}$ .

$$r = \|c - p\| = \|c - q\| = d(c, P) \leq \varepsilon < \varrho_{\mathcal{S}} \leq \text{lfs}(c)$$

$\Rightarrow B(c, r) \cap \mathcal{S}$  is a topological arc



if  $s \in P \setminus \{p, q\}$  belongs to this arc, then the arc is tangent to  $\partial B(c, r)$  in  $p$ ,  $q$  or  $s$  (say  $s$ )

$\Rightarrow d(c, P) = r = \|c - s\| \geq \text{lfs}(s) > \varepsilon.$   
(contradiction with the hypothesis of the theorem)

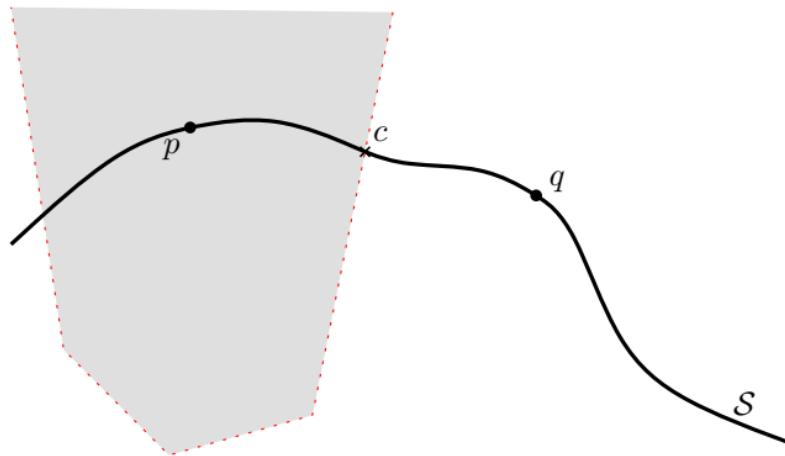
# Approximation power of the restricted Delaunay

Proof for curves:

→ show that every edge of  $\text{Del}_{\mathcal{S}}(P)$  connects consecutive points of  $P$  along  $\mathcal{S}$ , and vice-versa

Let  $c \in \text{arcs}_{\mathcal{S}}(pq) \cap \partial p^*$ .

$c \in ps^*$  for some  $s \in P \setminus \{p\}$

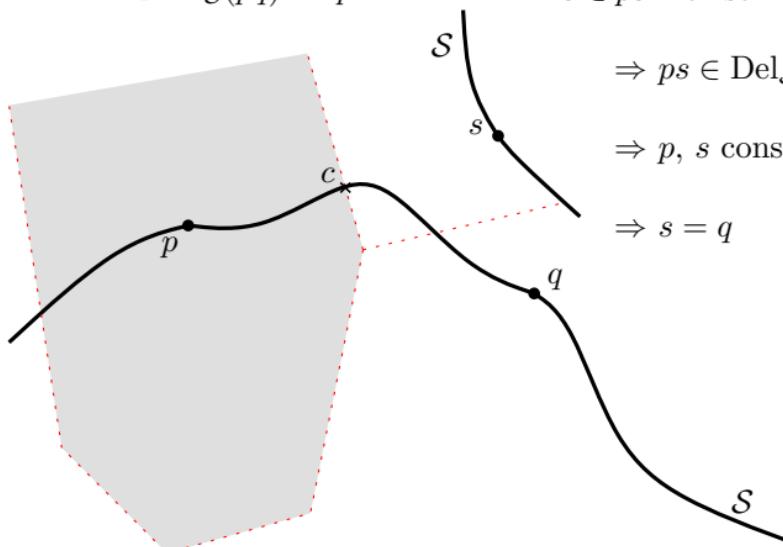


# Approximation power of the restricted Delaunay

Proof for curves:

→ show that every edge of  $\text{Del}_{\mathcal{S}}(P)$  connects consecutive points of  $P$  along  $\mathcal{S}$ , and vice-versa

Let  $c \in \text{arc}_{\mathcal{S}}(pq) \cap \partial p^*$ .



$c \in ps^*$  for some  $s \in P \setminus \{p\}$

$\Rightarrow ps \in \text{Del}_{\mathcal{S}}(P)$

$\Rightarrow p, s$  consecutive along  $\mathcal{S}$ , with  $c \in \text{arc}_{\mathcal{S}}(ps)$

$\Rightarrow s = q$

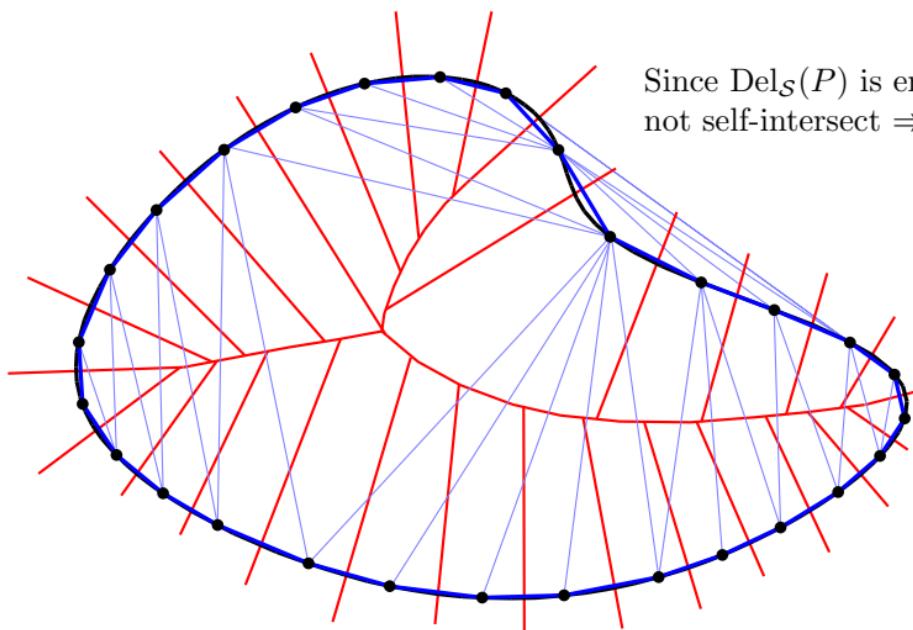
(by previous part of the proof)

# Approximation power of the restricted Delaunay

Proof for curves:

→ show that every edge of  $\text{Del}_{\mathcal{S}}(P)$  connects consecutive points of  $P$  along  $\mathcal{S}$ , and vice-versa

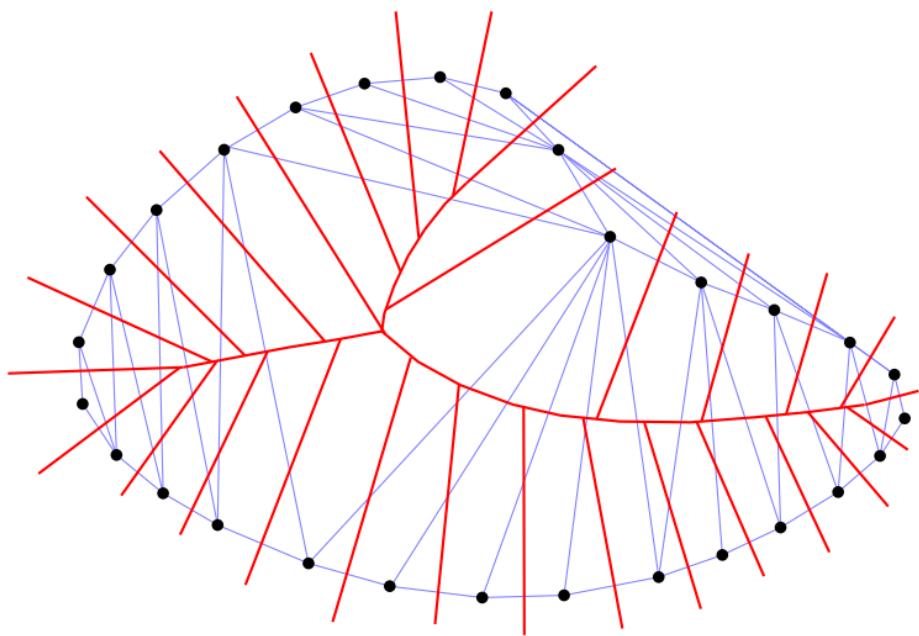
⇒  $\text{Del}_{\mathcal{S}}(P)$  is homeomorphic to  $\mathcal{S}$  between each pair of consecutive points of  $P$



Since  $\text{Del}_{\mathcal{S}}(P)$  is embedded in  $\text{Del}(P)$ , it does not self-intersect ⇒ global homeomorphism

# Computing the restricted Delaunay

Q How to compute  $\text{Del}_{\mathcal{S}}(P)$  when  $\mathcal{S}$  is unknown?

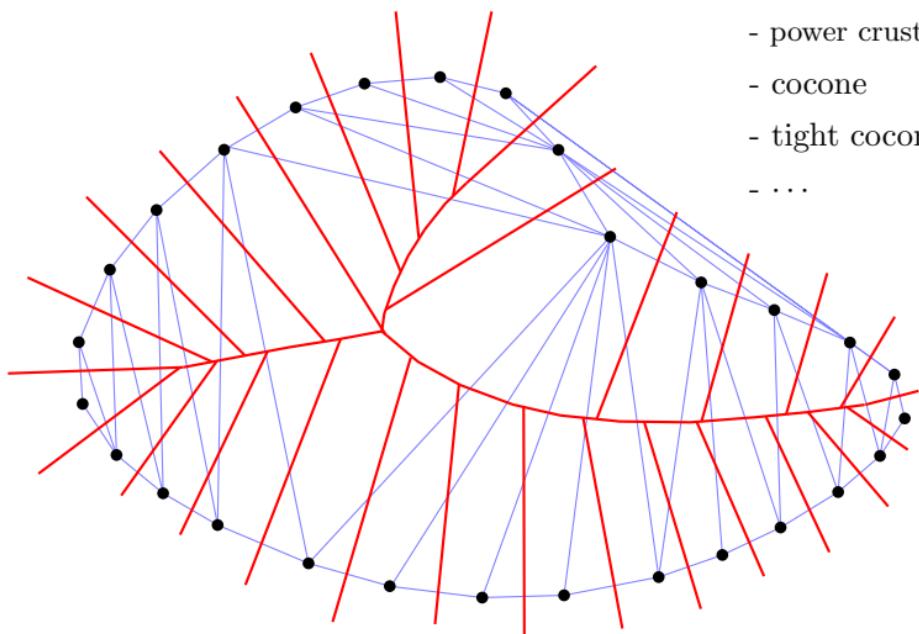


# Computing the restricted Delaunay

Q How to compute  $\text{Del}_{\mathcal{S}}(P)$  when  $\mathcal{S}$  is unknown?

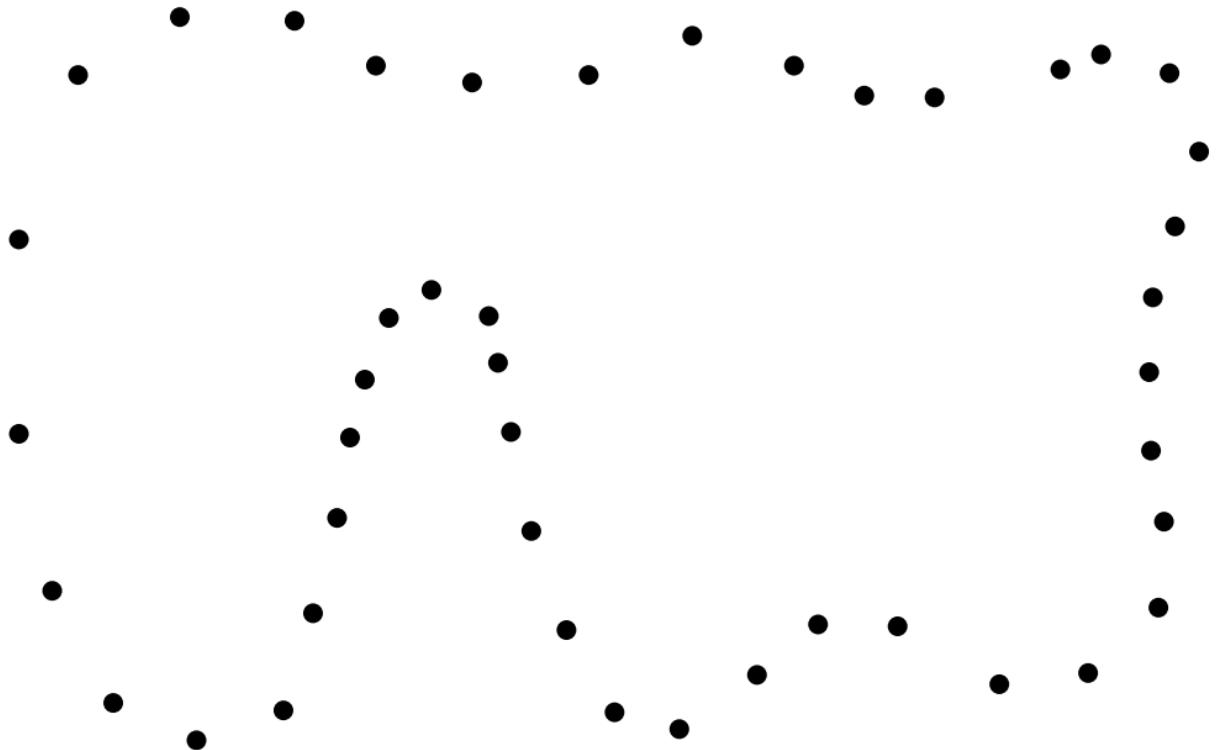
→ a whole family of algorithms use various Delaunay extraction criteria:

- crust
- power crust
- cocone
- tight cocone
- ...



# Crust algorithm

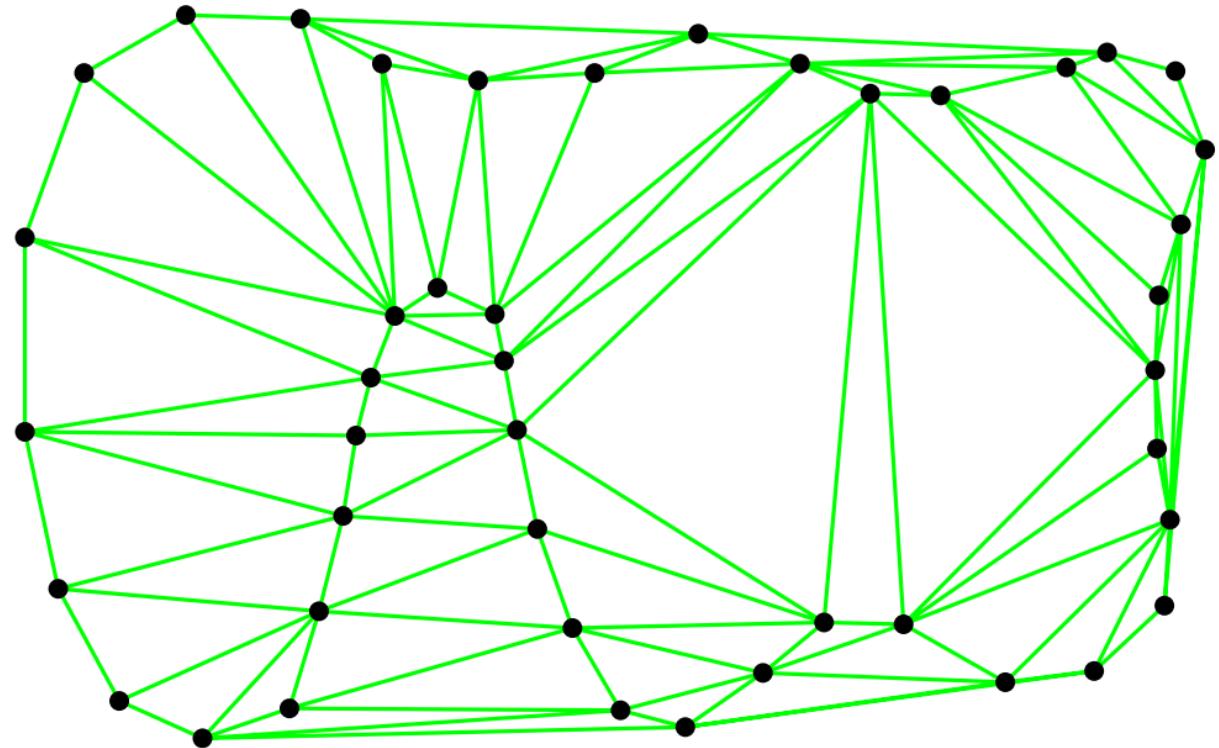
[Amenta et al. 1997-98]



# Crust algorithm

[Amenta et al. 1997-98]

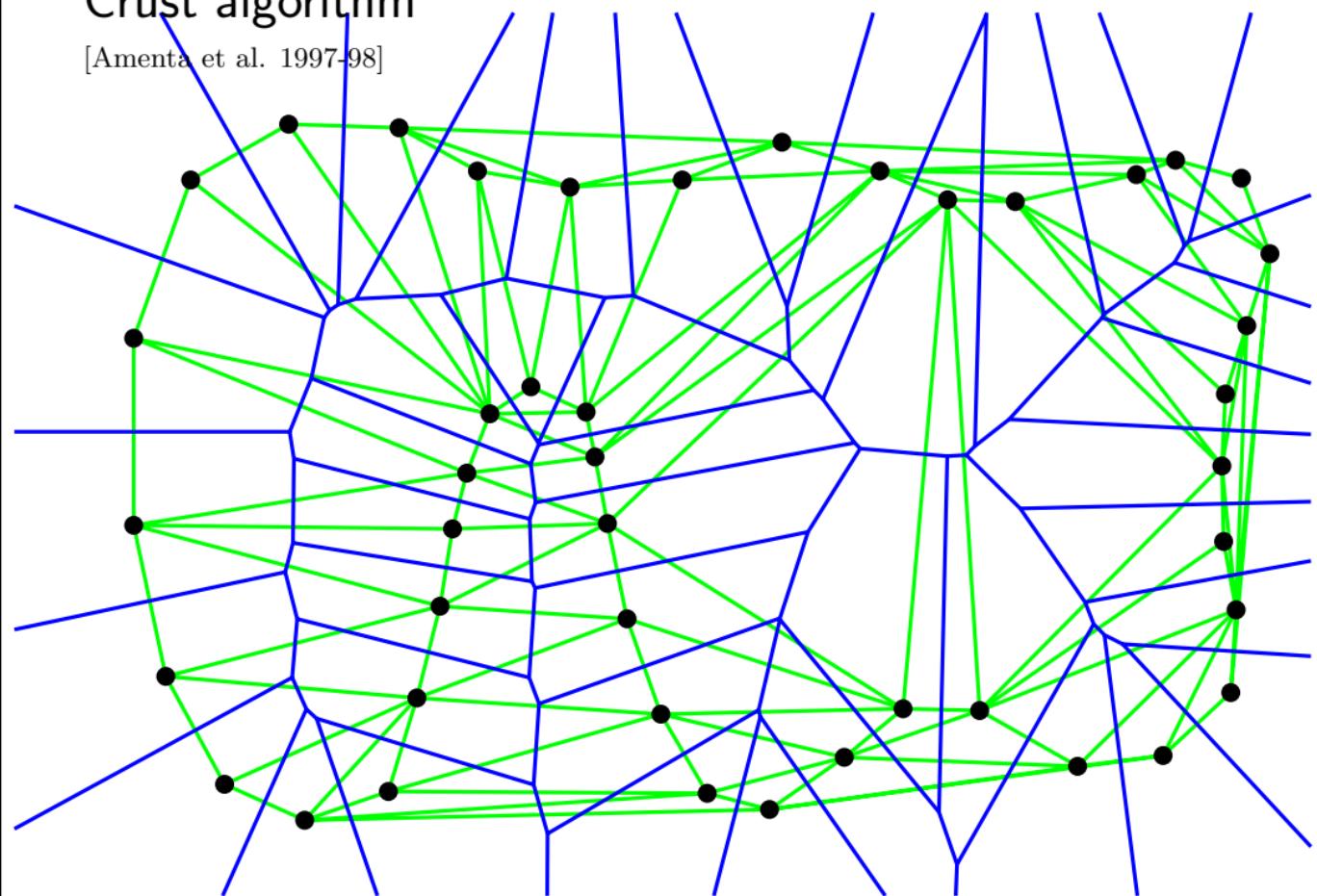
1. Compute Delaunay triangulation of  $P$



# Crust algorithm

[Amenta et al. 1997-98]

