

# Computational Topology in Reconstruction, Mesh Generation, and Data Analysis

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# Outline

- Topological concepts:

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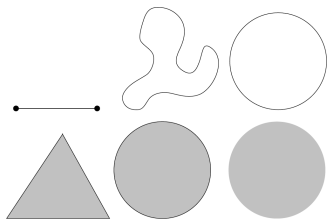
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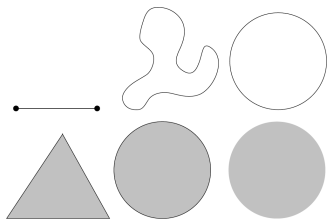
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- A point set with **open** subsets closed under **union** and **finite intersections**



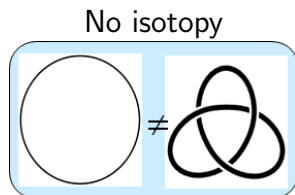
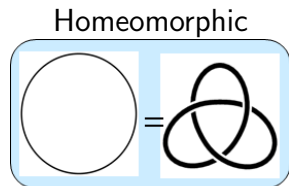
# Topological spaces

- A point set with **open** subsets closed under **union** and **finite intersections**
- $d$ -ball  $B^d = \{x \in \mathbb{R}^d \mid \|x\| \leq 1\}$
- $d$ -sphere  $S^d = \{x \in \mathbb{R}^d \mid \|x\| = 1\}$
- $k$ -manifold: neighborhoods 'homeomorphic' to open  $k$ -ball
  - 2-sphere, torus, double torus are 2-manifolds



# Maps

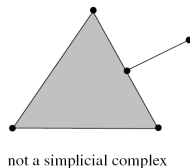
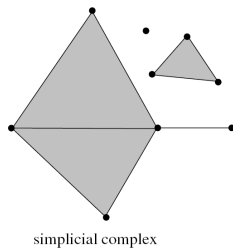
- **Homeomorphism**  $h : T_1 \rightarrow T_2$   
where  $h$  is continuous, bijective  
and has continuous inverse
- **Isotopy** : continuous deformation that  
maintains homeomorphism
- **homotopy equivalence**: map linked to  
continuous deformation only



# Simplicial complex

- **Abstract**

- $V(K)$ : vertex set,  $k$ -simplex:  
 $(k + 1)$ -subset  $\sigma \subseteq V(K)$
- Complex  
 $K = \{\sigma \mid \sigma' \subseteq \sigma \implies \sigma' \in K\}$



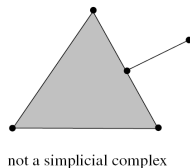
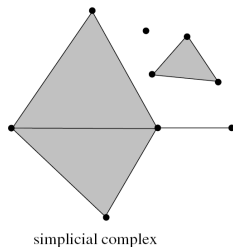
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- **Geometric**

- $k$ -simplex:  $k + 1$ -point convex hull
- Complex  $K$ :
  - $t \in K$  if  $t$  is a face of  $t' \in K$
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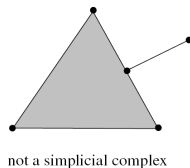
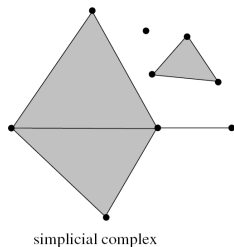
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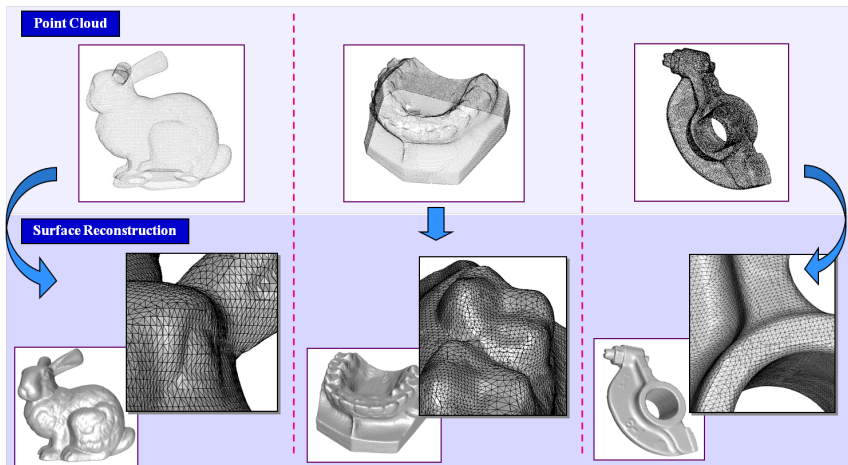
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- **Triangulation:**  $K$  is a triangulation of a topological space  $T$  if  $T \approx |K|$