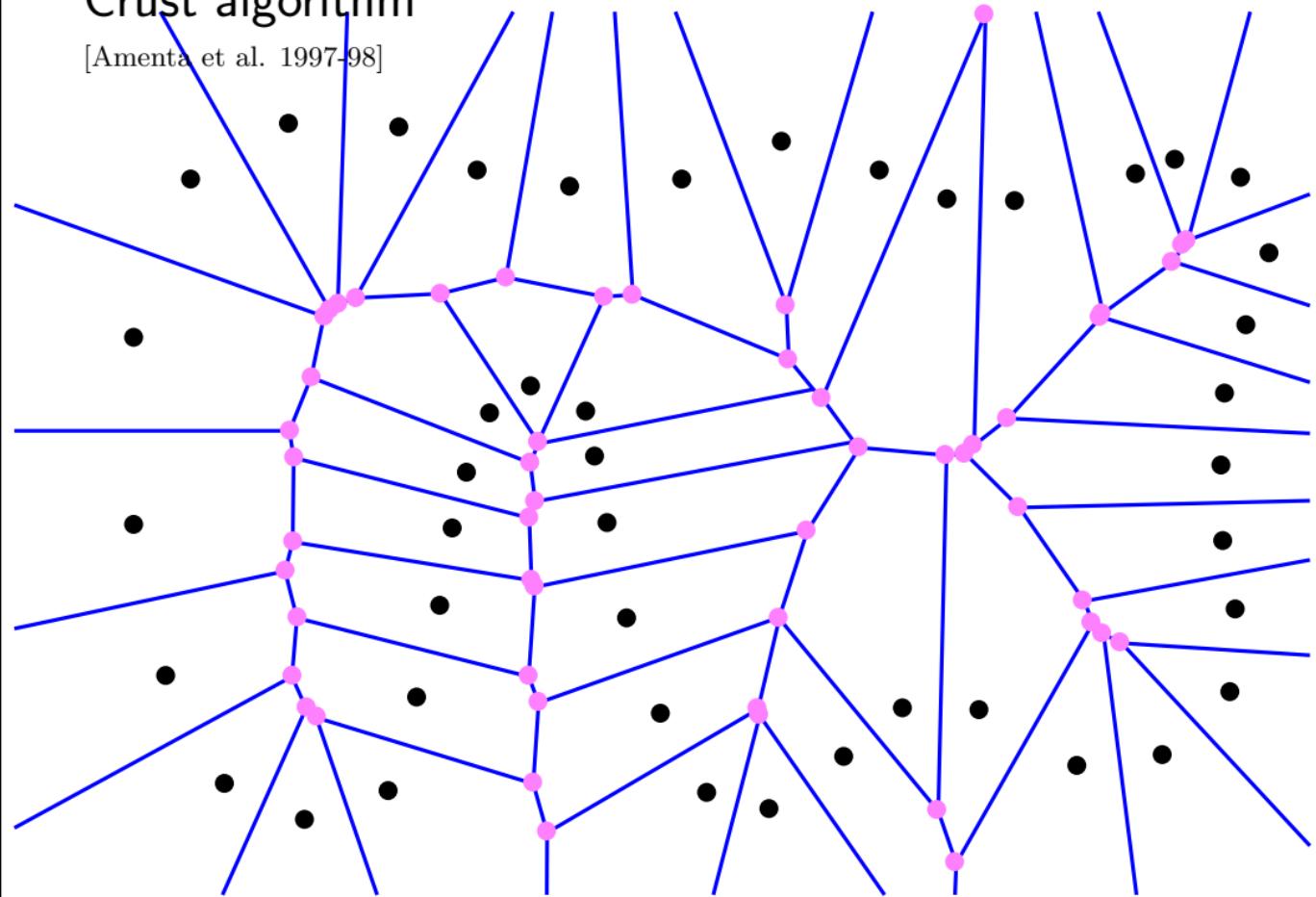
1. Compute Delaunay triangulation of  $P$



# Crust algorithm

[Amenta et al. 1997-98]

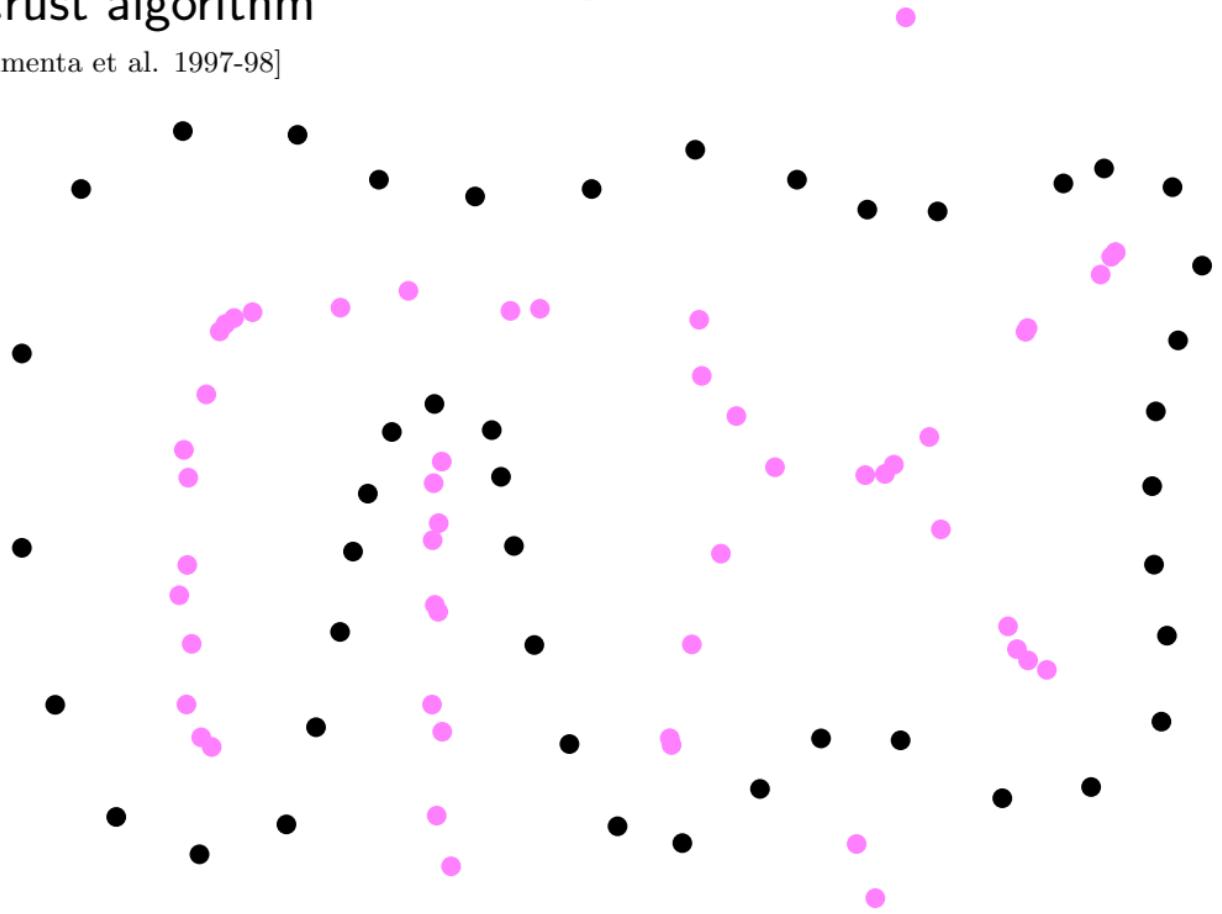
2. Compute *poles* (furthest Voronoi vertices)



# Crust algorithm

[Amenta et al. 1997-98]

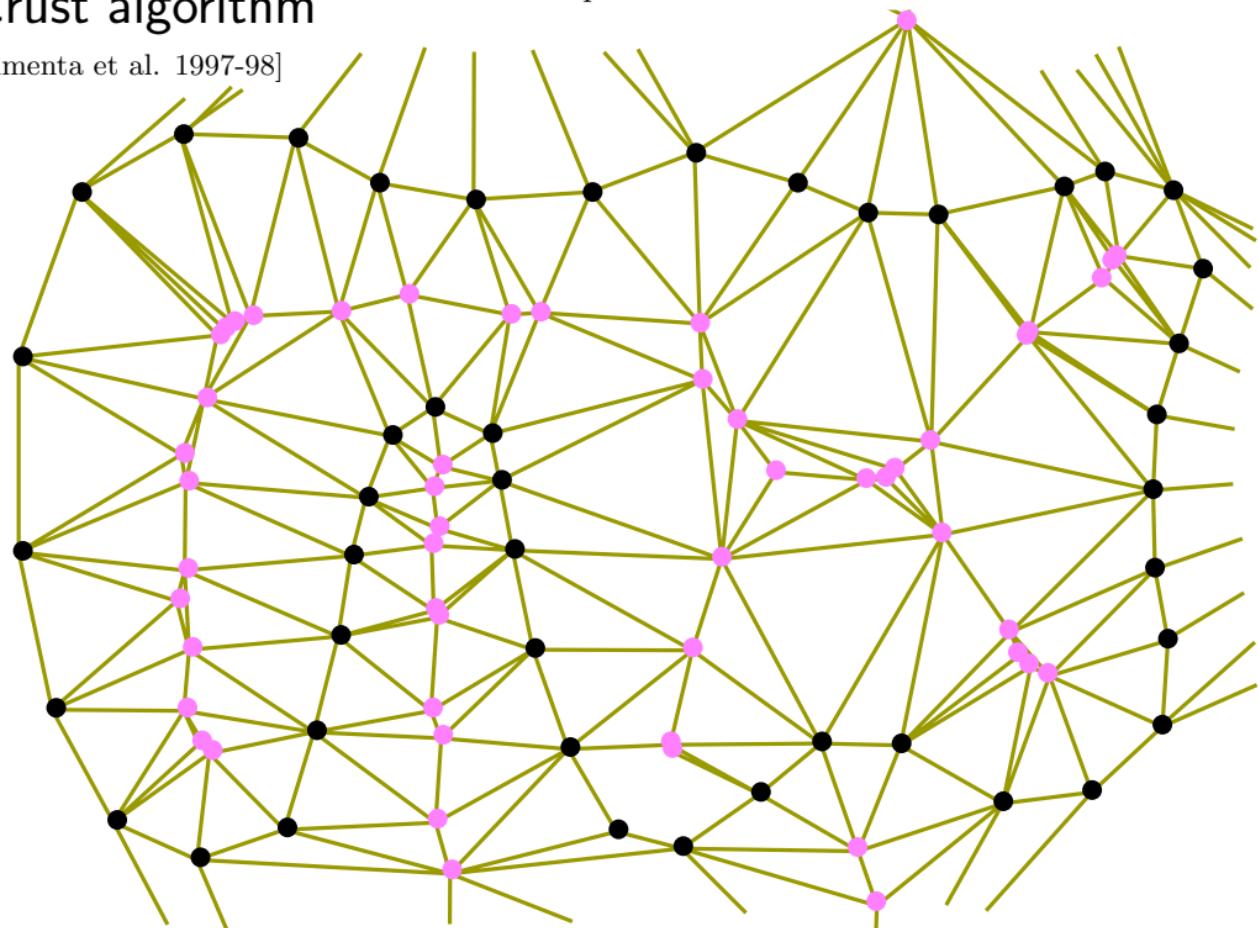
3. Add poles to the set of vertices



# Crust algorithm

[Amenta et al. 1997-98]

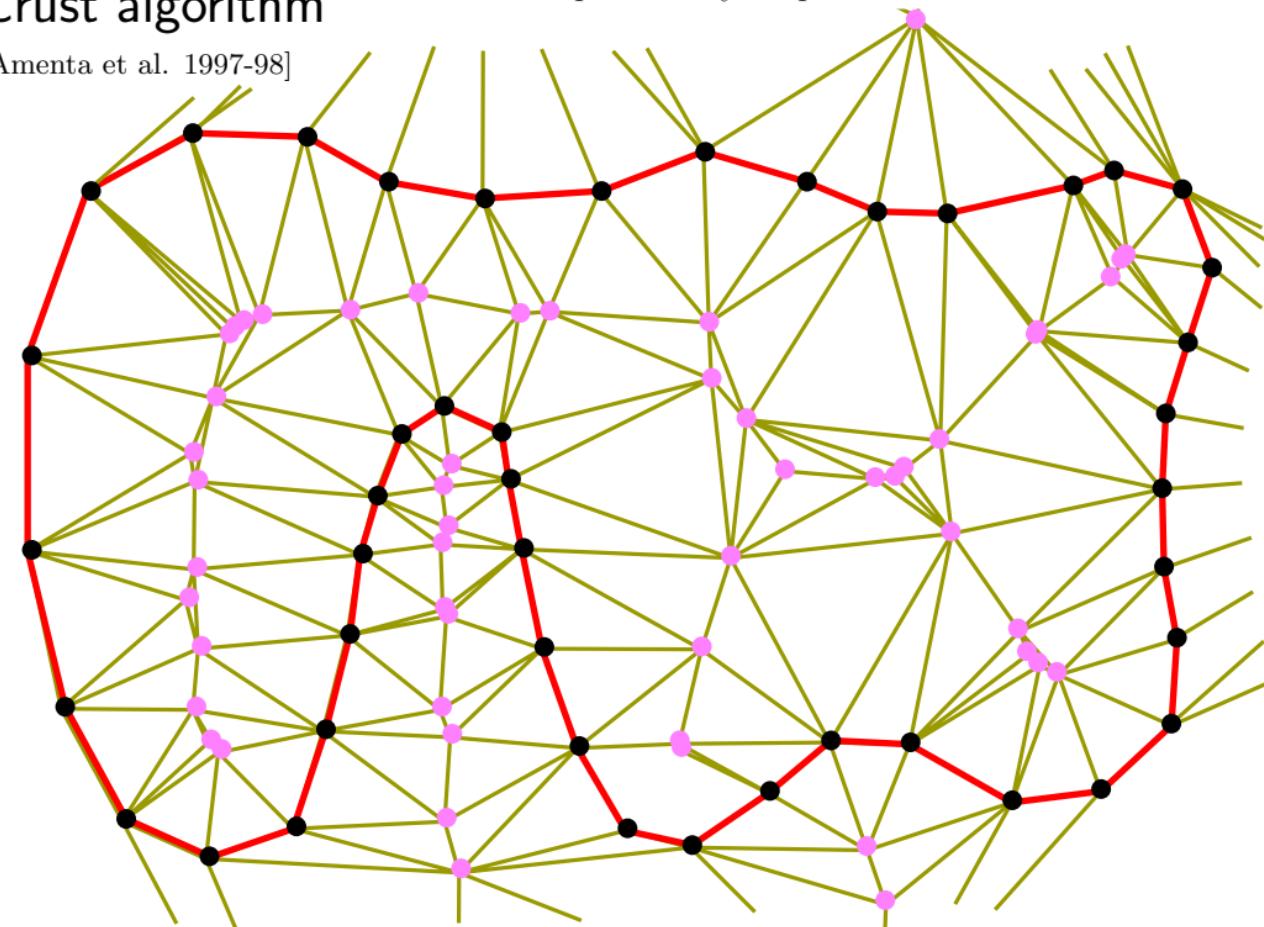
3. Add poles to the set of vertices



# Crust algorithm

[Amenta et al. 1997-98]

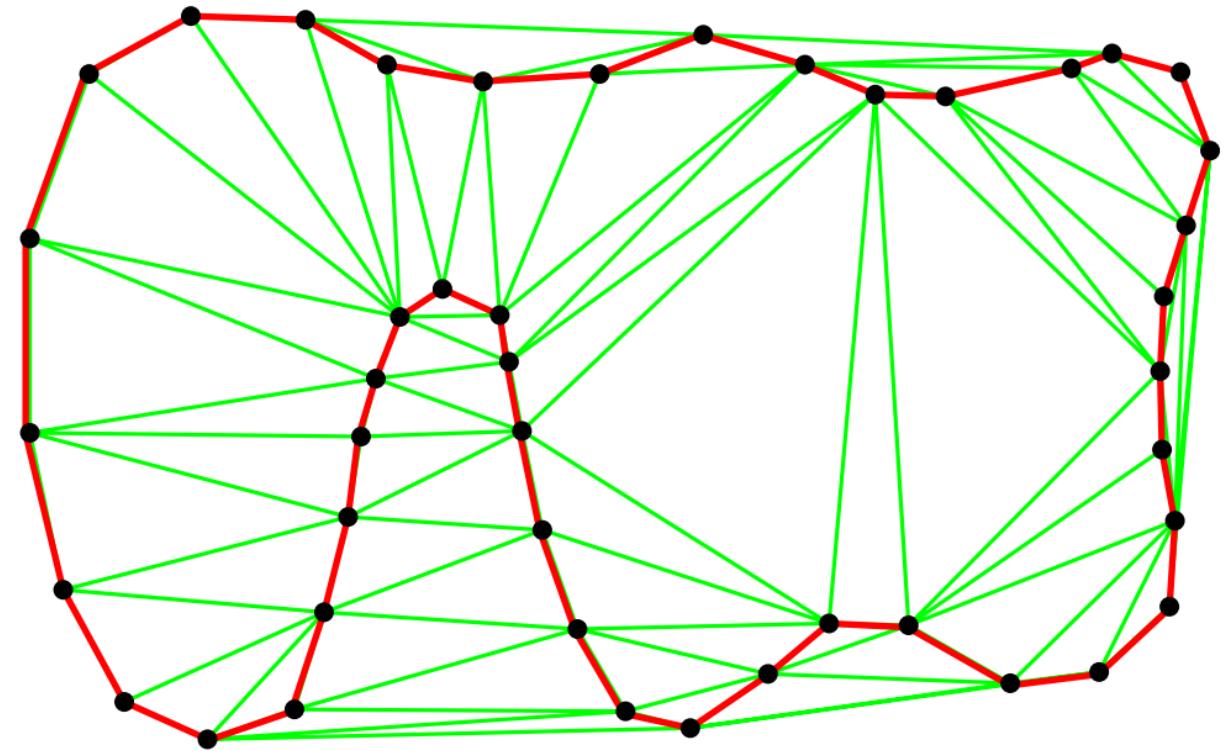
4. Keep Delaunay simplices whose vertices are in  $P$



# Crust algorithm

→ in 2-d, crust =  $\text{Del}_{\mathcal{S}}(P) \approx \mathcal{S}$

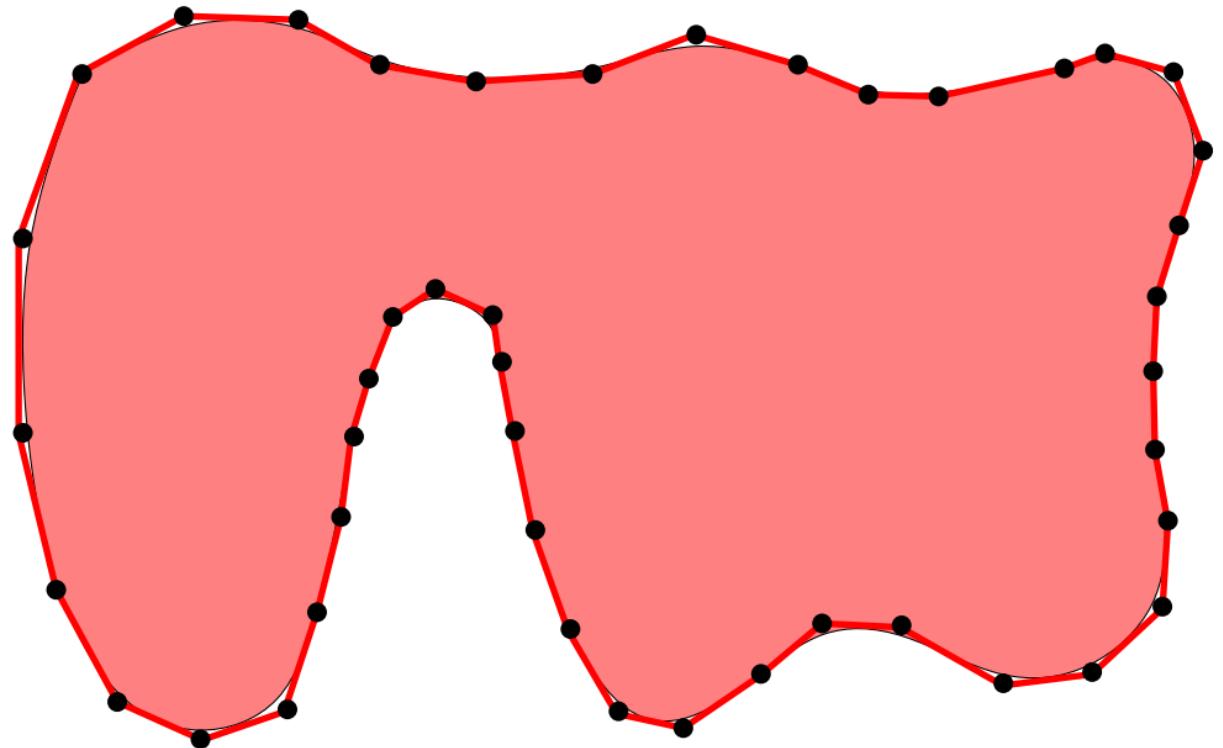
[Amenta et al. 1997-98]



# Crust algorithm

→ in 2-d, crust =  $\text{Del}_{\mathcal{S}}(P) \approx \mathcal{S}$

[Amenta et al. 1997-98]

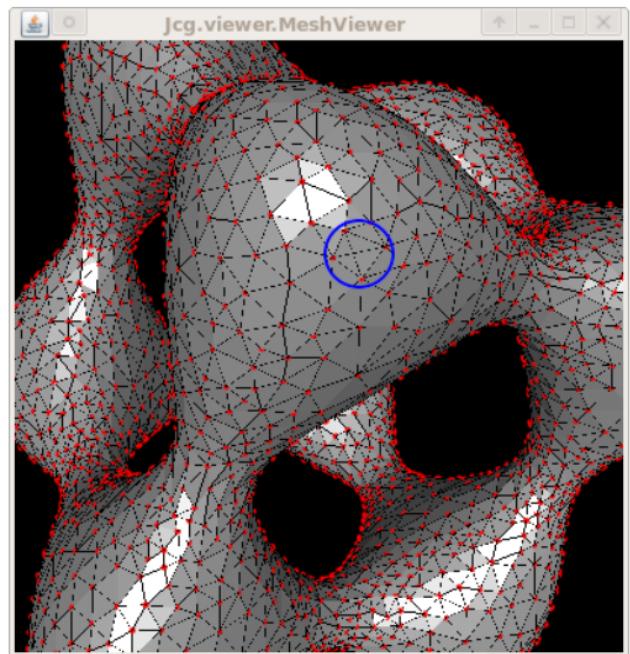
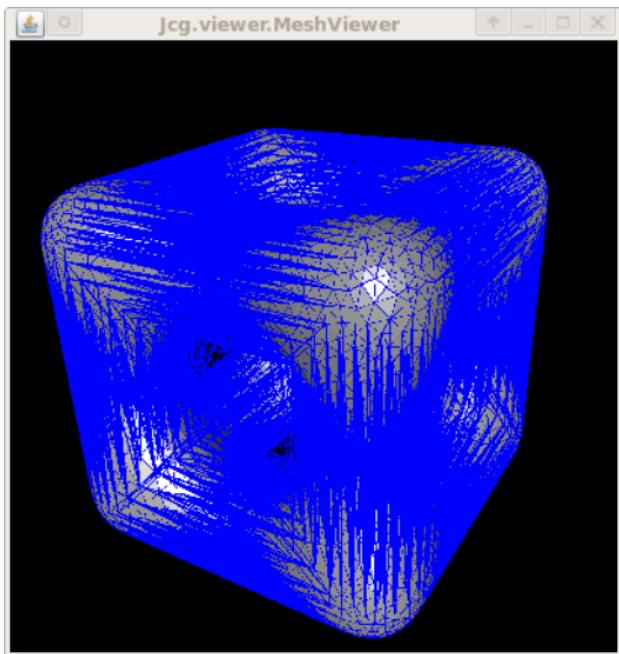


# Crust algorithm

[Amenta et al. 1997-98]

→ in 2-d, crust =  $\text{Del}_{\mathcal{S}}(P) \approx \mathcal{S}$

→ in 3-d, crust  $\supseteq \text{Del}_{\mathcal{S}}(P) \approx \mathcal{S}$



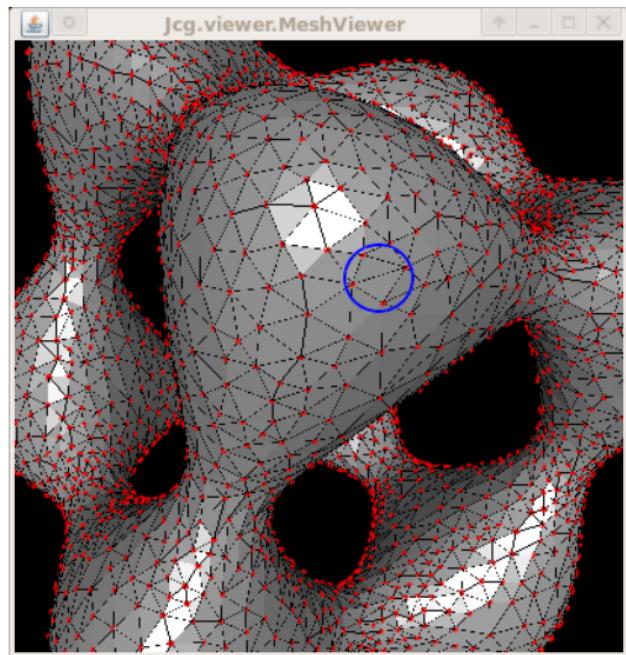
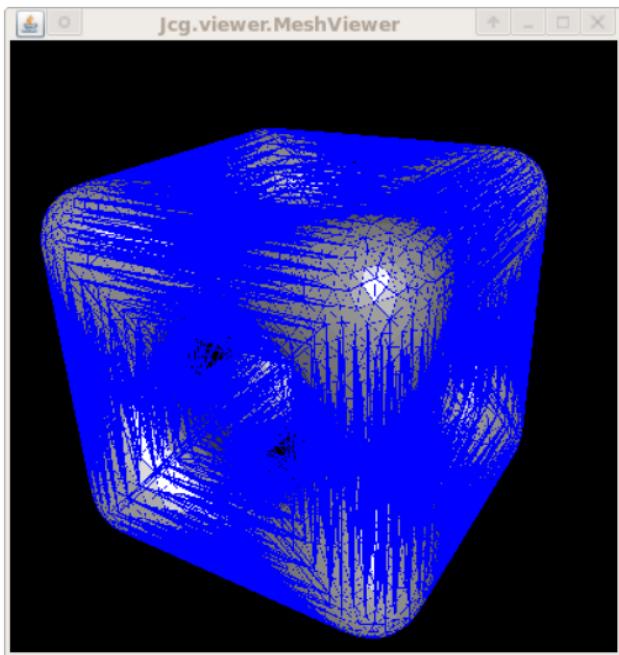
# Crust algorithm

[Amenta et al. 1997-98]

→ in 2-d, crust =  $\text{Del}_{\mathcal{S}}(P) \approx \mathcal{S}$

→ in 3-d, crust  $\supseteq \text{Del}_{\mathcal{S}}(P) \approx \mathcal{S}$

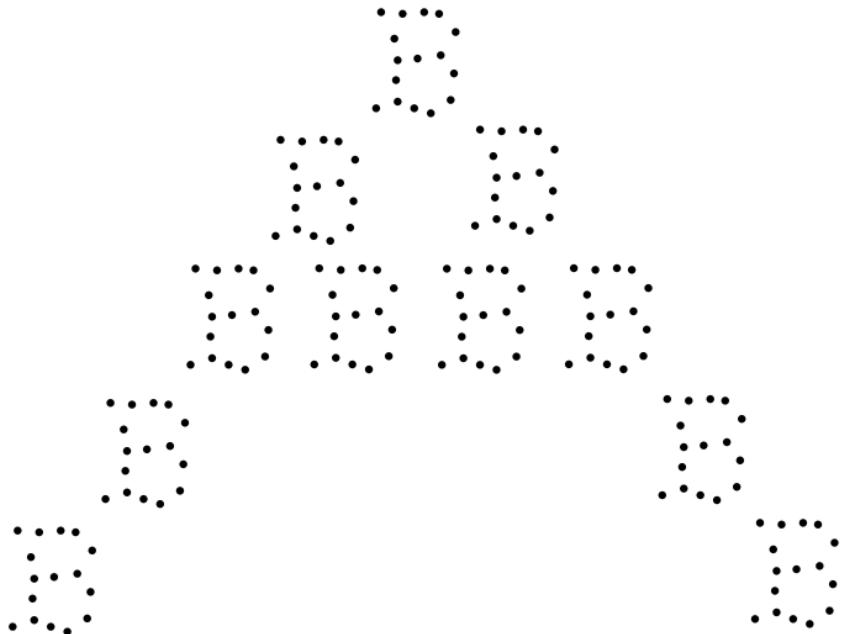
⇒ manifold extraction step in post-processing



# Back to the reconstruction paradigm

**Q** What do you see?

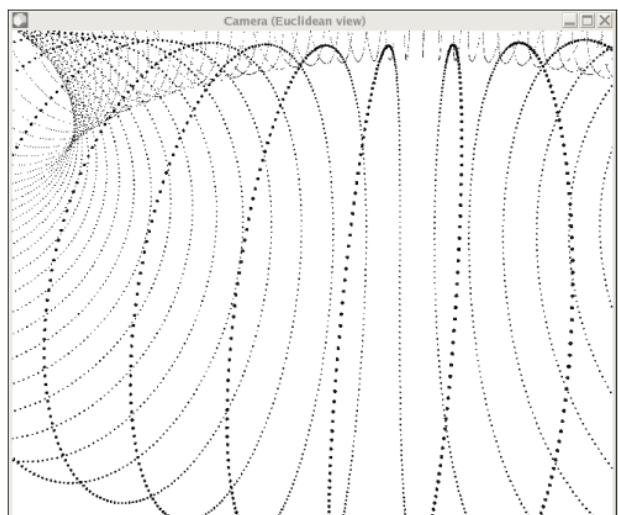
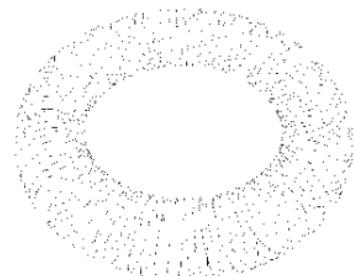
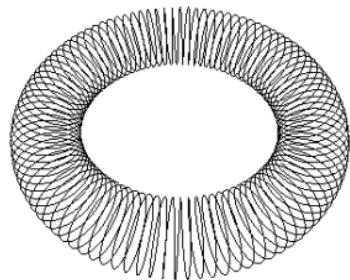
Why?



# Back to the reconstruction paradigm

**Q** What do you see?

Why?

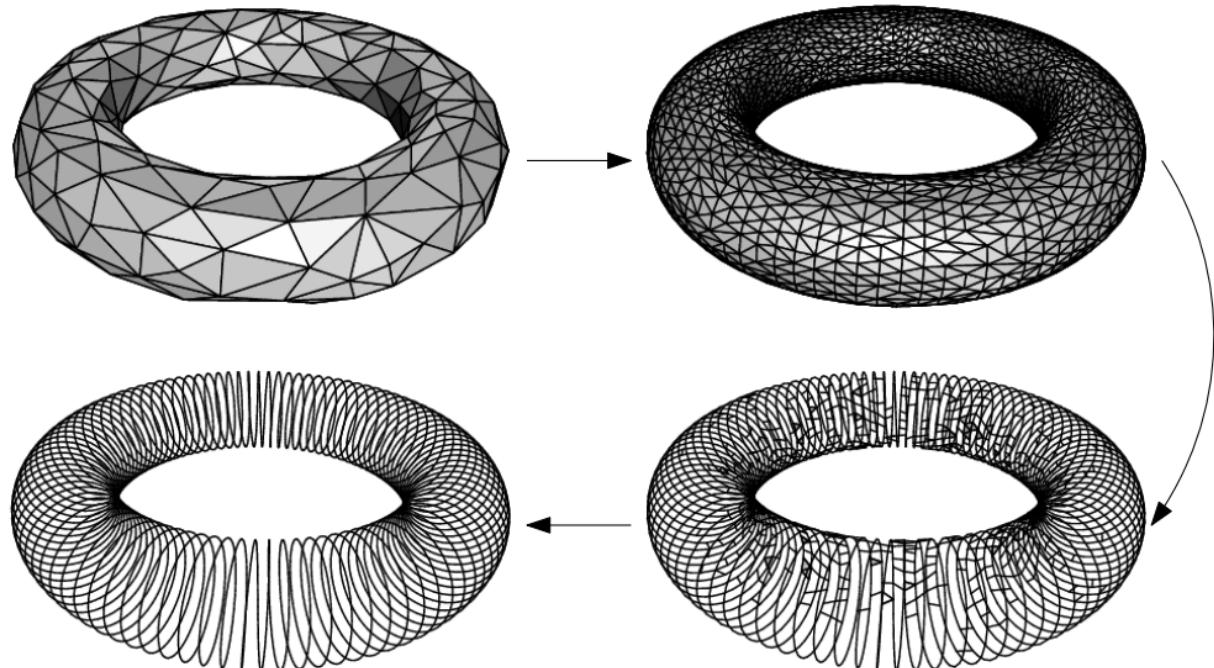


## Back to the reconstruction paradigm

- When the dimensionality of the data is unknown or there is noise, the reconstruction result depends on the scale at which the data is looked at.
- need for multi-scale reconstruction techniques

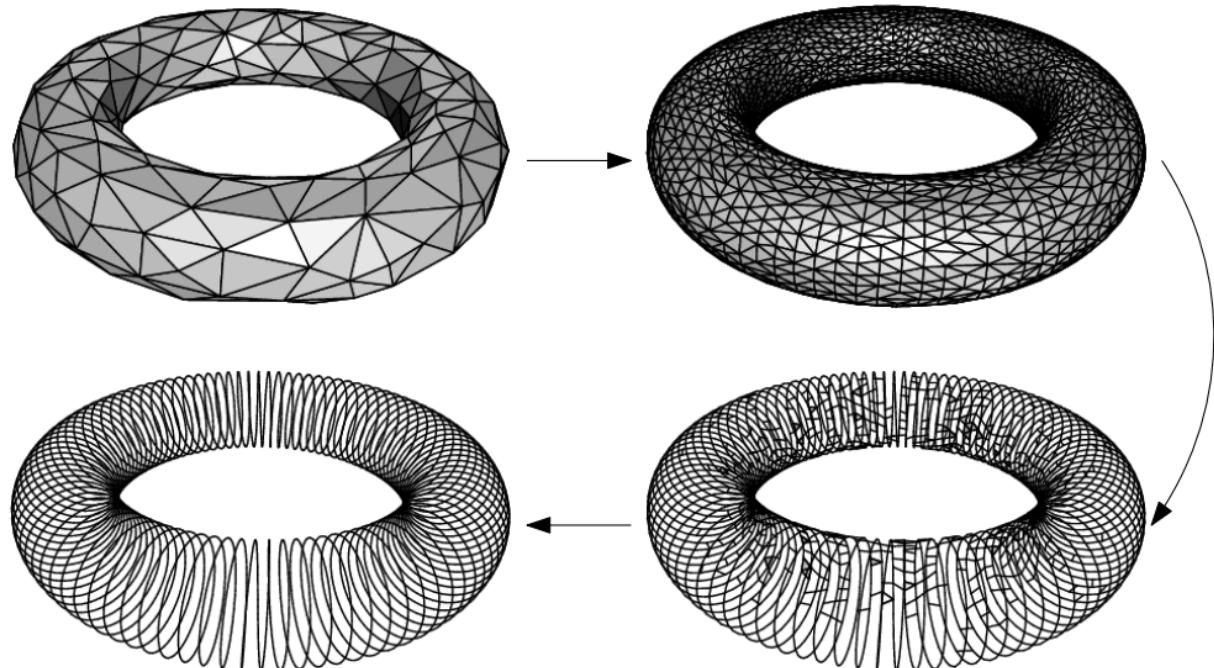
# Multi-scale approach in a nutshell

→ build a one-parameter family of complexes approximating the input at various scales



# Multi-scale approach in a nutshell

→ build a one-parameter family of complexes approximating the input at various scales

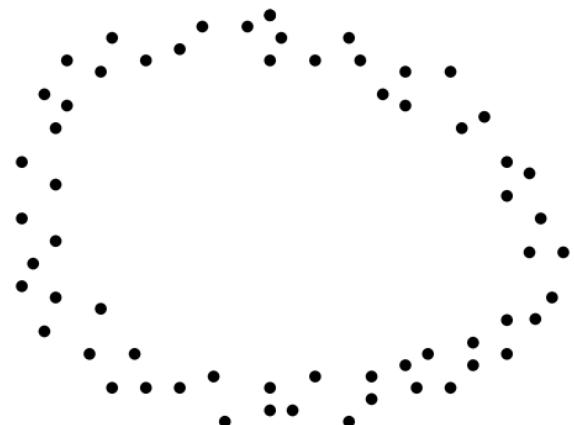


→ connections with manifold learning and topological persistence

# Multi-scale algorithm [Guibas, Oudot 2007]

Input: a finite point set  $W \subset \mathbb{R}^n$

→ resample  $W$  iteratively, and maintain a simplicial complex:

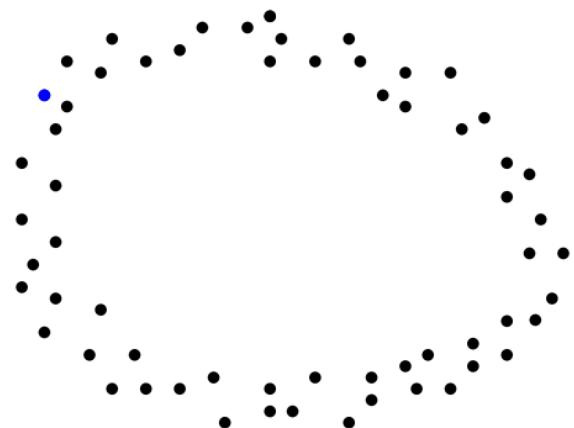


# Multi-scale algorithm [Guibas, Oudot 2007]

Input: a finite point set  $W \subset \mathbb{R}^n$

→ resample  $W$  iteratively, and maintain a simplicial complex:

Let  $L := \{p\}$ , for some  $p \in W$ ;



# Multi-scale algorithm [Guibas, Oudot 2007]

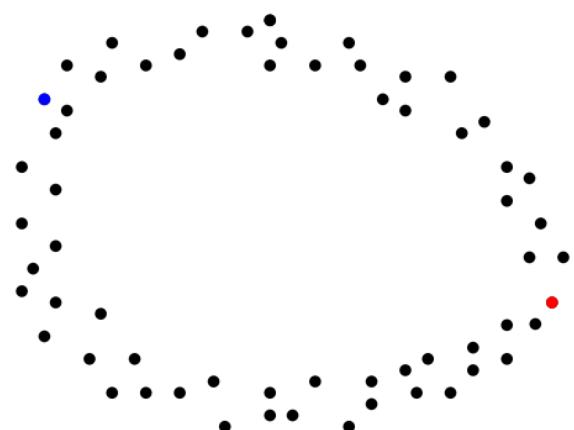
Input: a finite point set  $W \subset \mathbb{R}^n$

→ resample  $W$  iteratively, and maintain a simplicial complex:

Let  $L := \{p\}$ , for some  $p \in W$ ;

WHILE  $L \subsetneq W$

    Let  $q := \operatorname{argmax}_{w \in W} d(w, L)$ ;



# Multi-scale algorithm [Guibas, Oudot 2007]

Input: a finite point set  $W \subset \mathbb{R}^n$

→ resample  $W$  iteratively, and maintain a simplicial complex:

Let  $L := \{p\}$ , for some  $p \in W$ ;

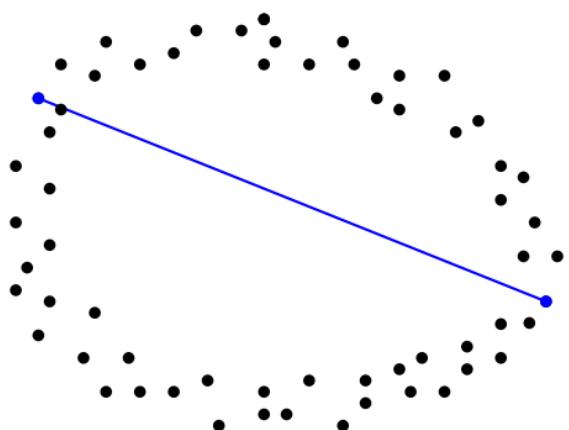
WHILE  $L \subsetneq W$

    Let  $q := \operatorname{argmax}_{w \in W} d(w, L)$ ;

$L := L \cup \{q\}$ ;

    update simplicial complex;

END WHILE



# Multi-scale algorithm [Guibas, Oudot 2007]

Input: a finite point set  $W \subset \mathbb{R}^n$

→ resample  $W$  iteratively, and maintain a simplicial complex:

Let  $L := \{p\}$ , for some  $p \in W$ ;

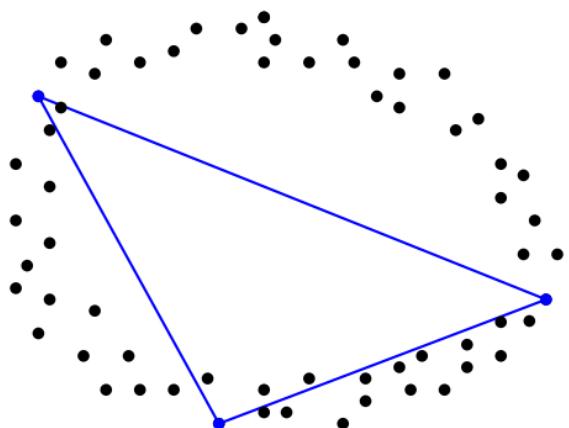
WHILE  $L \subsetneq W$

    Let  $q := \operatorname{argmax}_{w \in W} d(w, L)$ ;

$L := L \cup \{q\}$ ;

    update simplicial complex;

END WHILE



# Multi-scale algorithm [Guibas, Oudot 2007]

Input: a finite point set  $W \subset \mathbb{R}^n$

→ resample  $W$  iteratively, and maintain a simplicial complex:

Let  $L := \{p\}$ , for some  $p \in W$ ;

WHILE  $L \subsetneq W$

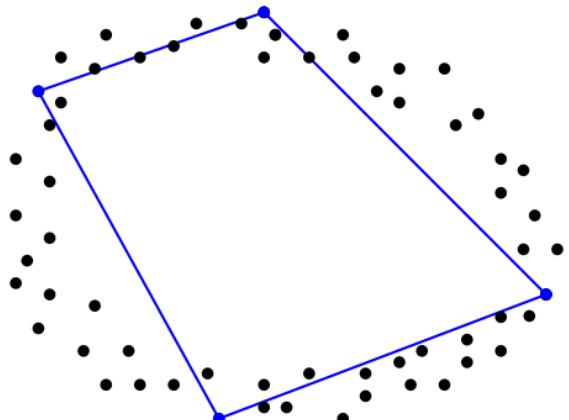
    Let  $q := \operatorname{argmax}_{w \in W} d(w, L)$ ;