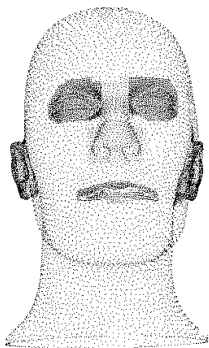


# Surface Reconstruction

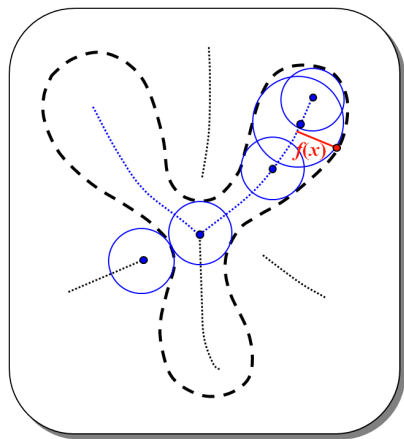


# Sampling

- Sample  $P \subset \Sigma \subset \mathbb{R}^3$

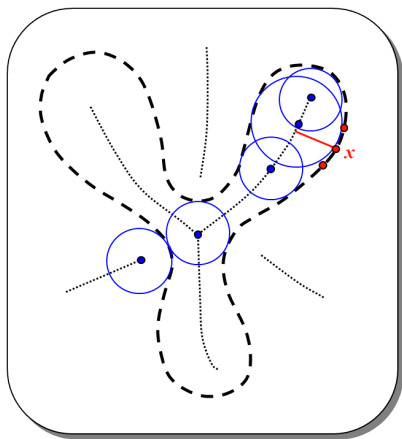


# Local Feature Size



- $Lfs(x)$  is the distance to medial axis

# $\varepsilon$ -sample (Amenta-Bern-Eppstein 98)



- Each  $x$  has a sample within  $\varepsilon \text{Lfs}(x)$  distance

# Crust and Cocone Guarantees

Theorem (Crust: Amenta-Bern 1999)

*Any point  $x \in \Sigma$  is within  $O(\varepsilon)\text{Lfs}(x)$  distance from a point in the output. Conversely, any point of the output surface has a point  $x \in \Sigma$  within  $O(\varepsilon)\text{Lfs}(x)$  distance for  $\varepsilon < 0.06$ .*

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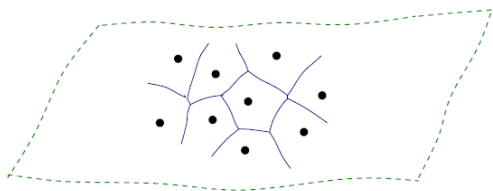
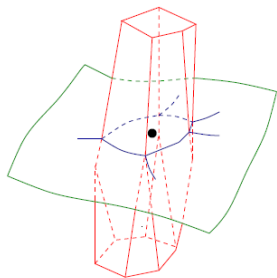
## Theorem (Cocone: Amenta-Choi-Dey-Leekha 2000)

*The output surface computed by COCONE from an  $\varepsilon$  – sample is homeomorphic to the sampled surface for  $\varepsilon < 0.06$ .*

# Restricted Voronoi/Delaunay

## Definition

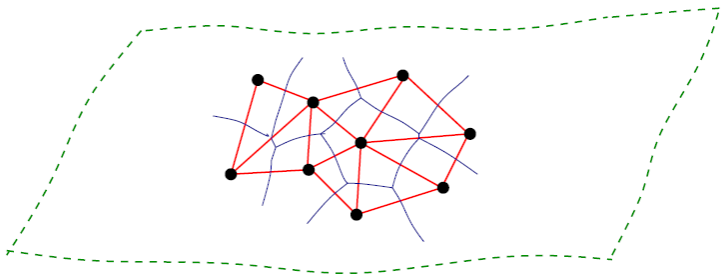
**Restricted Voronoi:**  $\text{Vor } P|_{\Sigma}$ : Intersection of  $\text{Vor}(P)$  with the surface/manifold  $\Sigma$ .



# Restricted Voronoi/Delaunay

## Definition

**Restricted Delaunay:**  $\text{Del } P|_{\Sigma}$ : dual of  $\text{Vor } P|_{\Sigma}$





# Topology

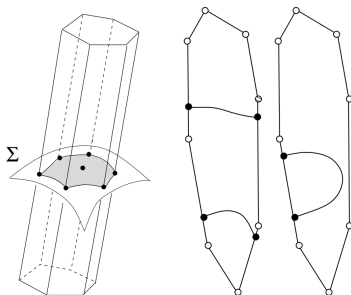
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*If restricted Voronoi cell is a closed ball in each dimension, then  $\text{Del } P|_{\Sigma}$  is homeomorphic to  $\Sigma$ .*

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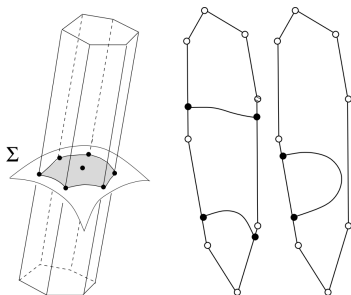
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## Theorem

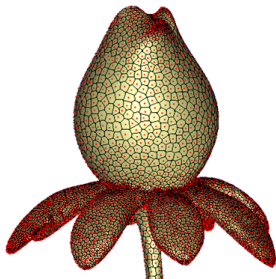
For a sufficiently small  $\varepsilon$  if  $P$  is an  $\varepsilon$ -sample of  $\Sigma$ , then  $(P, \Sigma)$  satisfies the closed ball property, and hence  $\text{Del } P|_{\Sigma} \approx \Sigma$ .



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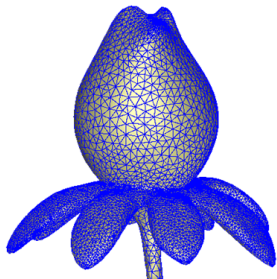
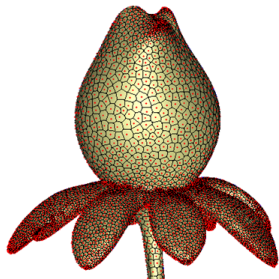
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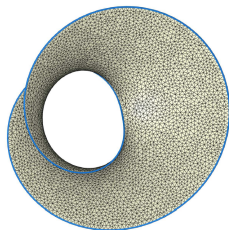
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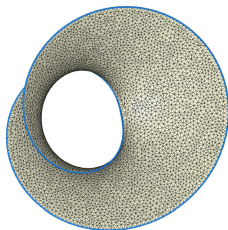


# Boundaries



- **Ambiguity** in reconstruction
- **Non-homeomorphic** Restricted Delaunay [DLRW09]
- **Non-orientability**

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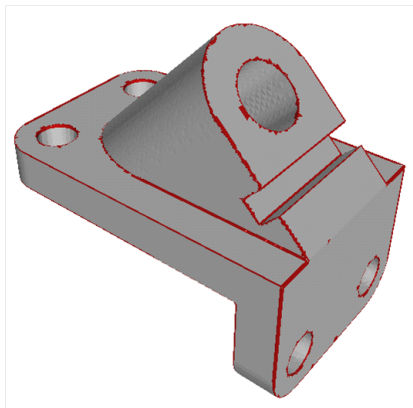
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Theorem (D.-Li-Ramos-Wenger 2009)

Given a sufficiently *dense* sample of a smooth compact surface  $\Sigma$  with boundary one can compute a Delaunay mesh *isotopic* to  $\Sigma$ .

# Open: Reconstructing nonsmooth surfaces

- Guarantee of homeomorphism is **open**





# High Dimensional PCD

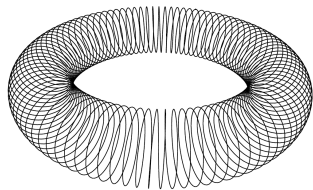
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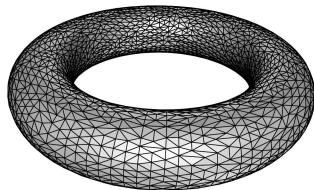
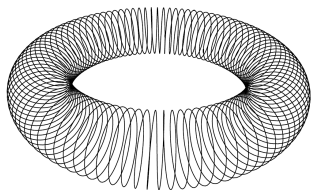
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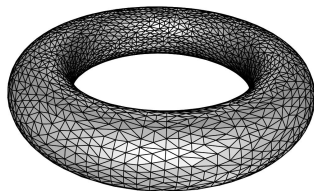
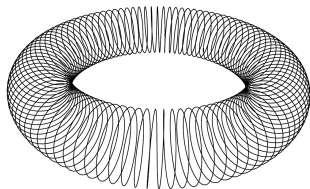
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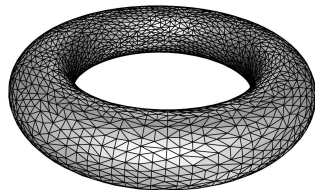
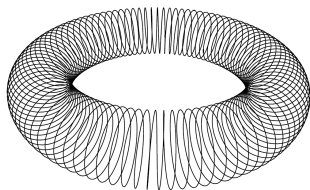
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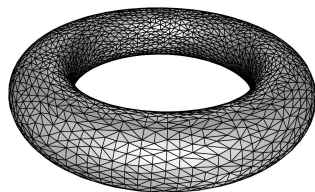
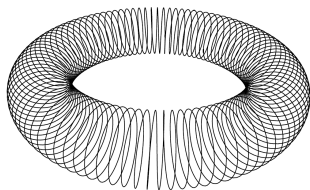
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- Delaunay triangulation becomes **harder**



# Reconstruction

## Theorem (Cheng-Dey-Ramos 2005)

Given an  $(\varepsilon, \delta)$ -sample  $P$  of a smooth manifold  $\Sigma \subset \mathbb{R}^d$  for appropriate  $\varepsilon, \delta > 0$ , there is a *weight assignment* of  $P$  so that  $\text{Del } \hat{P}|_{\Sigma} \approx \Sigma$  which can be computed efficiently.