$L := L \cup \{q\}$ ;

    update simplicial complex;

END WHILE

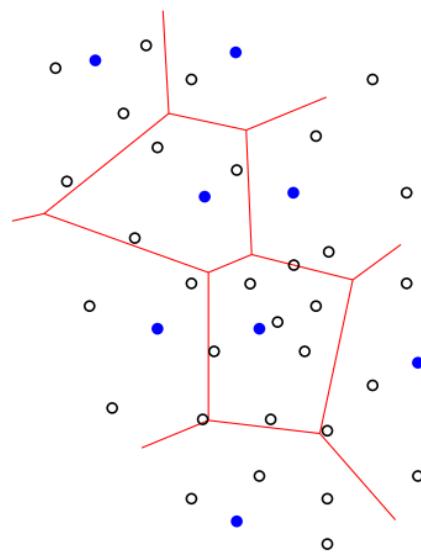
Output: the sequence of simplicial complexes



# The simplicial complex to maintain

→ maintain the **witness complex**  $C^W(L)$  [de Silva 2003]:

Let  $L \subseteq \mathbb{R}^d$  (landmarks) s.t.  $|L| < +\infty$  and  $W \subseteq \mathbb{R}^d$  (witnesses)

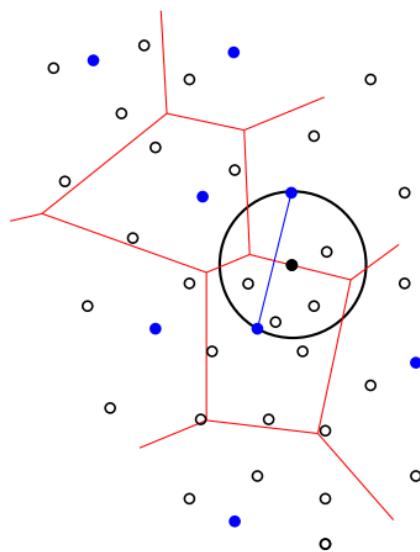


# The simplicial complex to maintain

→ maintain the **witness complex**  $C^W(L)$  [de Silva 2003]:

Let  $L \subseteq \mathbb{R}^d$  (landmarks) s.t.  $|L| < +\infty$  and  $W \subseteq \mathbb{R}^d$  (witnesses)

**Def.**  $w \in W$  strongly witnesses  $[v_0, \dots, v_k]$  if  $\|w - v_i\| = \|w - v_j\| \leq \|w - u\|$  for all  $i, j = 0, \dots, k$  and all  $u \in L \setminus \{v_0, \dots, v_k\}$  (Delaunay test)



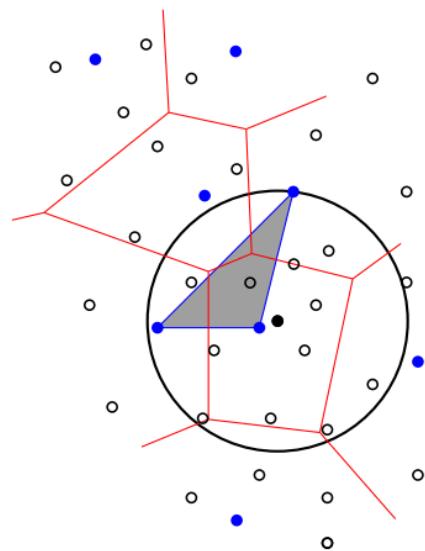
# The simplicial complex to maintain

→ maintain the **witness complex**  $C^W(L)$  [de Silva 2003]:

Let  $L \subseteq \mathbb{R}^d$  (landmarks) s.t.  $|L| < +\infty$  and  $W \subseteq \mathbb{R}^d$  (witnesses)

**Def.**  $w \in W$  strongly witnesses  $[v_0, \dots, v_k]$  if  $\|w - v_i\| = \|w - v_j\| \leq \|w - u\|$  for all  $i, j = 0, \dots, k$  and all  $u \in L \setminus \{v_0, \dots, v_k\}$  (Delaunay test)

**Def.**  $w \in W$  weakly witnesses  $[v_0, \dots, v_k]$  if  $\|w - v_i\| \leq \|w - u\|$  for all  $i = 0, \dots, k$  and all  $u \in L \setminus \{v_0, \dots, v_k\}$ .



# The simplicial complex to maintain

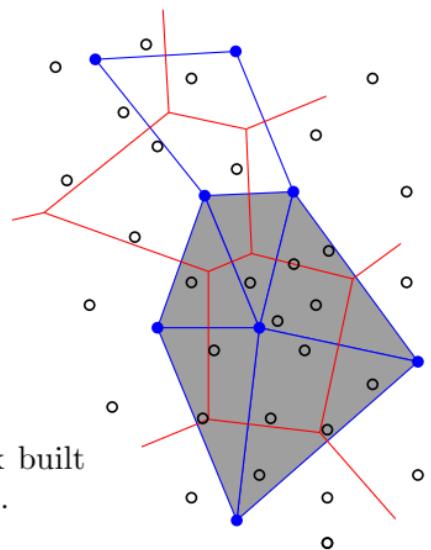
→ maintain the **witness complex**  $C^W(L)$  [de Silva 2003]:

Let  $L \subseteq \mathbb{R}^d$  (landmarks) s.t.  $|L| < +\infty$  and  $W \subseteq \mathbb{R}^d$  (witnesses)

**Def.**  $w \in W$  strongly witnesses  $[v_0, \dots, v_k]$  if  $\|w - v_i\| = \|w - v_j\| \leq \|w - u\|$  for all  $i, j = 0, \dots, k$  and all  $u \in L \setminus \{v_0, \dots, v_k\}$  (Delaunay test)

**Def.**  $w \in W$  weakly witnesses  $[v_0, \dots, v_k]$  if  $\|w - v_i\| \leq \|w - u\|$  for all  $i = 0, \dots, k$  and all  $u \in L \setminus \{v_0, \dots, v_k\}$ .

**Def.**  $C^W(L)$  is the largest abstract simplicial complex built over  $L$ , whose faces are weakly witnessed by points of  $W$ .

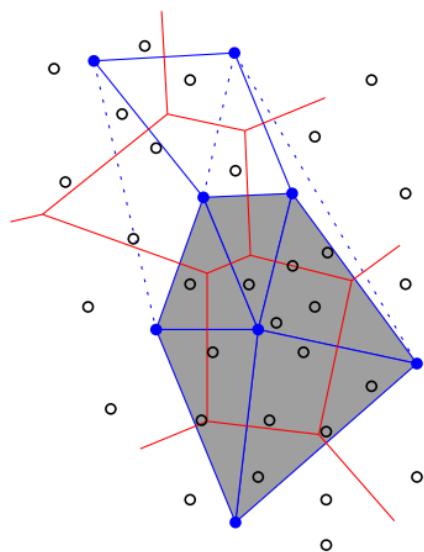


# The witness complex (properties)

**Thm. 1** [de Silva 2003]  $\forall W, L, \forall \sigma \in C^W(L), \exists c \in \mathbb{R}^d$  that strongly witnesses  $\sigma$ .

$\Rightarrow C^W(L)$  is a subcomplex of  $\text{Del}(L)$

$\Rightarrow C^W(L)$  is embedded in  $\mathbb{R}^d$



# The witness complex (properties)

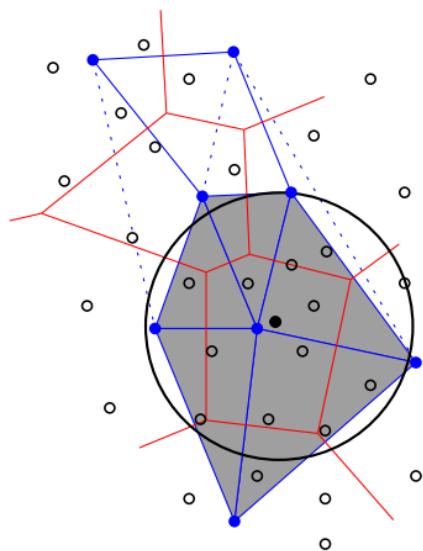
**Thm. 1** [de Silva 2003]  $\forall W, L, \forall \sigma \in C^W(L), \exists c \in \mathbb{R}^d$  that strongly witnesses  $\sigma$ .

$\Rightarrow C^W(L)$  is a subcomplex of  $\text{Del}(L)$

$\Rightarrow C^W(L)$  is embedded in  $\mathbb{R}^d$

**Thm. 2** [de Silva, Carlsson 2004]

- The size of  $C^W(L)$  is  $O(d|W|)$
- The time to compute  $C^W(L)$  is  $O(d|W||L|)$



# The witness complex (properties)

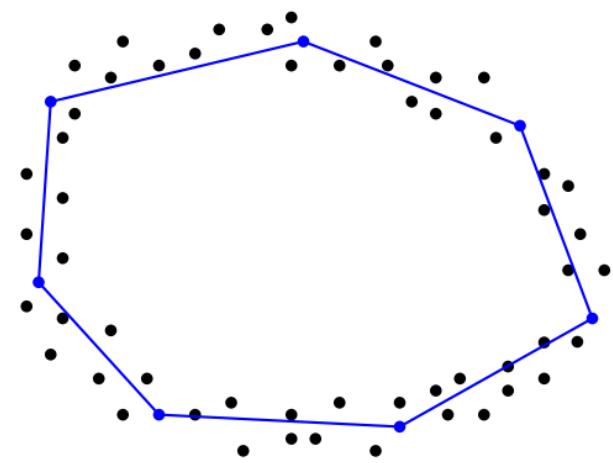
**Thm. 1** [de Silva 2003]  $\forall W, L, \forall \sigma \in C^W(L), \exists c \in \mathbb{R}^d$  that strongly witnesses  $\sigma$ .

$\Rightarrow C^W(L)$  is a subcomplex of  $\text{Del}(L)$

$\Rightarrow C^W(L)$  is embedded in  $\mathbb{R}^d$

**Thm. 2** [de Silva, Carlsson 2004]

- The size of  $C^W(L)$  is  $O(d|W|)$
- The time to compute  $C^W(L)$  is  $O(d|W||L|)$



**Thm. 3** [Guibas, Oudot 2007]

[Attali, Edelsbrunner, Mileyko 2007]

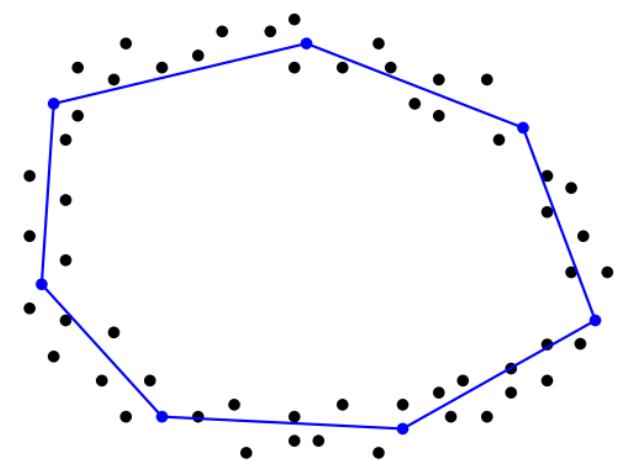
Under some conditions,  $C^W(L) = \text{Del}_S(L) \approx S$

# The witness complex (properties)

→ connection with reconstruction:

- $W \subset \mathbb{R}^d$  is given as input
- $L \subseteq W$  is generated
- underlying manifold  $\mathcal{S}$  unknown
- only distance comparisons

⇒ algorithm is applicable in any metric space



# The witness complex (properties)

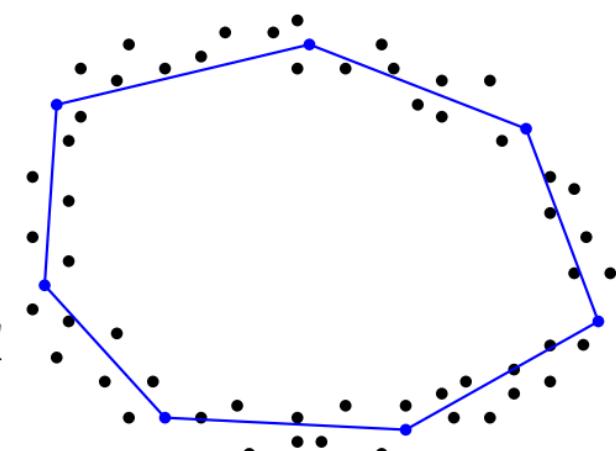
→ connection with reconstruction:

- $W \subset \mathbb{R}^d$  is given as input
- $L \subseteq W$  is generated
- underlying manifold  $\mathcal{S}$  unknown
- only distance comparisons

⇒ algorithm is applicable in any metric space

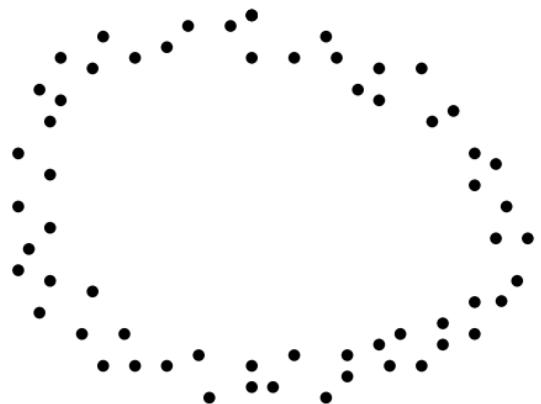
- In  $\mathbb{R}^d$ ,  $C^W(L)$  can be maintained by updating, for each witness  $w$ , the list of  $d + 1$  nearest landmarks of  $w$ .

⇒ space  $\leq O(d|W|)$   
⇒ time  $\leq O(d|W|^2)$



# The full algorithm

Input: a finite point set  $W \subset \mathbb{R}^d$ .

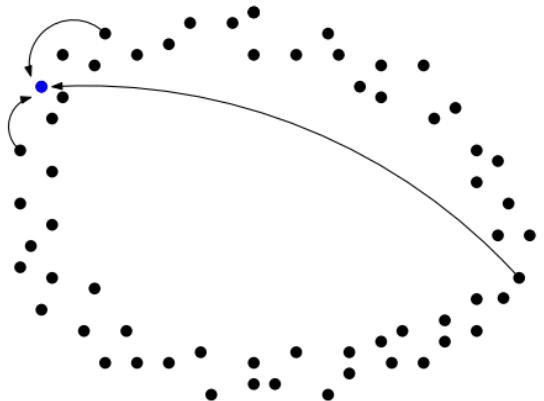


# The full algorithm

Input: a finite point set  $W \subset \mathbb{R}^d$ .

Init:  $L := \{p\}$ ; construct lists of nearest landmarks;  $C^W(L) = \{[p]\}$ ;

Invariant:  $\forall w \in W$ , the list of  $d + 1$  nearest landmarks of  $w$  is maintained throughout the process.



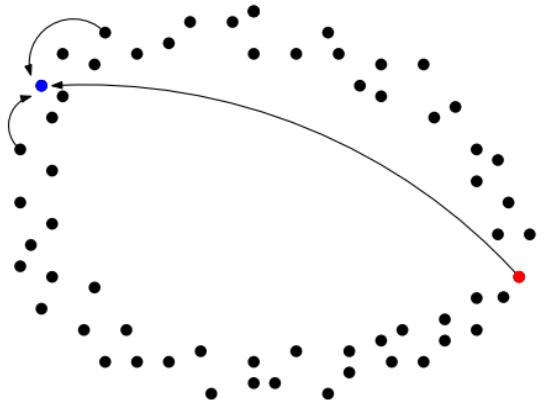
# The full algorithm

Input: a finite point set  $W \subset \mathbb{R}^d$ .

Init:  $L := \{p\}$ ; construct lists of nearest landmarks;  $C^W(L) = \{[p]\}$ ;

Invariant:  $\forall w \in W$ , the list of  $d + 1$  nearest landmarks of  $w$  is maintained throughout the process.

WHILE  $L \subsetneq W$



# The full algorithm

Input: a finite point set  $W \subset \mathbb{R}^d$ .

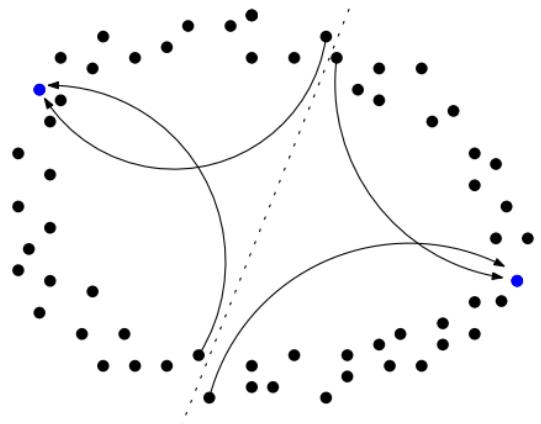
Init:  $L := \{p\}$ ; construct lists of nearest landmarks;  $C^W(L) = \{[p]\}$ ;

Invariant:  $\forall w \in W$ , the list of  $d + 1$  nearest landmarks of  $w$  is maintained throughout the process.

WHILE  $L \subsetneq W$

    insert  $\operatorname{argmax}_{w \in W} d(w, L)$  in  $L$ ;

    update the lists of nearest neighbors;



# The full algorithm

Input: a finite point set  $W \subset \mathbb{R}^d$ .

Init:  $L := \{p\}$ ; construct lists of nearest landmarks;  $C^W(L) = \{[p]\}$ ;

Invariant:  $\forall w \in W$ , the list of  $d + 1$  nearest landmarks of  $w$  is maintained throughout the process.

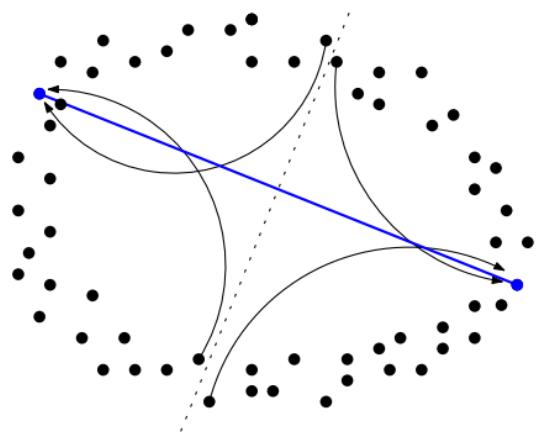
WHILE  $L \subsetneq W$

    insert  $\operatorname{argmax}_{w \in W} d(w, L)$  in  $L$ ;

    update the lists of nearest neighbors;

    update  $C^W(L)$ ;

END WHILE



# The full algorithm

Input: a finite point set  $W \subset \mathbb{R}^d$ .

Init:  $L := \{p\}$ ; construct lists of nearest landmarks;  $C^W(L) = \{[p]\}$ ;

Invariant:  $\forall w \in W$ , the list of  $d + 1$  nearest landmarks of  $w$  is maintained throughout the process.

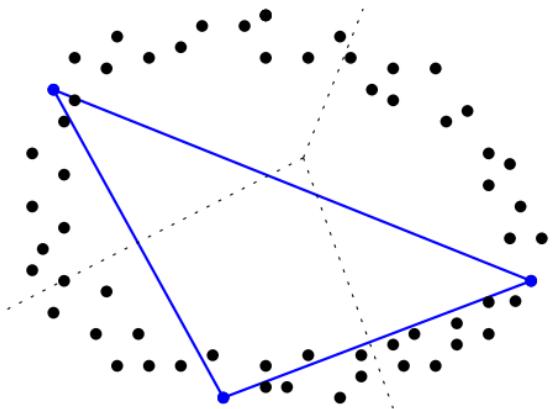
WHILE  $L \subsetneq W$

    insert  $\operatorname{argmax}_{w \in W} d(w, L)$  in  $L$ ;

    update the lists of nearest neighbors;

    update  $C^W(L)$ ;

END WHILE



# The full algorithm

Input: a finite point set  $W \subset \mathbb{R}^d$ .

Init:  $L := \{p\}$ ; construct lists of nearest landmarks;  $C^W(L) = \{[p]\}$ ;

Invariant:  $\forall w \in W$ , the list of  $d + 1$  nearest landmarks of  $w$  is maintained throughout the process.