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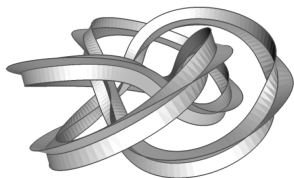
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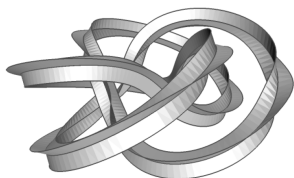
## Theorem (Chazal-Lieutier 2006)

Given an  $\varepsilon$ -noisy sample  $P$  of manifold  $\Sigma \subset \mathbb{R}^d$ , there exists  $r_p \leq \rho(\Sigma)$  for each  $p \in P$  so that the union of balls  $B(p, r_p)$  is *homotopy equivalent* to  $\Sigma$ .

# Reconstructing Compacts

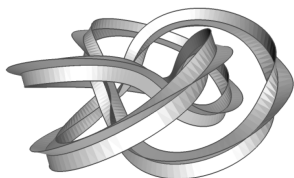


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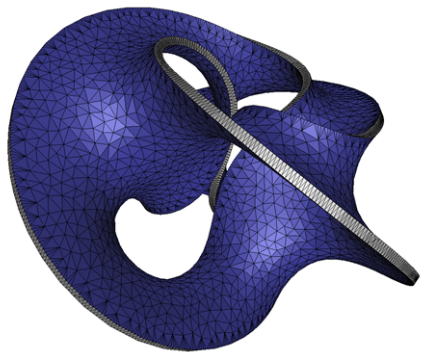


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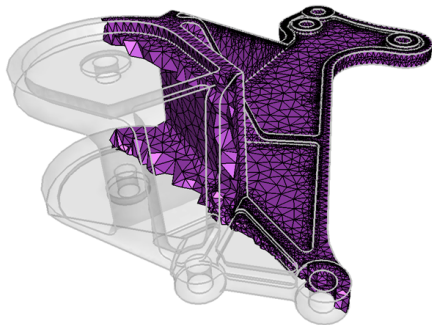
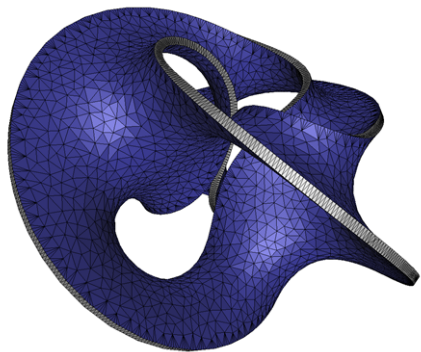
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- Burden is to show termination (by packing argument)

# Sampling Theorem

Theorem (Amenta-Bern 98, Cheng-D.-Edelsbrunner-Sullivan 01)

*If  $P \subset \Sigma$  is a discrete  $\varepsilon$ -sample of a smooth surface  $\Sigma$ , then for  $\varepsilon < 0.09$ ,  $\text{Del } P|_{\Sigma}$  satisfies:*

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- *Hausdorff distance* between  $\Sigma$  and  $\text{Del } P|_{\Sigma}$  is  $O(\varepsilon^2)$  of LFS.



# Sampling Theorem Modified

## Theorem (Boissonnat-Oudot 05)

If  $P \in \Sigma$  is such that each Voronoi edge-surface intersection  $x$  lies within  $\varepsilon \text{Lfs}(x)$  from a sample, then for  $\varepsilon < 0.09$ ,  $\text{Del } P|_{\Sigma}$  satisfies:

- It is *homeomorphic* to  $\Sigma$
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# Basic Delaunay Refinement

- 1 Initialize points  $P \subset \Sigma$ , compute  $\text{Del } P$
- 2 If some **condition** is not satisfied, insert a point  $c \in \Sigma$  into  $P$  and repeat
- 3 Return  $\text{Del } P|_{\Sigma}$

# Surface Delaunay Refinement

- 1 Initialize points  $P \subset \Sigma$ , compute  $\text{Del } P$
- 2 If some Voronoi edge intersects  $\Sigma$  at  $x$  with  $d(x, P) > \varepsilon LFS(x)$ , insert  $x$  in  $P$  and repeat
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- Require **topological disks** around vertices

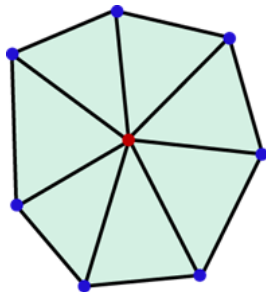


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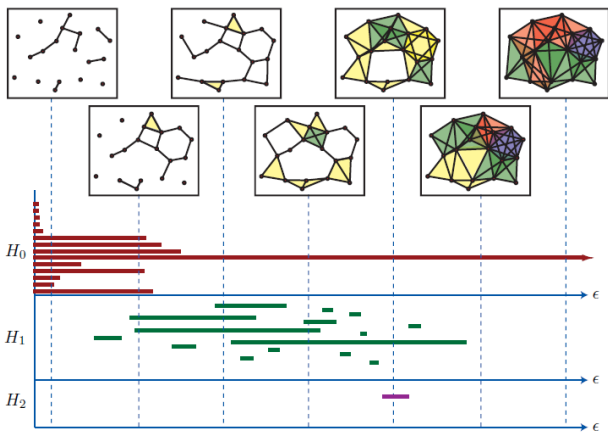
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  - ③ ***Hausdorff** distance between  $\Sigma$  and  $\text{Del } P|_{\Sigma}$  is  $O(\lambda^2)$  of LFS.*

# Data Analysis by Persistent Homology

- Persistent homology [Edelsbrunner-Letscher-Zomorodian 00], [Zomorodian-Carlsson 02]

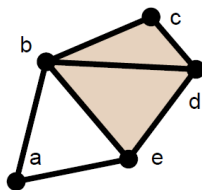


# Chain

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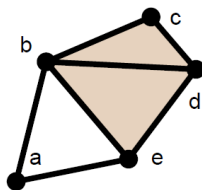
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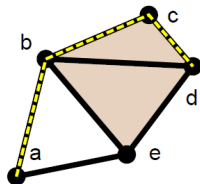
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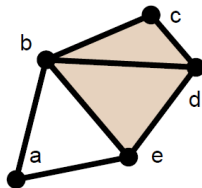
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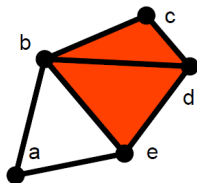


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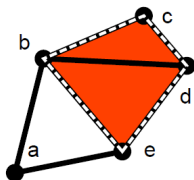


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1-boundary  $bc + cd + db + bd + de + eb = bc + cd + de + eb = \partial_2(bcd + bde)$   
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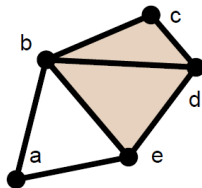
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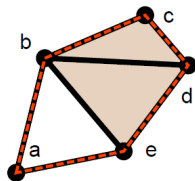


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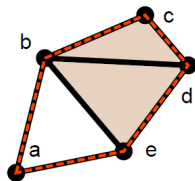


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- Each  $p$ -boundary is a  $p$ -cycle:  $\partial_p \circ \partial_{p+1} = 0$

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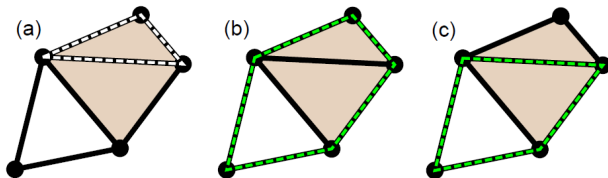
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(a) trivial (null-homologous) cycle; (b), (c) nontrivial homologous cycles

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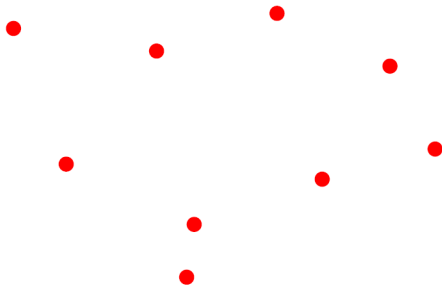
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## Proposition

For any finite set  $P \subset \mathbb{R}^d$  and any  $r \geq 0$ ,  $\mathcal{C}^r(P) \subseteq \mathcal{R}^r(P) \subseteq \mathcal{C}^{2r}(P)$