WHILE  $L \subsetneq W$

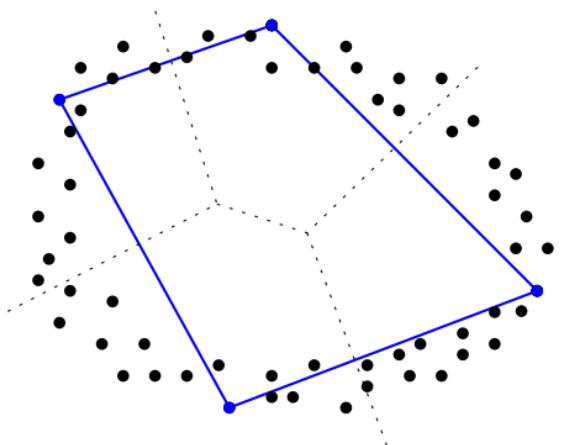
    insert  $\operatorname{argmax}_{w \in W} d(w, L)$  in  $L$ ;

    update the lists of nearest neighbors;

    update  $C^W(L)$ ;

END WHILE

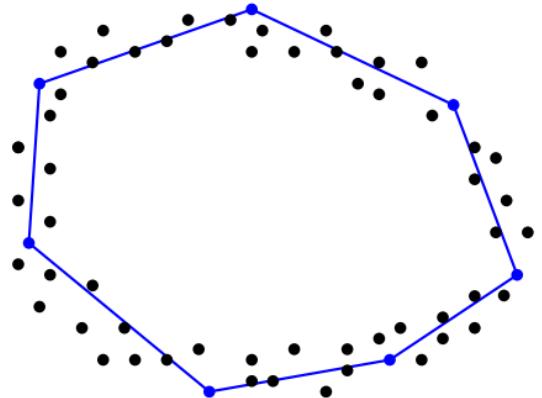
Output: the sequence of complexes  $C^W(L)$



# Theoretical guarantees

→ case of curves:

**Conjecture** [Carlsson, de Silva 2004]:  
 $C^W(L)$  coincides with  $\text{Del}_S(L)$ ...

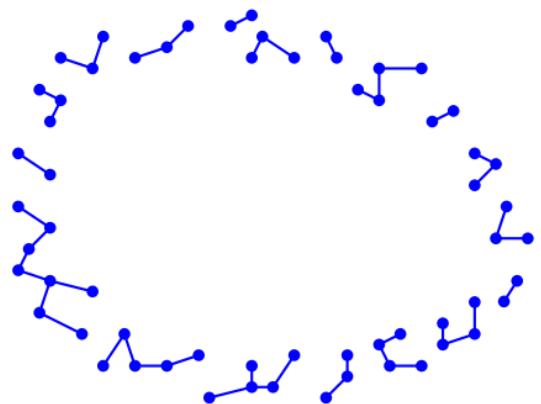


# Theoretical guarantees

→ case of curves:

**Conjecture** [Carlsson, de Silva 2004]:  
 $C^W(L)$  coincides with  $\text{Del}_S(L)$ ...

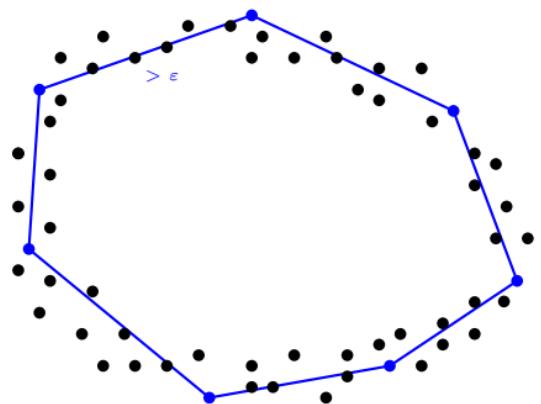
... under some conditions on  $W$  and  $L$



# Theoretical guarantees

→ case of curves:

**Thm. 3** If  $\mathcal{S}$  is a closed curve with positive reach,  $W \subset \mathbb{R}^d$  s.t.  $d_H(W, \mathcal{S}) \leq \delta$ ,  $L \subseteq W$   $\varepsilon$ -sparse  $\varepsilon$ -sample of  $W$  with  $\delta \ll \varepsilon \ll \varrho_{\mathcal{S}}$ , then  $C^W(L) = \text{Del}_{\mathcal{S}}(L) \approx \mathcal{S}$ .

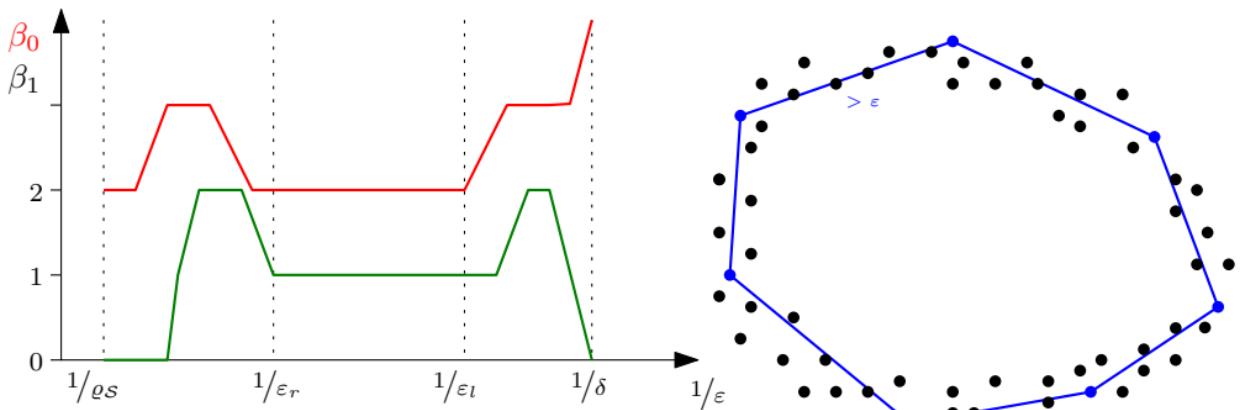


# Theoretical guarantees

→ case of curves:

**Thm. 3** If  $\mathcal{S}$  is a closed curve with positive reach,  $W \subset \mathbb{R}^d$  s.t.  $d_H(W, \mathcal{S}) \leq \delta$ ,  $L \subseteq W$   $\varepsilon$ -sparse  $\varepsilon$ -sample of  $W$  with  $\delta \ll \varepsilon \ll \varrho_{\mathcal{S}}$ , then  $C^W(L) = \text{Del}_{\mathcal{S}}(L) \approx \mathcal{S}$ .

$$\varepsilon_l \quad \varepsilon_r$$



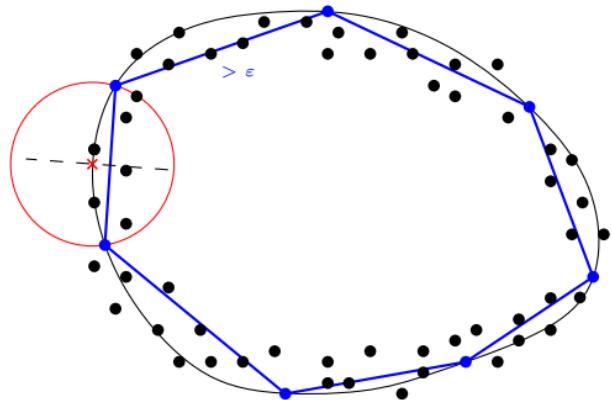
→ There is a plateau in the diagram of Betti numbers of  $C^W(L)$ .

# Theoretical guarantees

→ case of curves:

**Thm. 3** If  $\mathcal{S}$  is a closed curve with positive reach,  $W \subset \mathbb{R}^d$  s.t.  $d_H(W, \mathcal{S}) \leq \delta$ ,  $L \subseteq W$   $\varepsilon$ -sparse  $\varepsilon$ -sample of  $W$  with  $\delta \ll \varepsilon \ll \varrho_{\mathcal{S}}$ , then  $C^W(L) = \text{Del}_{\mathcal{S}}(L) \approx \mathcal{S}$ .

- $\text{Del}_{\mathcal{S}}(L) \subseteq C^W(L)$

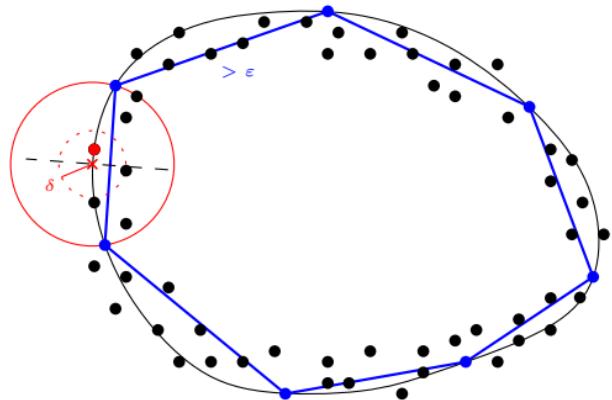


# Theoretical guarantees

→ case of curves:

**Thm. 3** If  $\mathcal{S}$  is a closed curve with positive reach,  $W \subset \mathbb{R}^d$  s.t.  $d_H(W, \mathcal{S}) \leq \delta$ ,  $L \subseteq W$   $\varepsilon$ -sparse  $\varepsilon$ -sample of  $W$  with  $\delta \ll \varepsilon \ll \varrho_{\mathcal{S}}$ , then  $C^W(L) = \text{Del}_{\mathcal{S}}(L) \approx \mathcal{S}$ .

- $\text{Del}_{\mathcal{S}}(L) \subseteq C^W(L)$

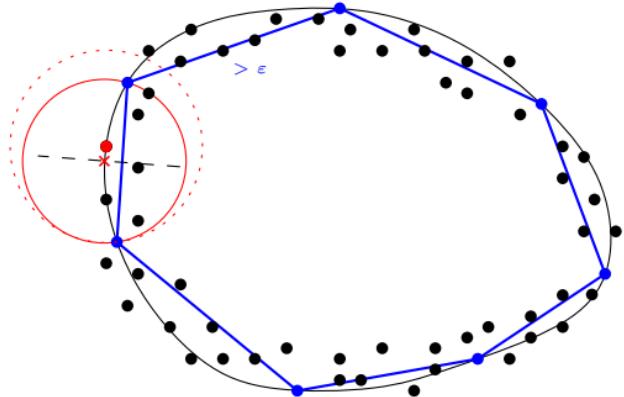


# Theoretical guarantees

→ case of curves:

**Thm. 3** If  $\mathcal{S}$  is a closed curve with positive reach,  $W \subset \mathbb{R}^d$  s.t.  $d_H(W, \mathcal{S}) \leq \delta$ ,  $L \subseteq W$   $\varepsilon$ -sparse  $\varepsilon$ -sample of  $W$  with  $\delta \ll \varepsilon \ll \varrho_{\mathcal{S}}$ , then  $C^W(L) = \text{Del}_{\mathcal{S}}(L) \approx \mathcal{S}$ .

- $\text{Del}_{\mathcal{S}}(L) \subseteq C^W(L)$

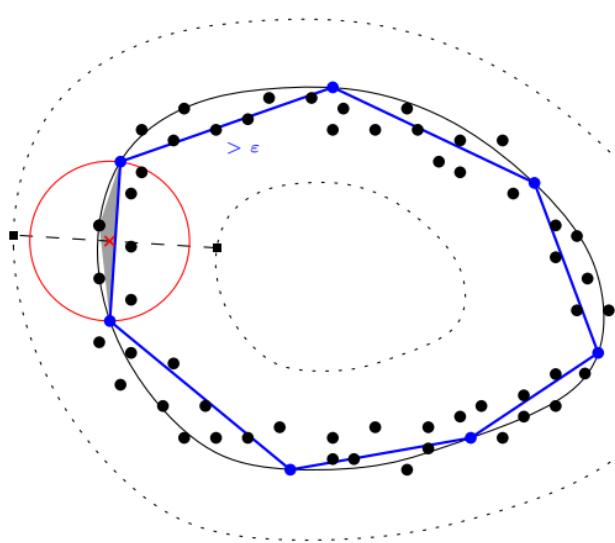


# Theoretical guarantees

→ case of curves:

**Thm. 3** If  $\mathcal{S}$  is a closed curve with positive reach,  $W \subset \mathbb{R}^d$  s.t.  $d_H(W, \mathcal{S}) \leq \delta$ ,  $L \subseteq W$   $\varepsilon$ -sparse  $\varepsilon$ -sample of  $W$  with  $\delta \ll \varepsilon \ll \varrho_{\mathcal{S}}$ , then  $C^W(L) = \text{Del}_{\mathcal{S}}(L) \approx \mathcal{S}$ .

- $\text{Del}_{\mathcal{S}}(L) \subseteq C^W(L)$
- $C^W(L) \subseteq \text{Del}_{\mathcal{S}}(L)$



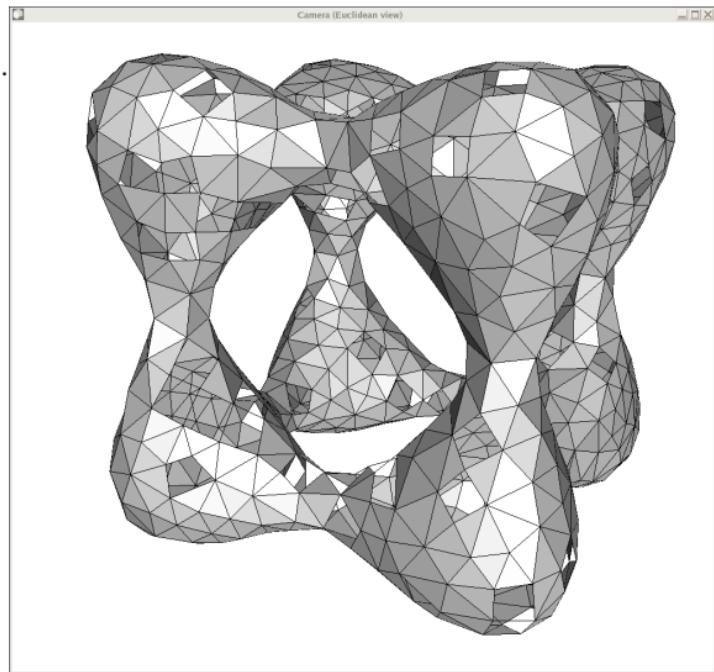
# Theoretical guarantees

→ case of surfaces:

**Thm** [Attali, Edelsbrunner, Mileyko]

If  $\varepsilon \ll \varrho_S$ , then  $\forall W \subseteq S, C^W(L) \subseteq \text{Dels}(L)$ .

$$\Rightarrow C^S(L) = \text{Dels}(L)$$



$$\varepsilon = 0.2, \varrho_S \approx 0.25$$

# Theoretical guarantees

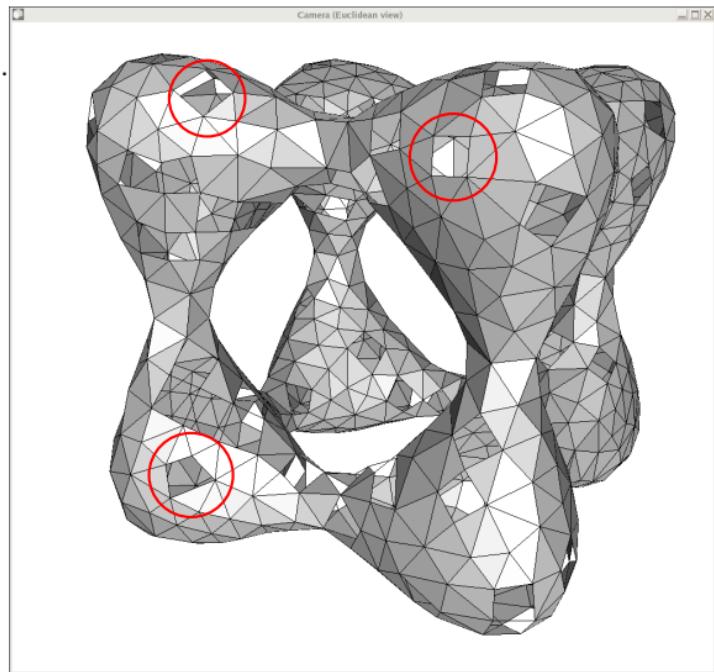
→ case of surfaces:

**Thm** [Attali, Edelsbrunner, Mileyko]

If  $\varepsilon \ll \varrho_S$ , then  $\forall W \subseteq S, C^W(L) \subseteq \text{Dels}(L)$ .

$$\Rightarrow C^S(L) = \text{Dels}(L)$$

**Pb**  $\text{Dels}(L) \not\subseteq C^W(L)$  if  $W \subsetneq S$



$$\varepsilon = 0.2, \varrho_S \approx 0.25$$

# Theoretical guarantees

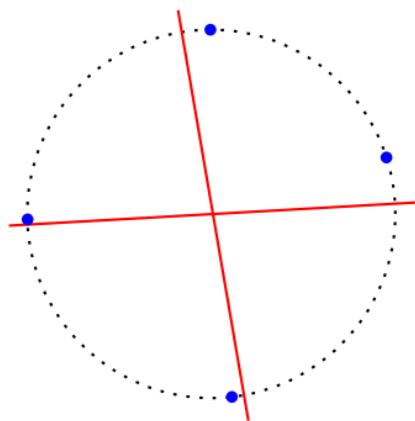
→ case of surfaces:

**Thm** [Attali, Edelsbrunner, Mileyko]

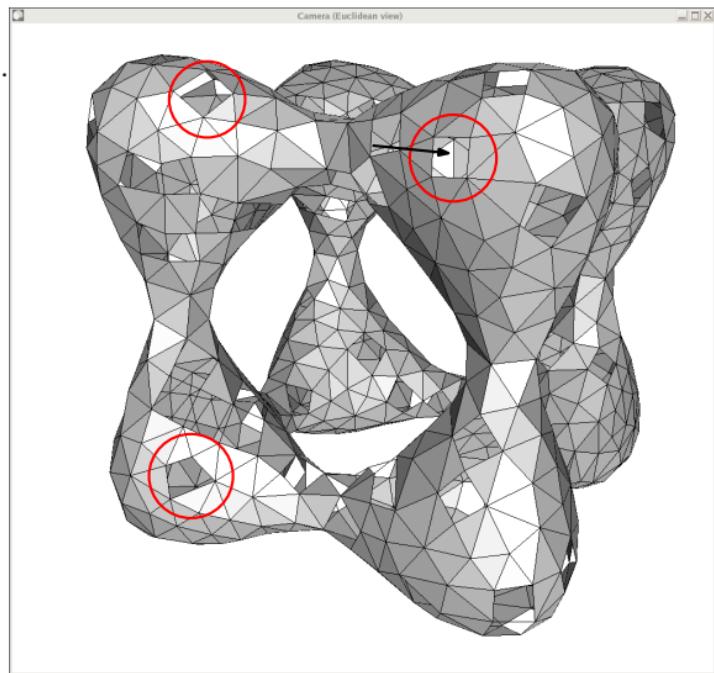
If  $\varepsilon \ll \varrho_S$ , then  $\forall W \subseteq S, C^W(L) \subseteq \text{Dels}(L)$ .

$$\Rightarrow C^S(L) = \text{Dels}(L)$$

**Pb**  $\text{Dels}(L) \not\subseteq C^W(L)$  if  $W \subsetneq S$



order-2 Voronoi diagram



$$\varepsilon = 0.2, \varrho_S \approx 0.25$$

# Theoretical guarantees

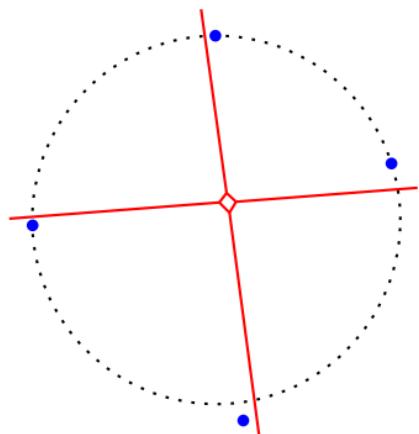
→ case of surfaces:

**Thm** [Attali, Edelsbrunner, Mileyko]

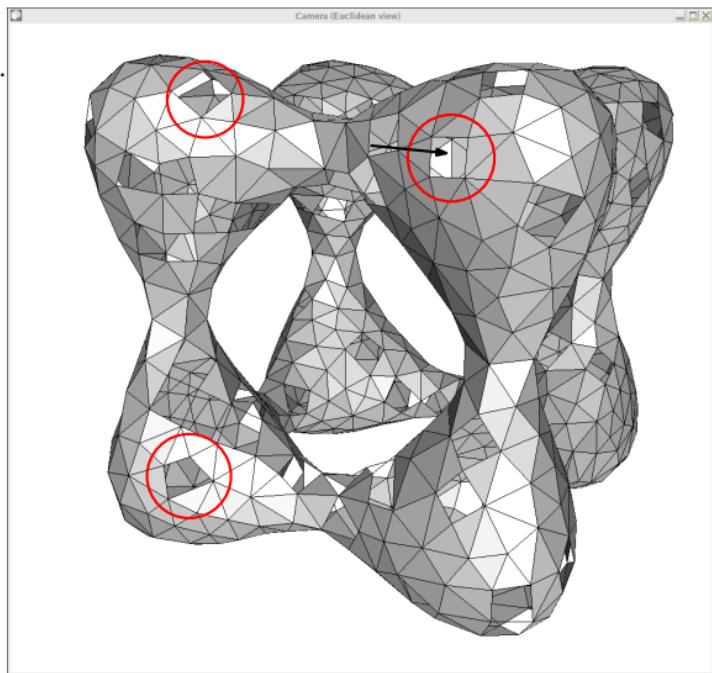
If  $\varepsilon \ll \varrho_S$ , then  $\forall W \subseteq S, C^W(L) \subseteq \text{Dels}(L)$ .

$$\Rightarrow C^S(L) = \text{Dels}(L)$$

**Pb**  $\text{Dels}(L) \not\subseteq C^W(L)$  if  $W \subsetneq S$



order-2 Voronoi diagram



$$\varepsilon = 0.2, \varrho_S \approx 0.25$$

# Theoretical guarantees

→ case of surfaces:

**Thm** [Attali, Edelsbrunner, Mileyko]

If  $\varepsilon \ll \varrho_S$ , then  $\forall W \subseteq S, C^W(L) \subseteq \text{Dels}(L)$ .

$$\Rightarrow C^S(L) = \text{Dels}(L)$$

**Pb**  $\text{Del}_S(L) \not\subseteq C^W(L)$  if  $W \subsetneq S$

**Solution** relax witness test

[Guibas, Oudot]

$$\Rightarrow C_\nu^W(L) = \text{Del}_S(L) + \text{slivers}$$

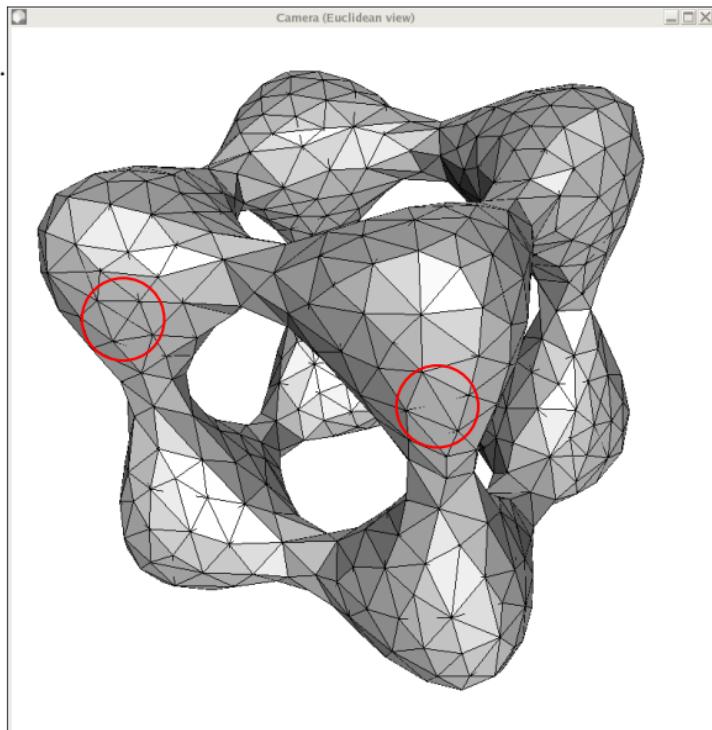
$$\Rightarrow C_\nu^W(L) \not\subseteq \text{Del}(L)$$

$\Rightarrow C_\nu^W(L)$  not embedded.