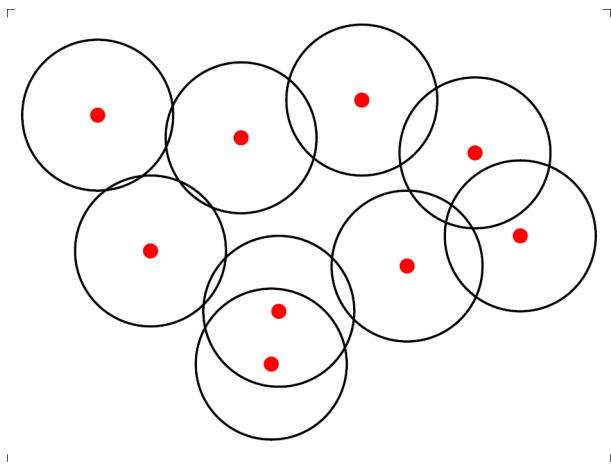
# Point set $P$

|

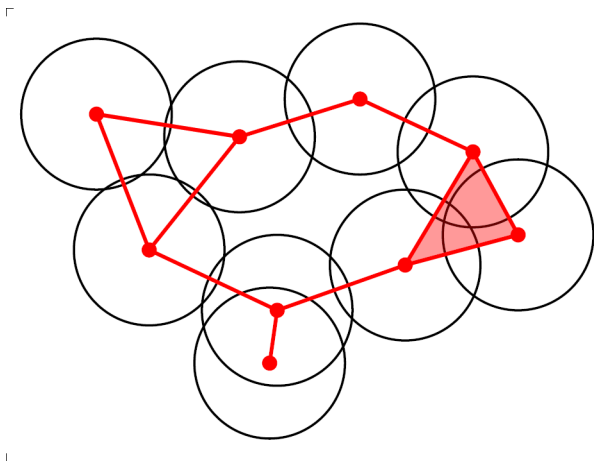


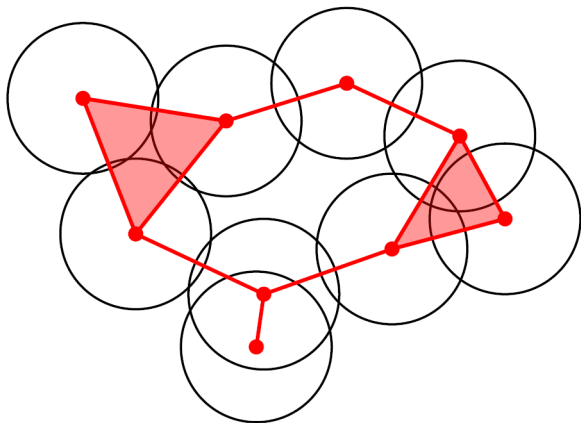
|

# Balls $B(p, r/2)$ for $p \in P$



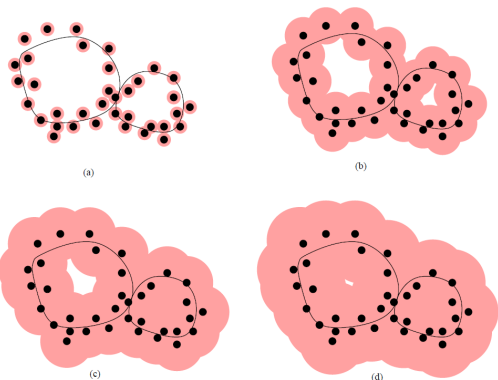
# Čech complex $\mathcal{C}^r(P)$



Rips complex  $\mathcal{R}^r(P)$ 

# Topological persistence

- $r(x) = d(x, P)$ : distance to point cloud  $P$
- **Sublevel sets**  $r^{-1}[0, a]$  are union of balls
- Evolution of the sublevel sets with increasing  $a$ —left hole **persists longer**
- Persistent homology quantizes this idea



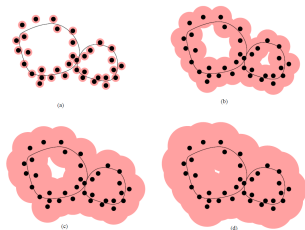


# Persistent Homology

- $f : \mathbb{T} \rightarrow \mathbb{R}$ ;  $\mathbb{T}_a = f^{-1}(-\infty, a]$ , the sublevel set
- $\mathbb{T}_a \subseteq \mathbb{T}_b$  for  $a \leq b$  provides inclusion map  $\iota : \mathbb{T}_a \rightarrow \mathbb{T}_b$
- Induced map  $\iota_* : H_p(\mathbb{T}_a) \rightarrow H_p(\mathbb{T}_b)$  giving the sequence

$$0 \rightarrow H_p(\mathbb{T}_{a_1}) \rightarrow H_p(\mathbb{T}_{a_2}) \rightarrow \cdots \rightarrow H_p(\mathbb{T}_{a_n}) \rightarrow H_p(\mathbb{T})$$

- Persistent homology classes: Image of  $f_p^{ij} : H_p(\mathbb{T}_{a_i}) \rightarrow H_p(\mathbb{T}_{a_j})$



# Continuous to Discrete

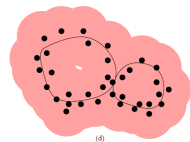
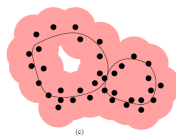
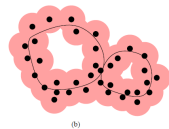
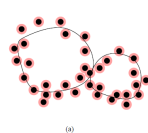
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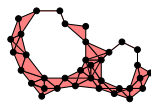


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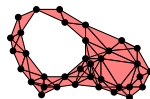
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(b)



(c)



(d)

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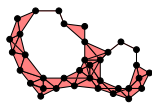
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$$\emptyset = K_0 \subseteq K_1 \subseteq \dots \subseteq K_n = K$$

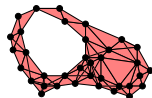
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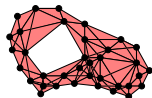
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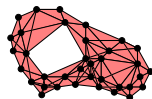
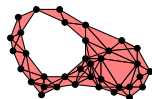
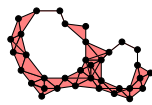
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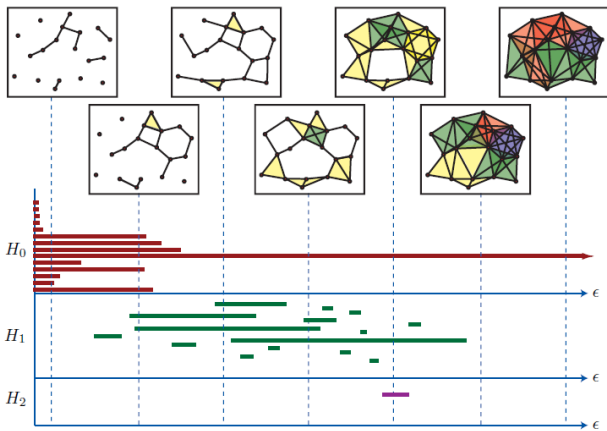
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- Birth and Death of homology classes



# Bar Codes

- birth-death and bar codes





# Persistence Diagram

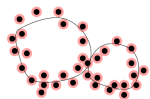
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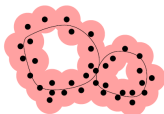
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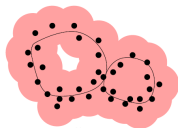
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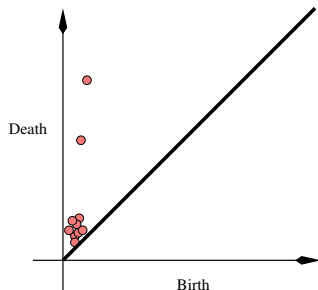
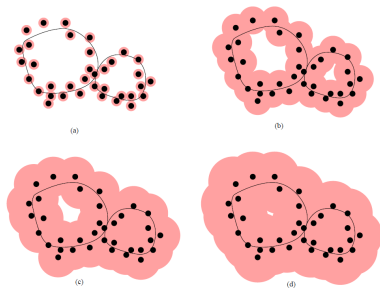
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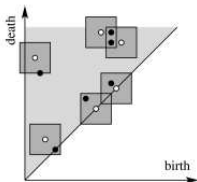
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# Stability of Persistence Diagram

- Bottleneck distance ( $\mathcal{C}$  all bijections)

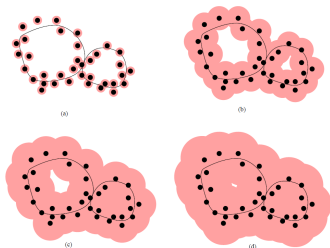
$$d_B(\text{Dgm}_\rho(f), \text{Dgm}_\rho(g)) := \inf_{c \in \mathcal{C}} \sup_{x \in \text{Dgm}_\rho(f)} \|x - c(x)\|$$



Theorem (Cohen-Steiner, Edelsbrunner, Harer 06)

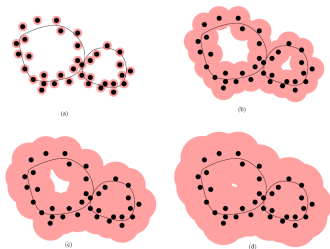
$$d_B(\text{Dgm}_\rho(f), \text{Dgm}_\rho(g)) \leq \|f - g\|_\infty$$

# Back to Point Data



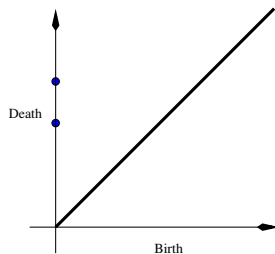
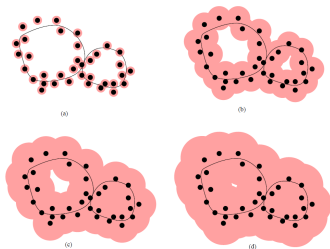
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- $d_P$  be the distance function from sample  $P$

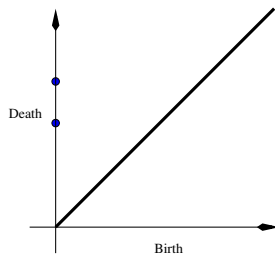
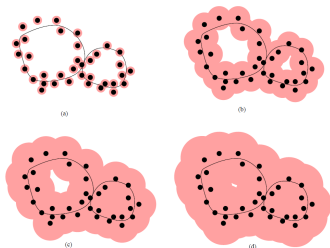
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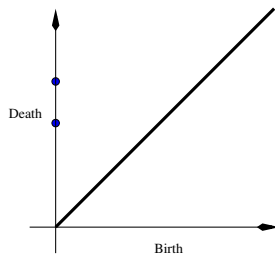
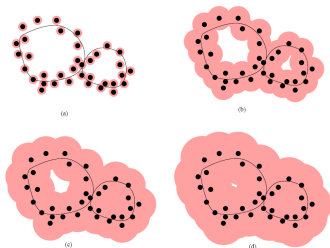