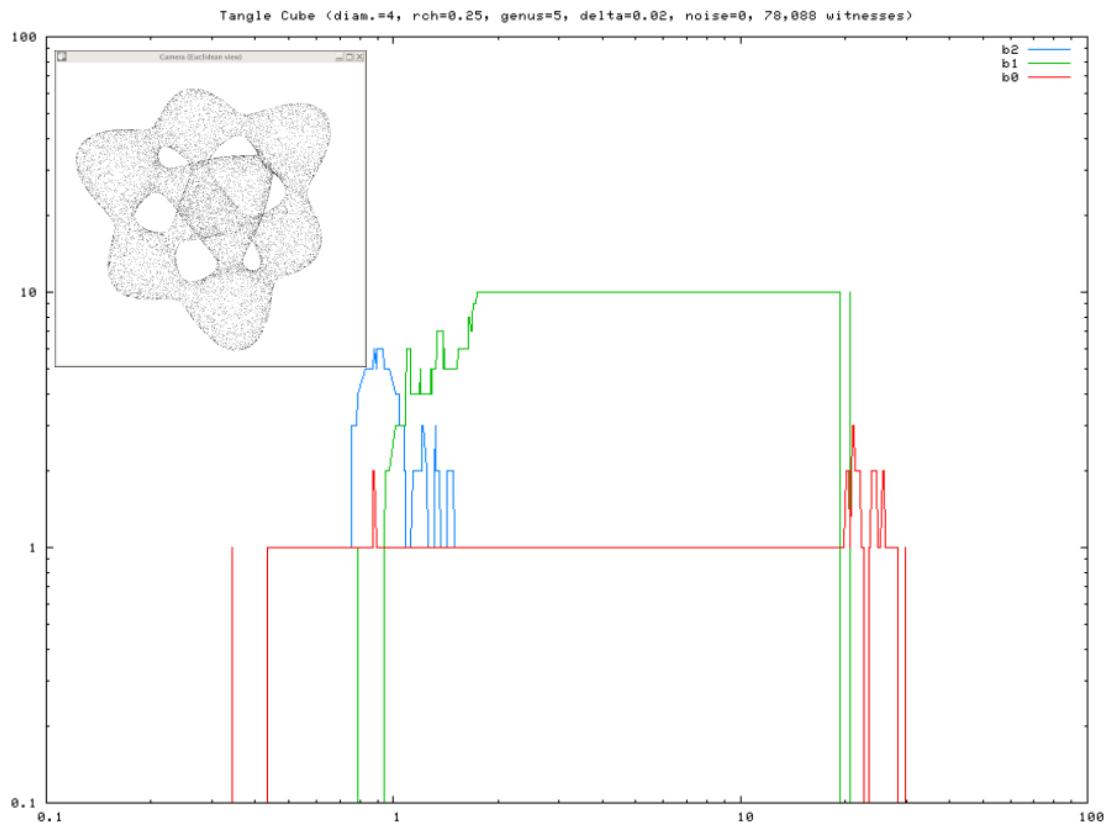
**Post-process** extract manifold  $M$

from  $C_\nu^W(L) \cap \text{Del}(L)$

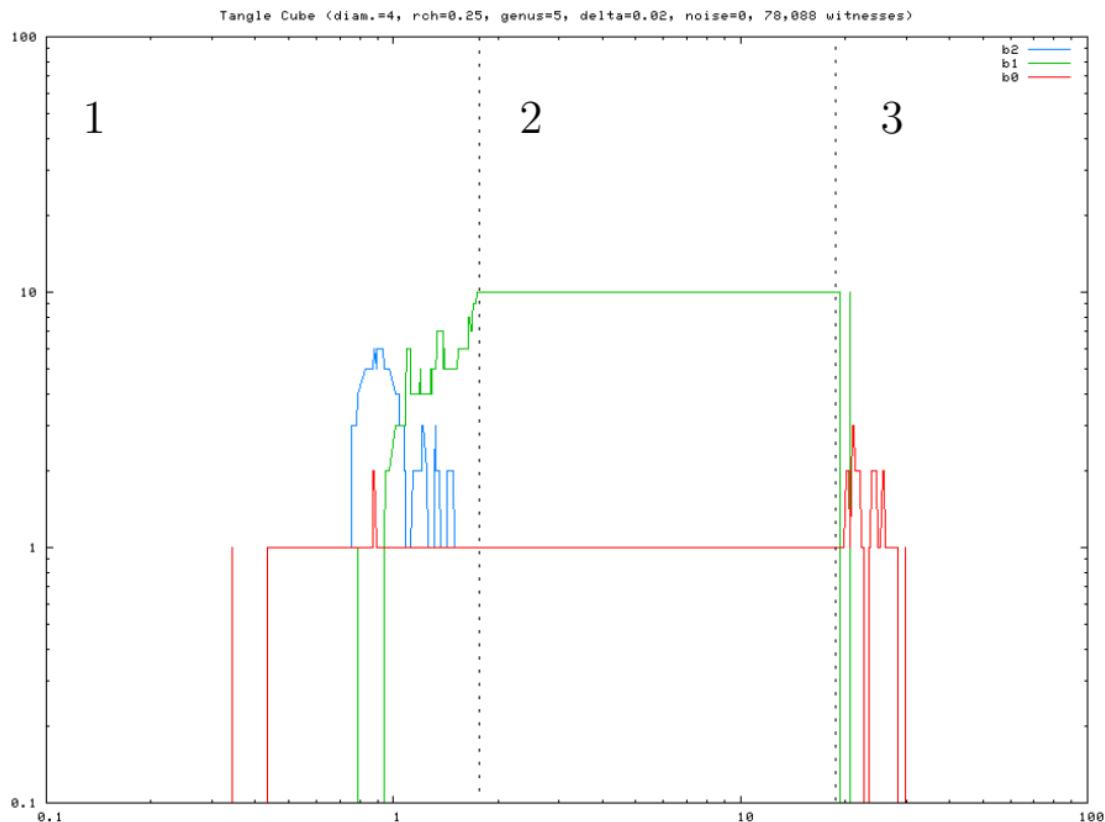
[Amenta, Choi, Dey, Leekha]