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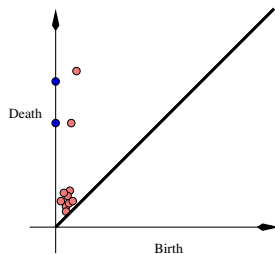
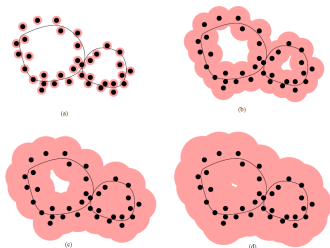
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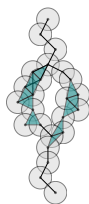
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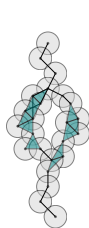
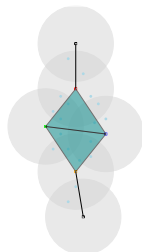
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Q



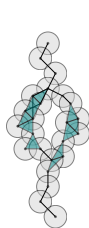
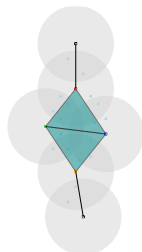
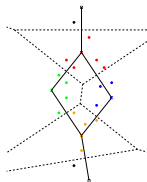
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- Efficient algorithm for Zigzag simplicial maps?

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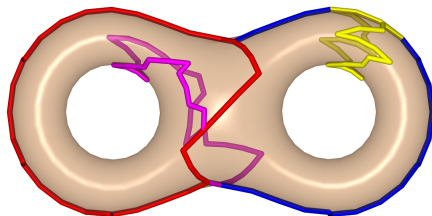
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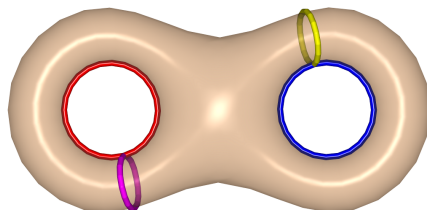
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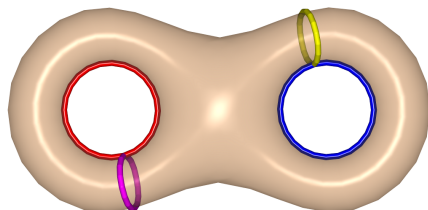
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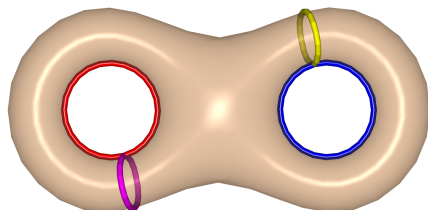
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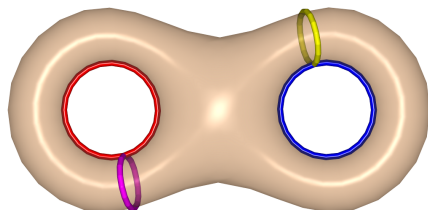
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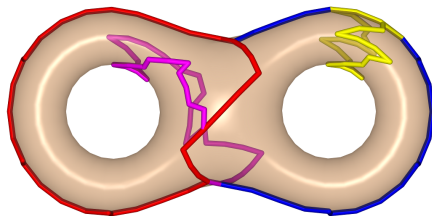
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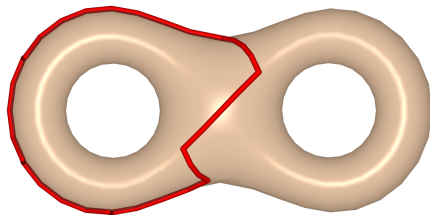
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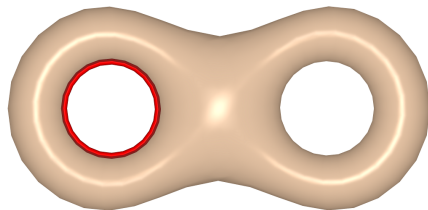
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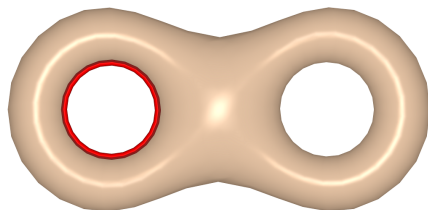
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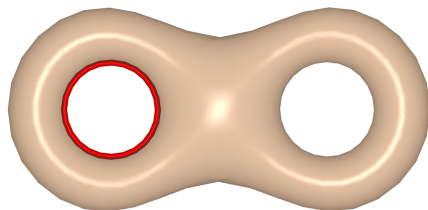


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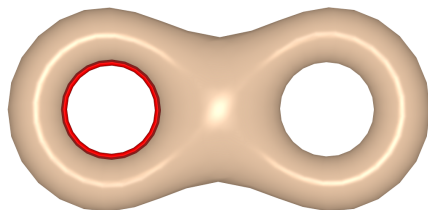
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# Thank You