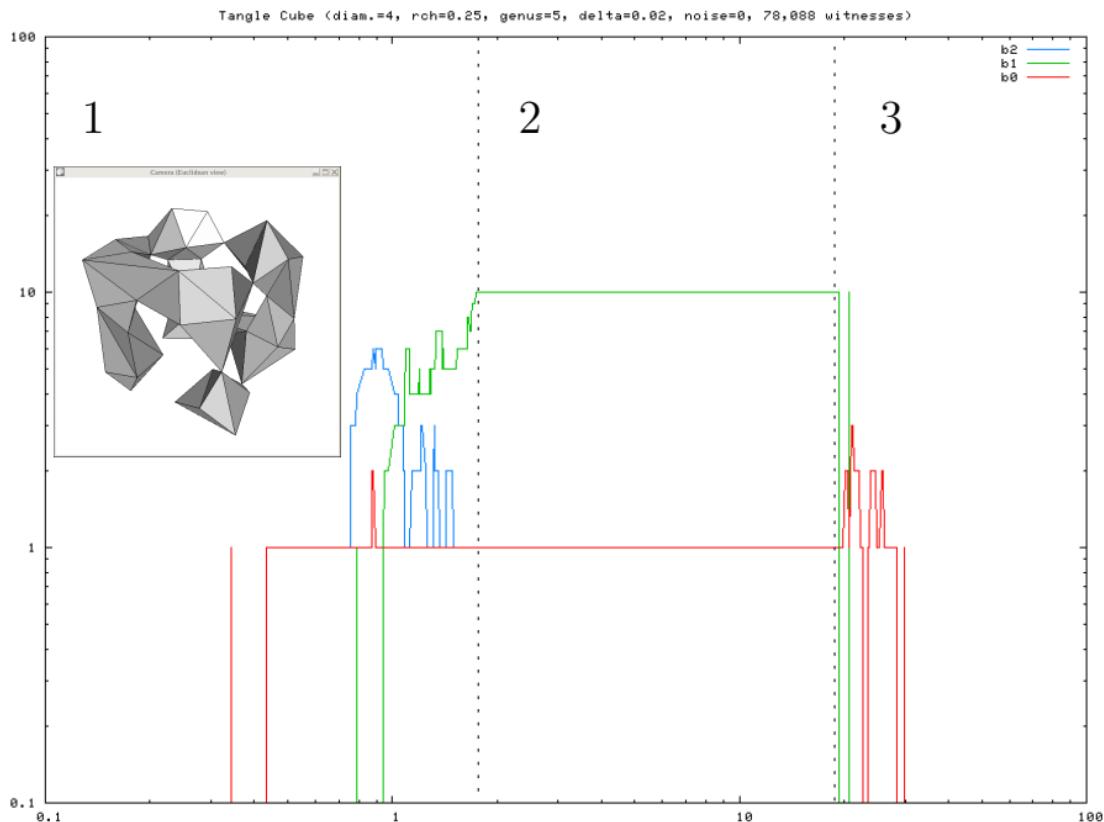
# Some results



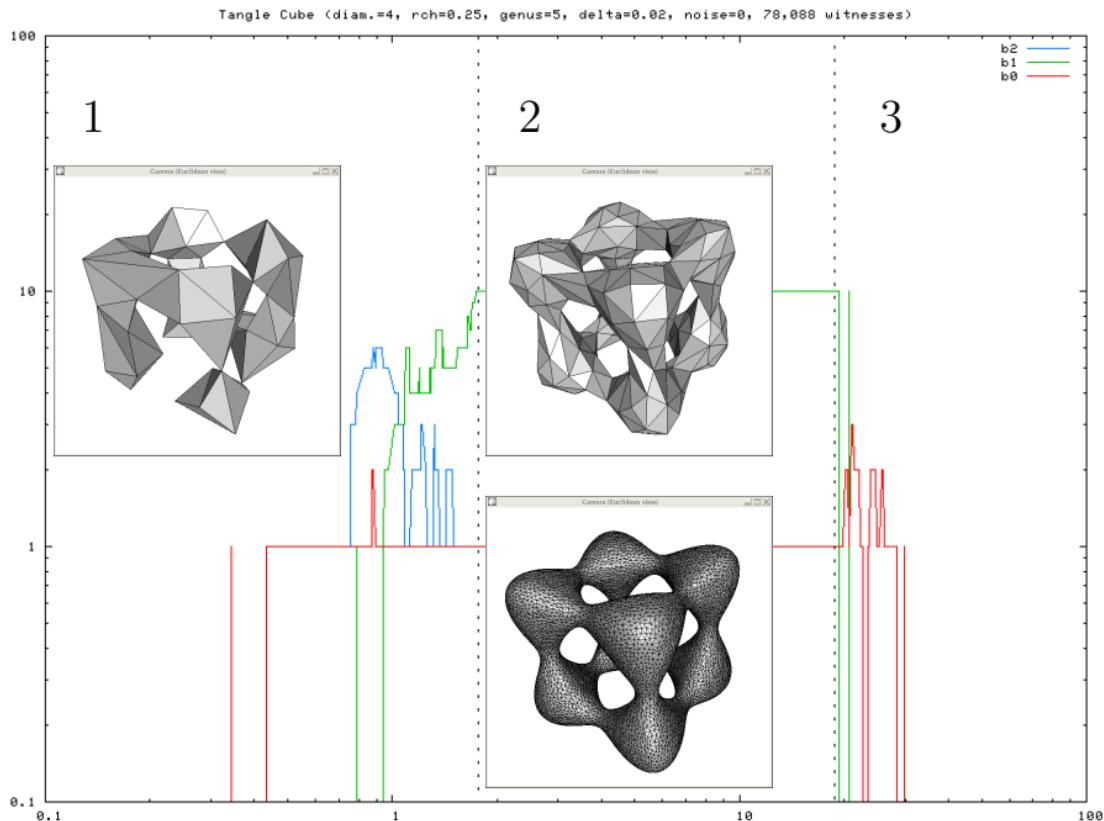
# Some results



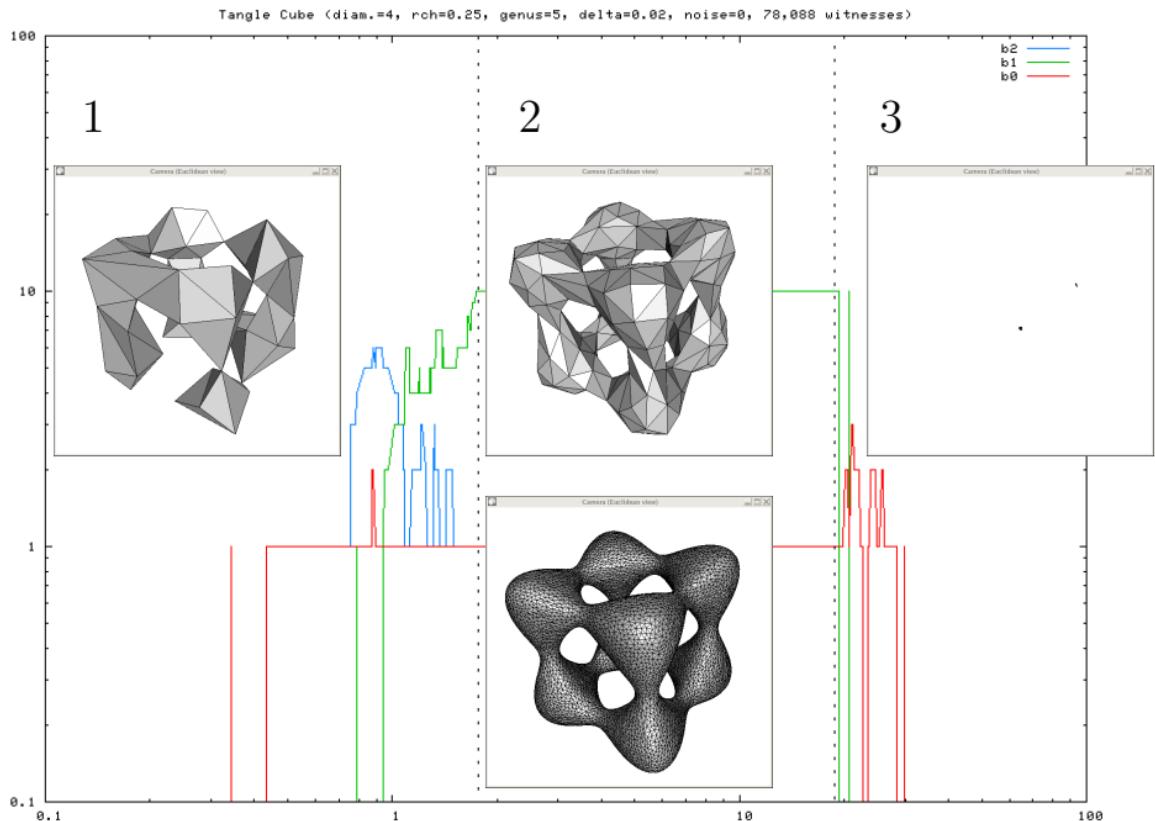
# Some results



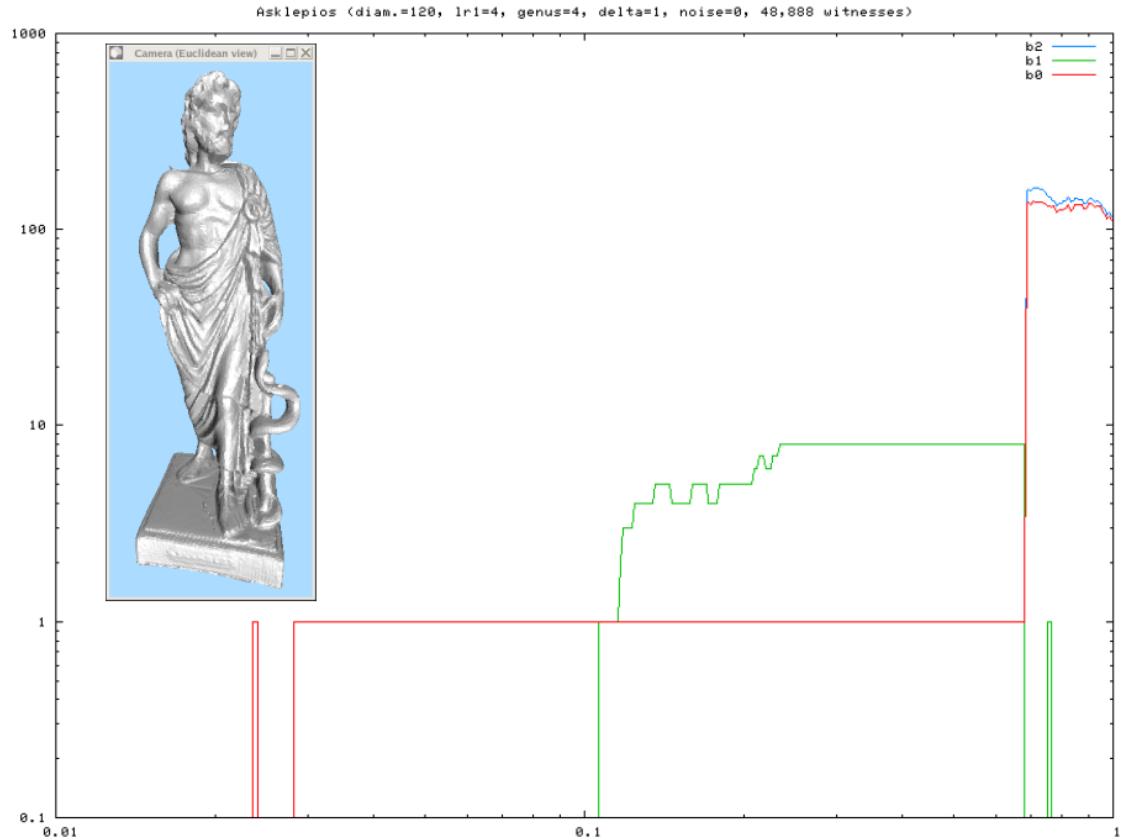
# Some results



# Some results

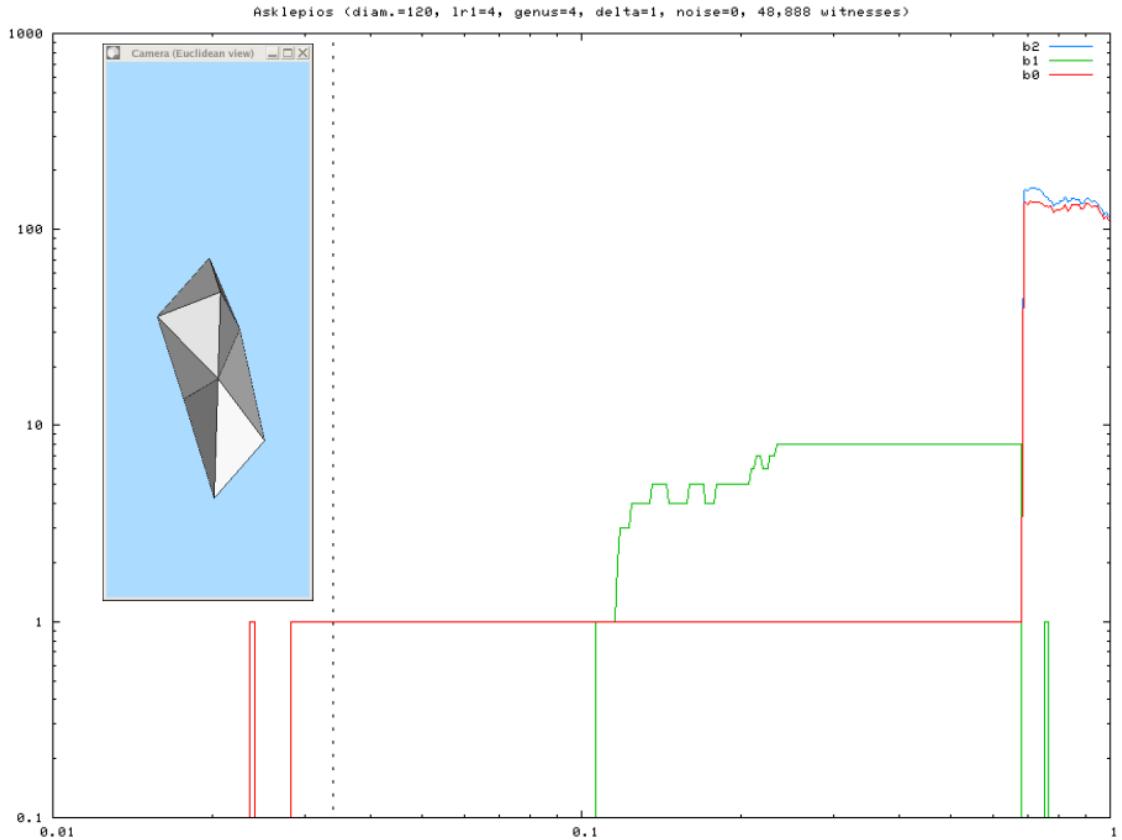


# Some results (cont'd)



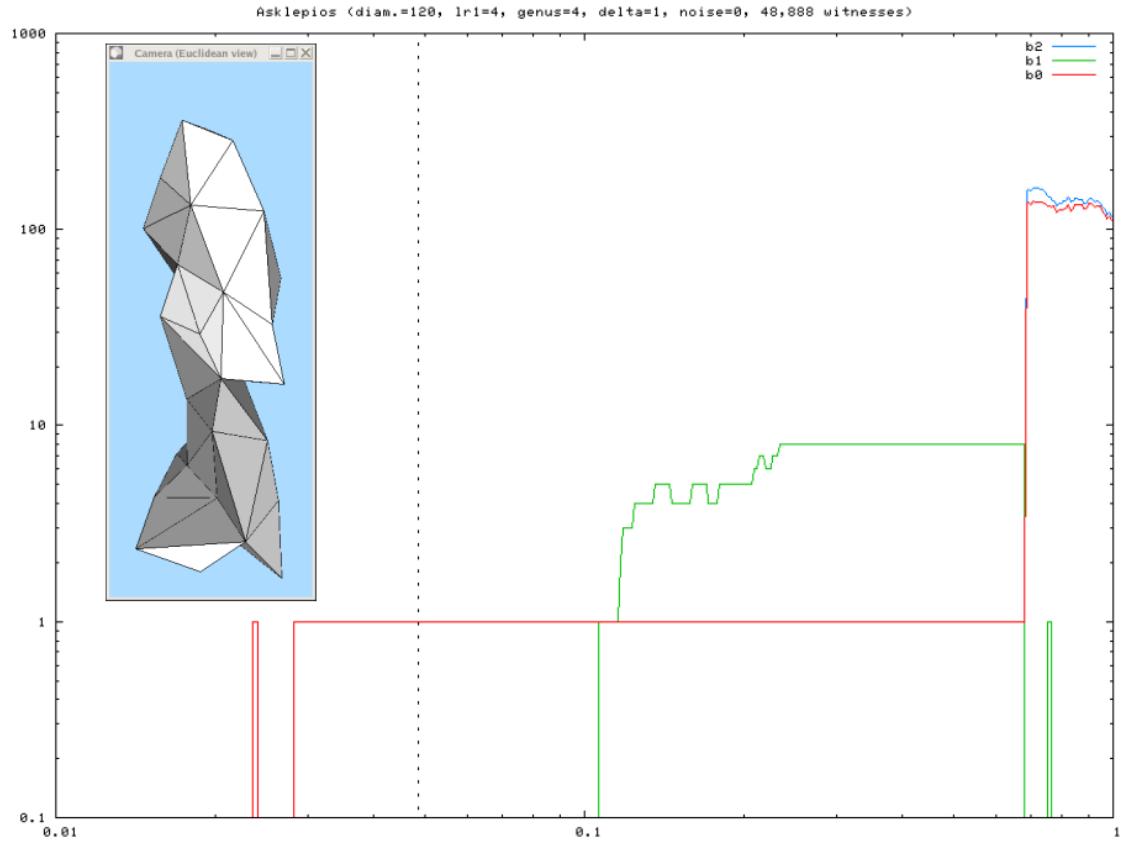
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)



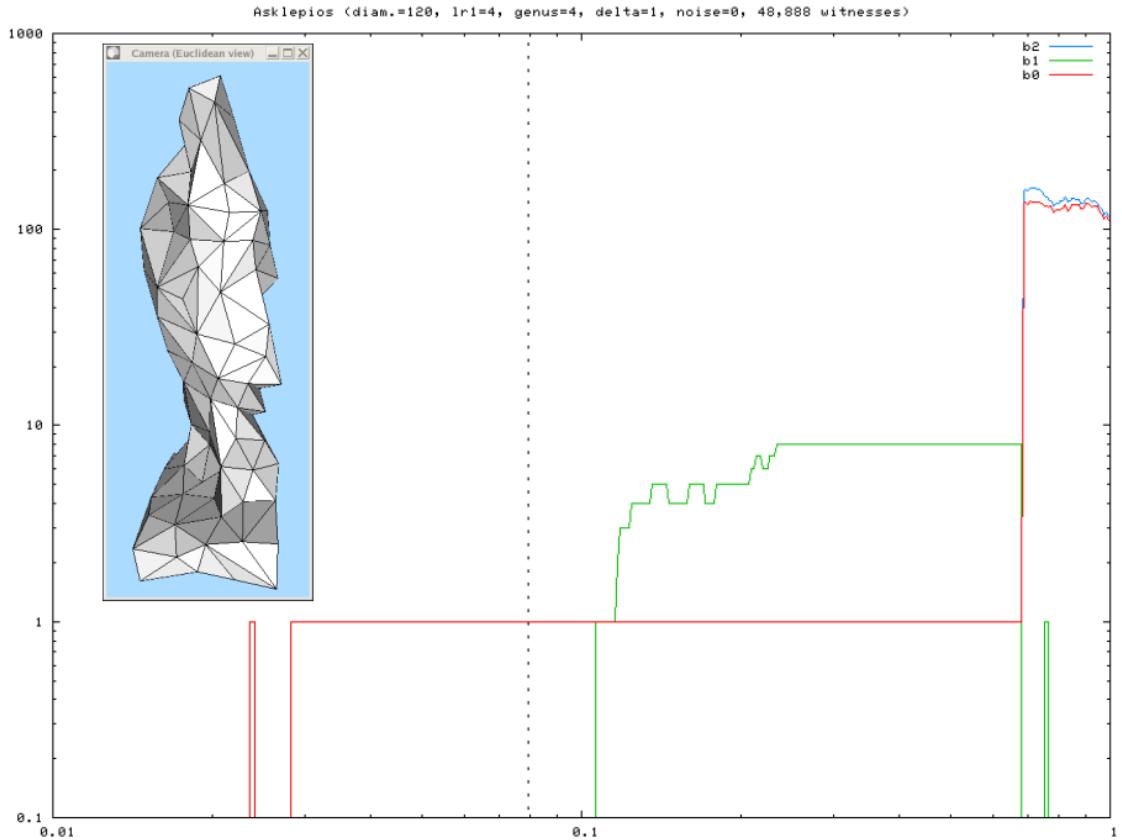
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)



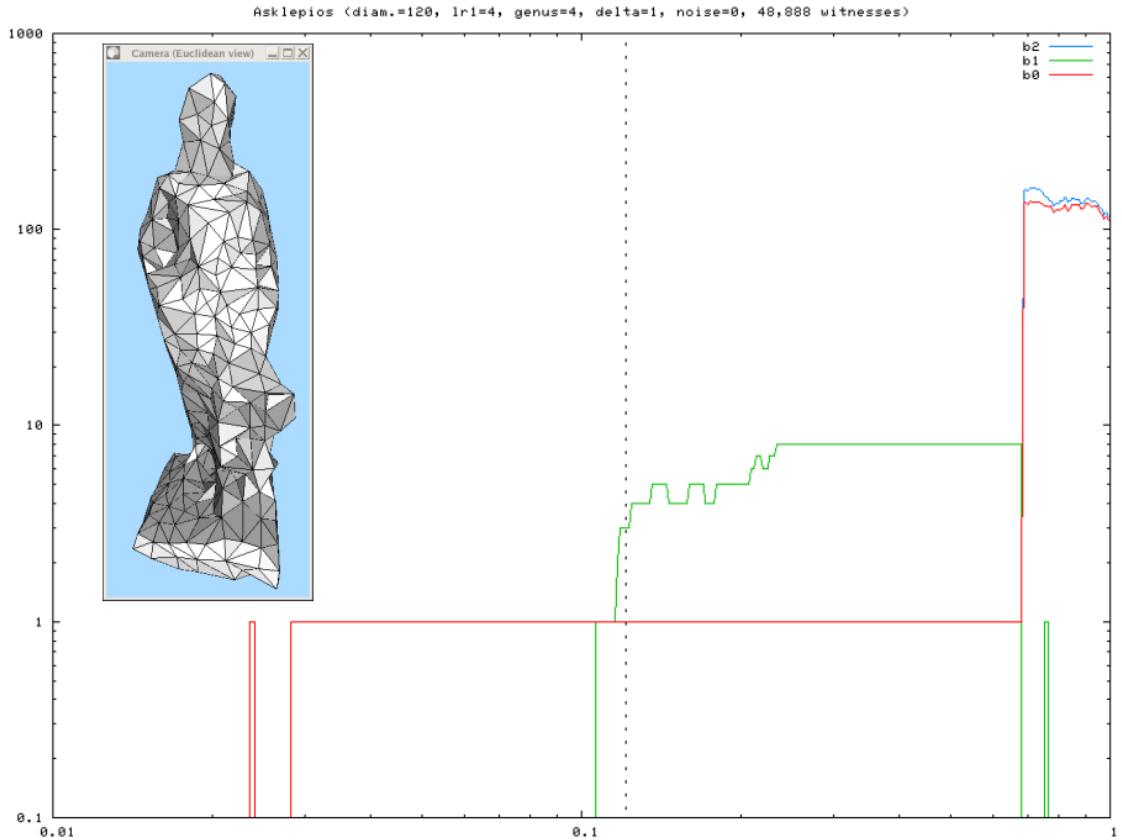
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)

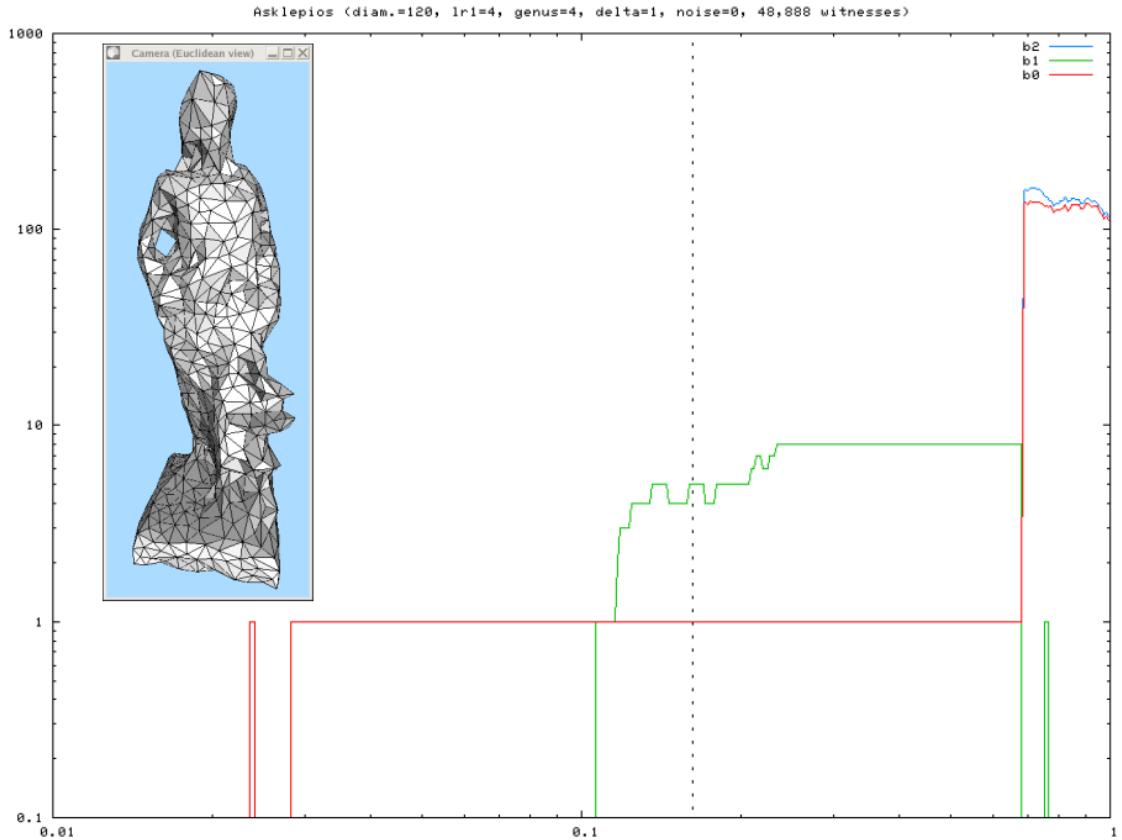


input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)

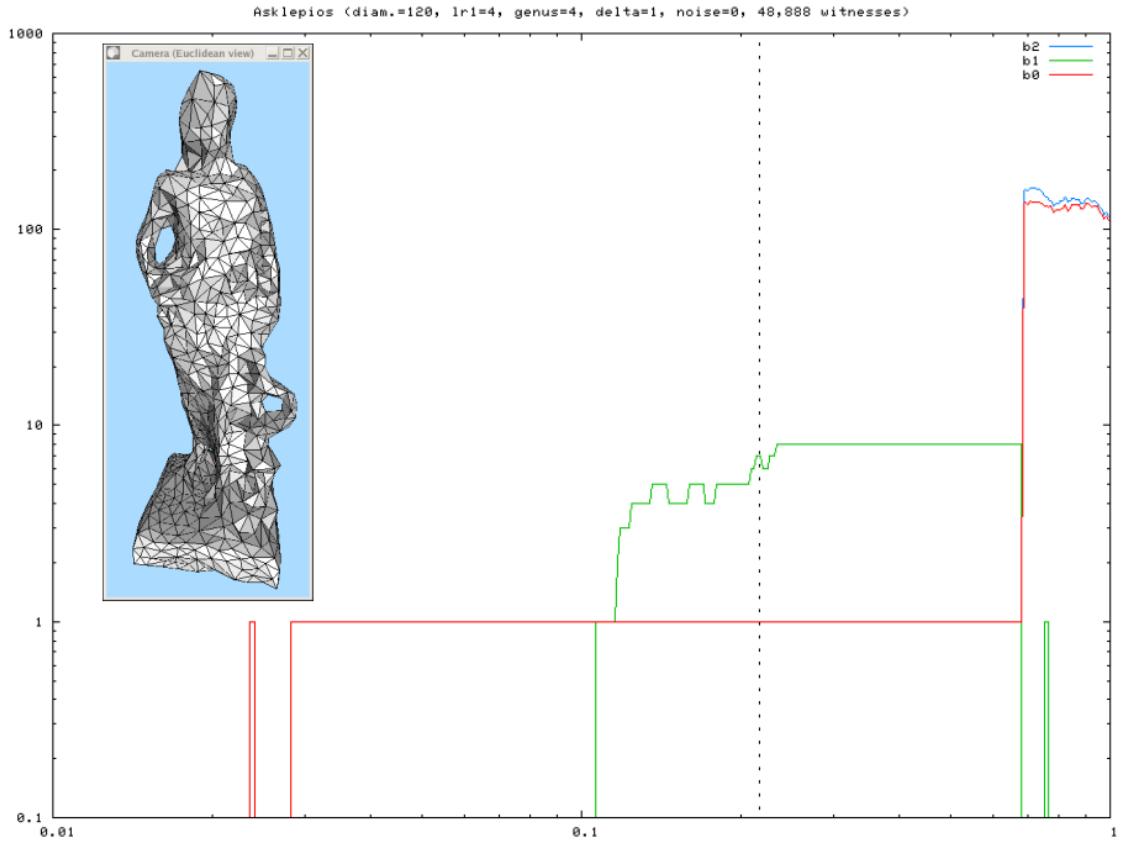


# Some results (cont'd)



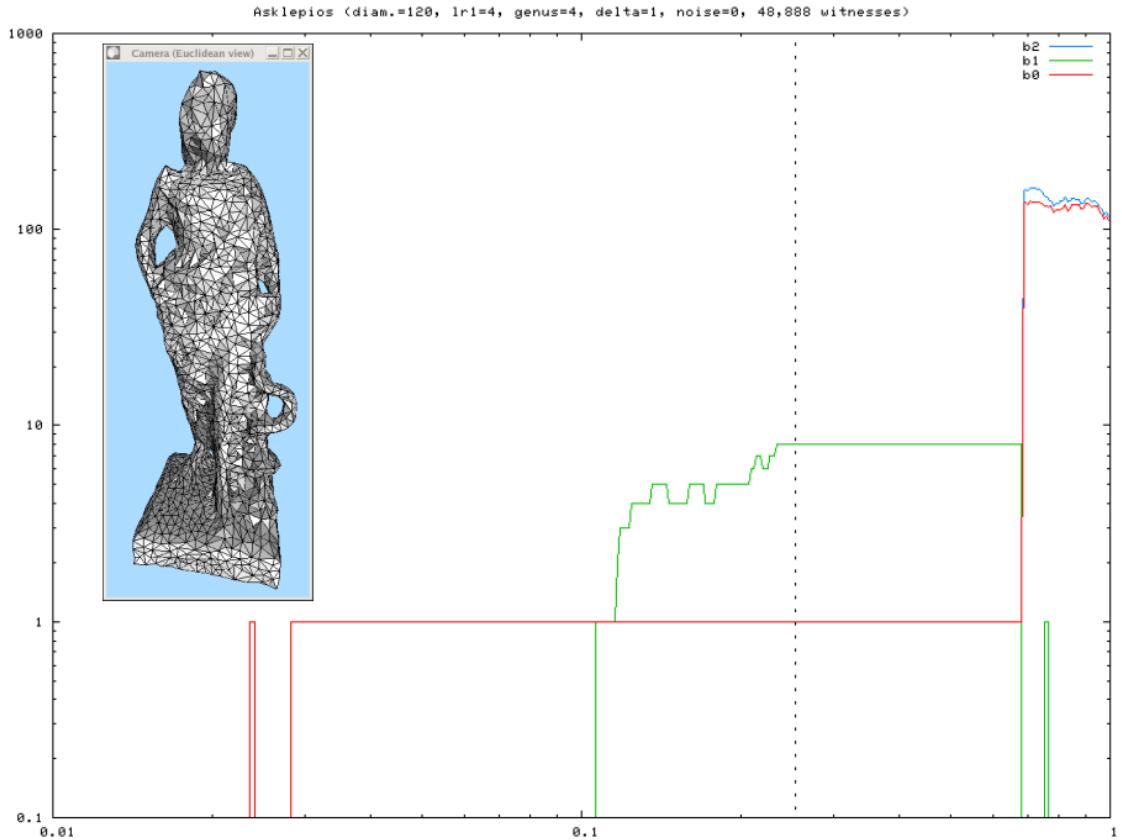
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)



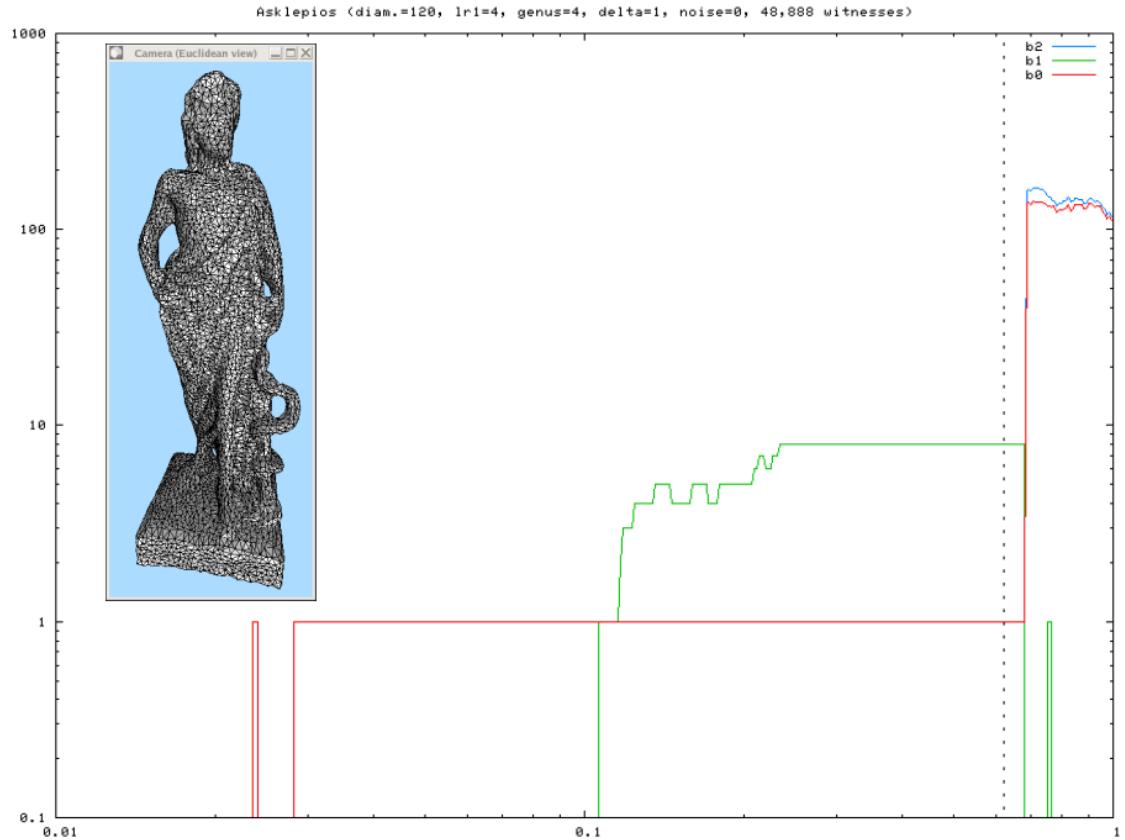
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)



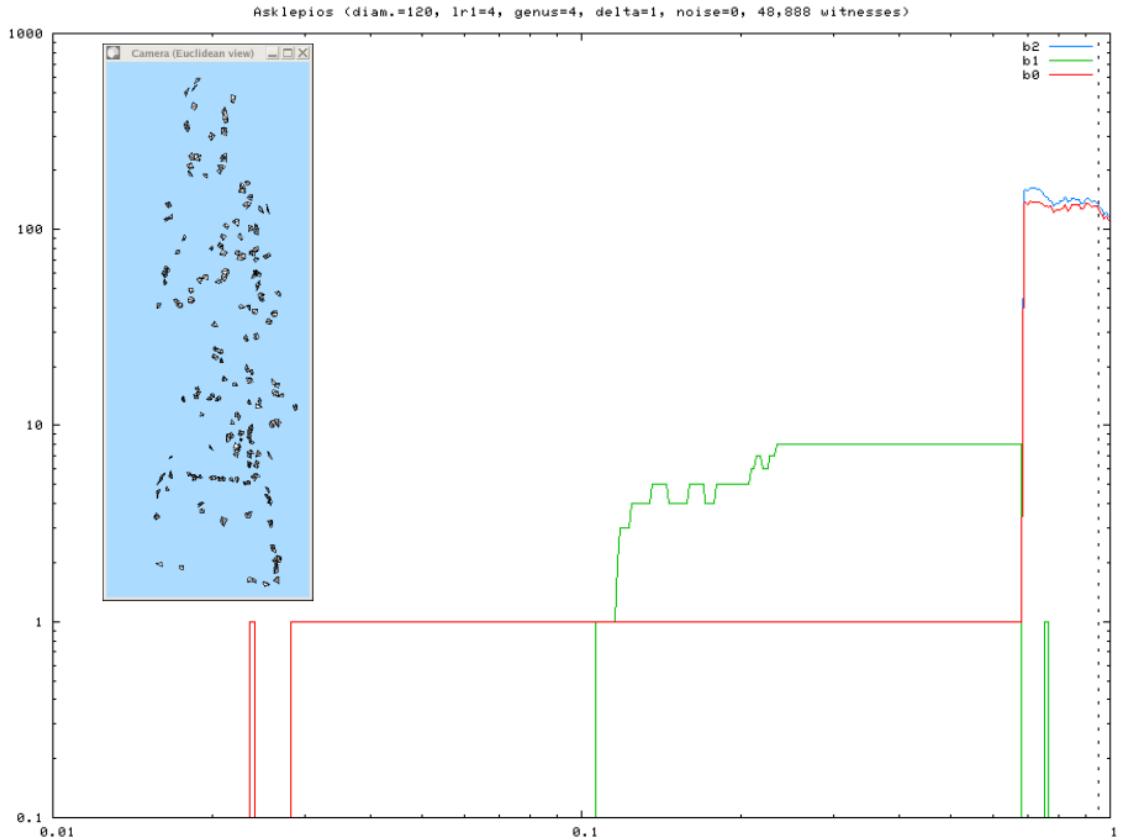
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)



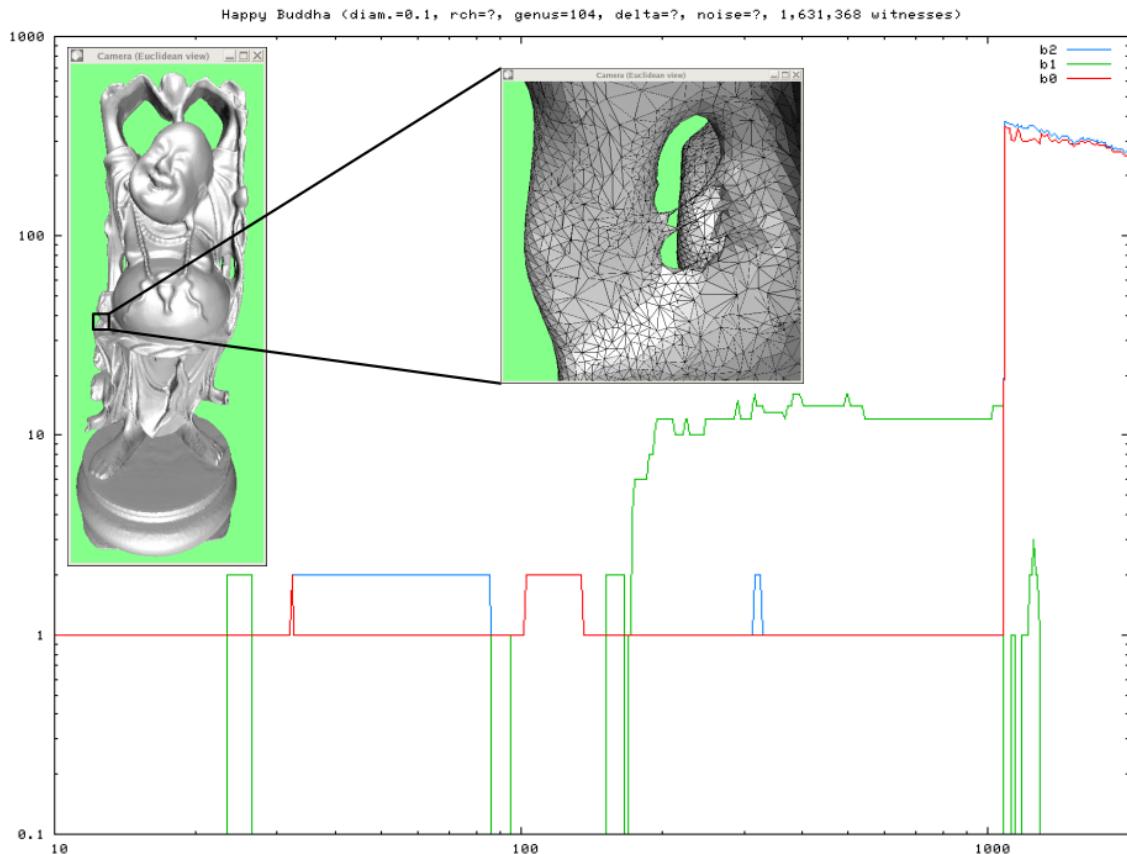
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)



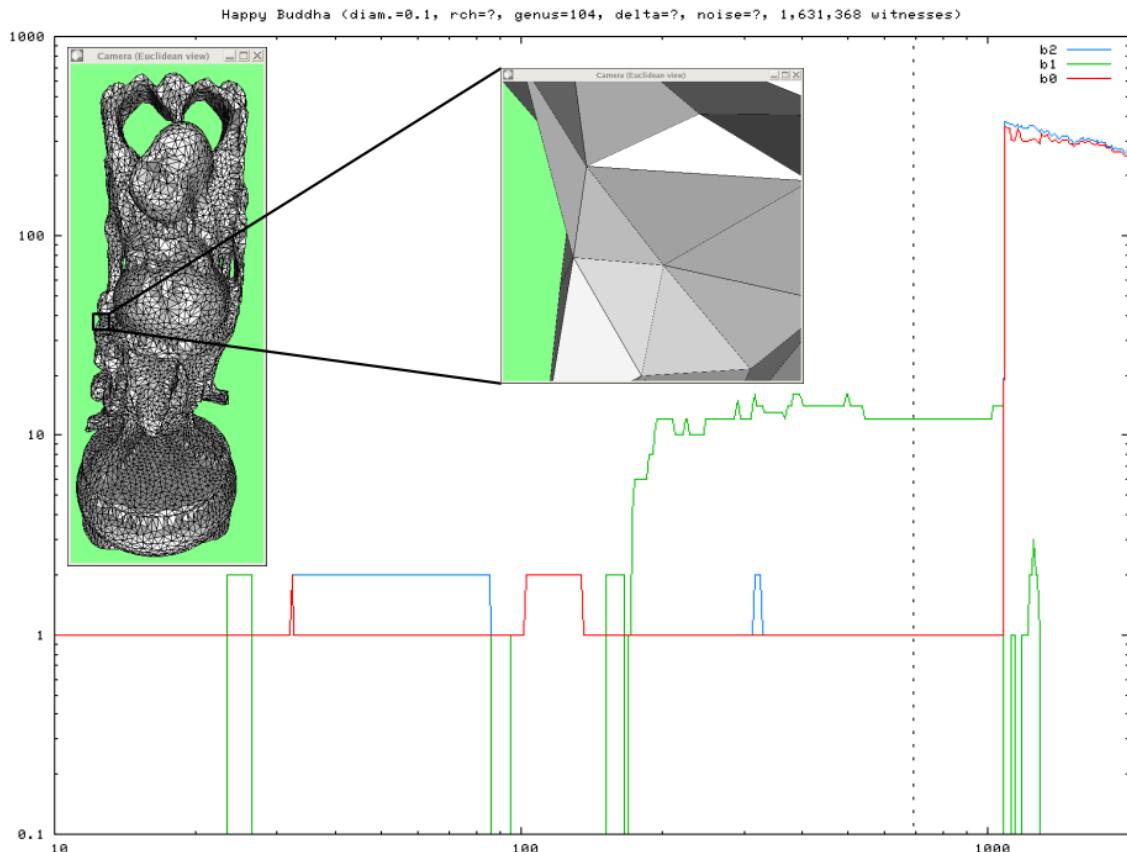
input model provided courtesy of IMATI by the Aim@Shape repository

# Some results (cont'd)

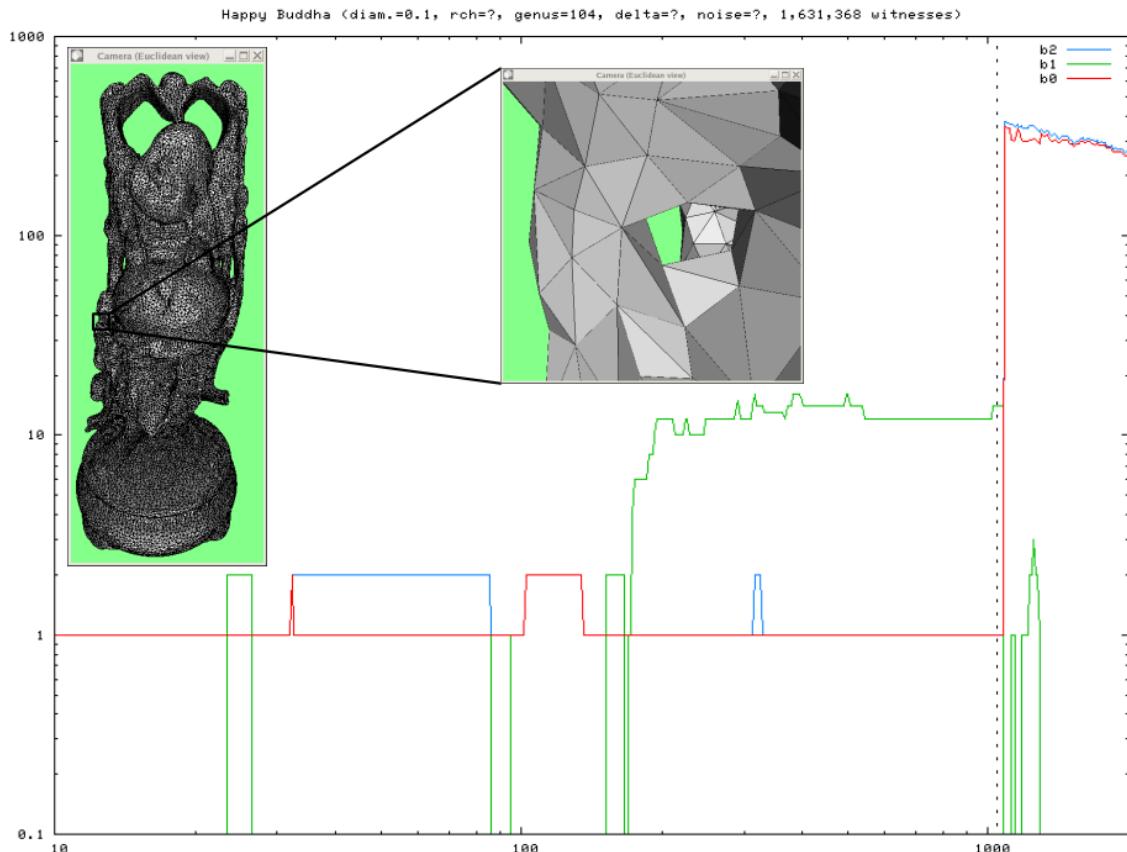


input data set courtesy of the Graphics Lab@Stanford

# Some results (cont'd)

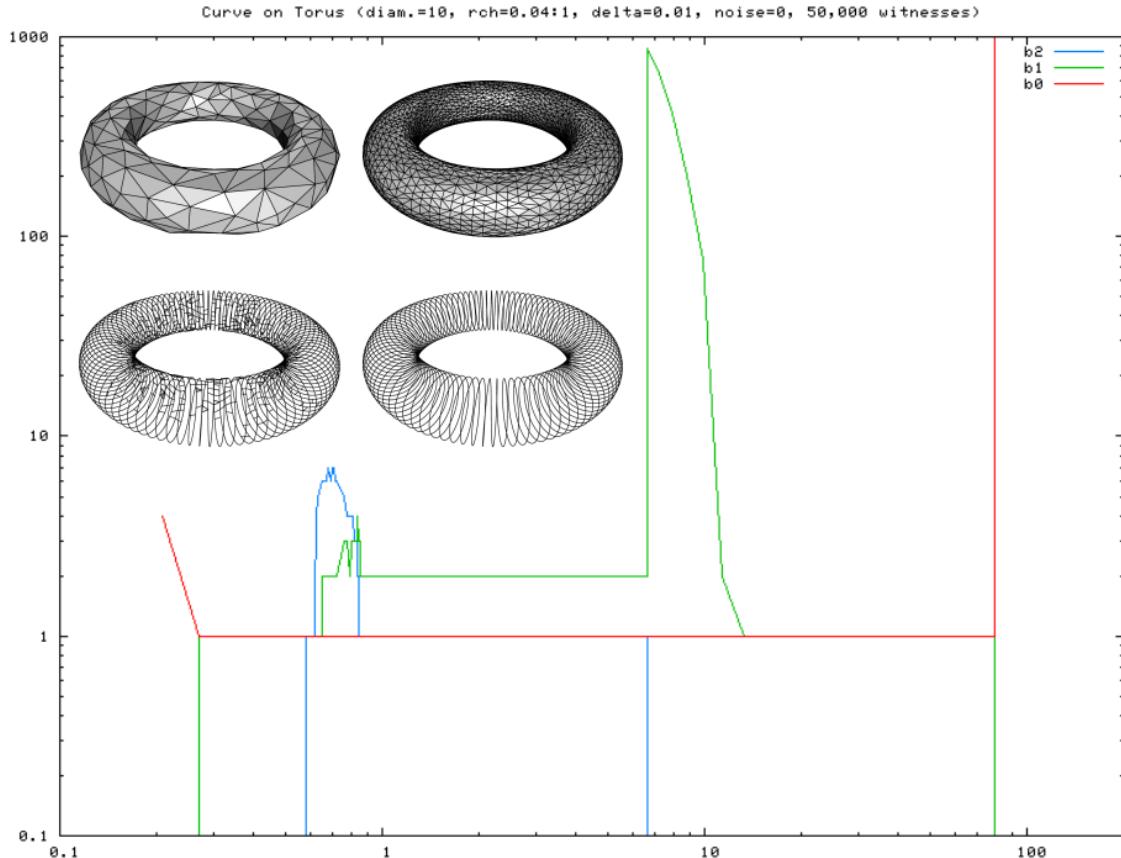


## Some results (cont'd)



input data set courtesy of the Graphics Lab@Stanford

## Some results (cont'd)



## Higher dimensions

→ Carlsson and de Silva's conjecture:

*Under some sampling conditions,  $C^W(L) = \text{Del}_{\mathcal{S}}(L) \approx \mathcal{S}$*

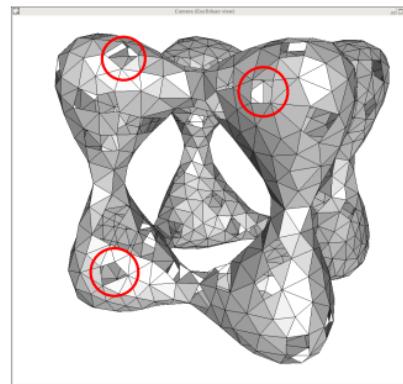
# Higher dimensions

→ Carlsson and de Silva's conjecture:

~~Under some sampling conditions,  $C^W(L) = \text{Del}_S(L) \approx \mathcal{S}$~~

non longer true

- $\text{Del}_S(L)$  may not be included in  $C^W(L)$  on  $d$ -manifolds,  $d \geq 2$  [Guibas, Oudot]



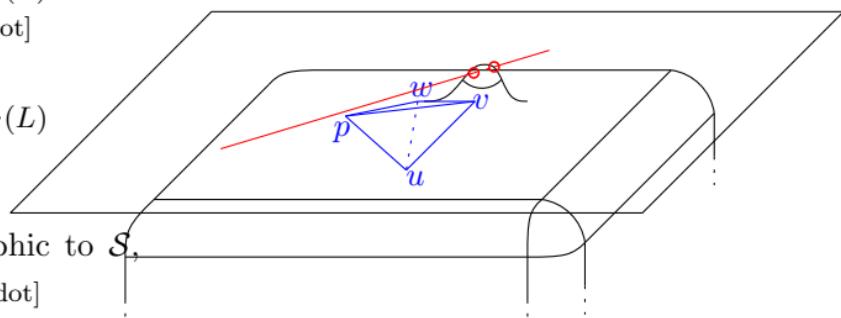
# Higher dimensions

→ Carlsson and de Silva's conjecture:

~~Under some sampling conditions,  $C^W(L) = \text{Del}_S(L) \approx S$~~

non longer true

- $\text{Del}_S(L)$  may not be included in  $C^W(L)$  on  $d$ -manifolds,  $d \geq 2$  [Guibas, Oudot]



- $C^W(L)$  may not be included in  $\text{Del}_S(L)$  on  $d$ -manifolds,  $d \geq 3$  [Oudot]

- $\text{Del}_S(L)$  may not be homeomorphic to  $S$ , nor even homotopy equivalent [Oudot]

→ source of problems: **slivers**

# Higher dimensions

→ Carlsson and de Silva's conjecture:

~~Under some sampling conditions,  $C^W(L) = \text{Del}_S(L) \approx S$~~

non longer true

- $\text{Del}_S(L)$  may not be included in  $C^W(L)$  on  $d$ -manifolds,  $d \geq 2$  [Guibas, Oudot]

) dilate  $W$  so that it includes  $S$   
[Boissonnat, Guibas, Oudot]

- $C^W(L)$  may not be included in  $\text{Del}_S(L)$  on  $d$ -manifolds,  $d \geq 3$  [Oudot]

) assign weights to the landmarks  
to remove slivers

- $\text{Del}_S(L)$  may not be homeomorphic to  $S$ , nor even homotopy equivalent [Oudot]

[Cheng, Dey, Ramos]

→ source of problems: **slivers**

# Higher dimensions

→ Carlsson and de Silva's conjecture:

~~Under some sampling conditions,  $C^W(L) = \text{Del}_S(L) \approx S$~~

non longer true

- $\text{Del}_S(L)$  may not be included in  $C^W(L)$  on  $d$ -manifolds,  $d \geq 2$  [Guibas, Oudot]

) dilate  $W$  so that it includes  $S$   
[Boissonnat, Guibas, Oudot]

- $C^W(L)$  may not be included in  $\text{Del}_S(L)$  on  $d$ -manifolds,  $d \geq 3$  [Oudot]

) assign weights to the landmarks  
to remove slivers

- $\text{Del}_S(L)$  may not be homeomorphic to  $S$ , nor even homotopy equivalent [Oudot]

[Cheng, Dey, Ramos]

→ source of problems: **slivers**

Higher-dimensional reconstruction is still widely open