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# Homology inference

F. Chazal  
Geometrica Group  
INRIA Saclay

If you have any question:  
[frederic.chazal@inria.fr](mailto:frederic.chazal@inria.fr)



# Weak feature size and stability

The *weak feature size* of a compact  $K \subset \mathbb{R}^d$ :

$$\text{wfs}(K) = \inf\{c > 0 : c \text{ is a critical value of } d_K\}$$

**Proposition:** [C-Lieutier'05] Let  $K, K' \subset \mathbb{R}^d$  be such that

$$d_H(K, K') < \varepsilon := \frac{1}{2} \min(\text{wfs}(K), \text{wfs}(K'))$$

Then for all  $0 < r \leq 2\varepsilon$ ,  $K^r$  and  $K'^r$  are homotopy equivalent.

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**Proof:** let  $\delta > 0$  be s.t.  $\delta + 2\varepsilon < \min(\text{wfs}(K), \text{wfs}(K'))$ .

$$\begin{array}{ccccc}
 K^\delta & \xrightarrow{a_0} & K^{\delta+\varepsilon} & \xrightarrow{a_1} & K^{\delta+2\varepsilon} \\
 & \searrow d_0 & & \searrow d_1 & \\
 & & K'^{\delta+\varepsilon} & & K'^{\delta+2\varepsilon} \\
 & \nearrow c_0 & & \nearrow c_1 & \\
 K'^\delta & \xrightarrow{b_0} & K'^{\delta+\varepsilon} & \xrightarrow{b_1} & K'^{\delta+2\varepsilon}
 \end{array}$$

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Compact set with positive wfs:



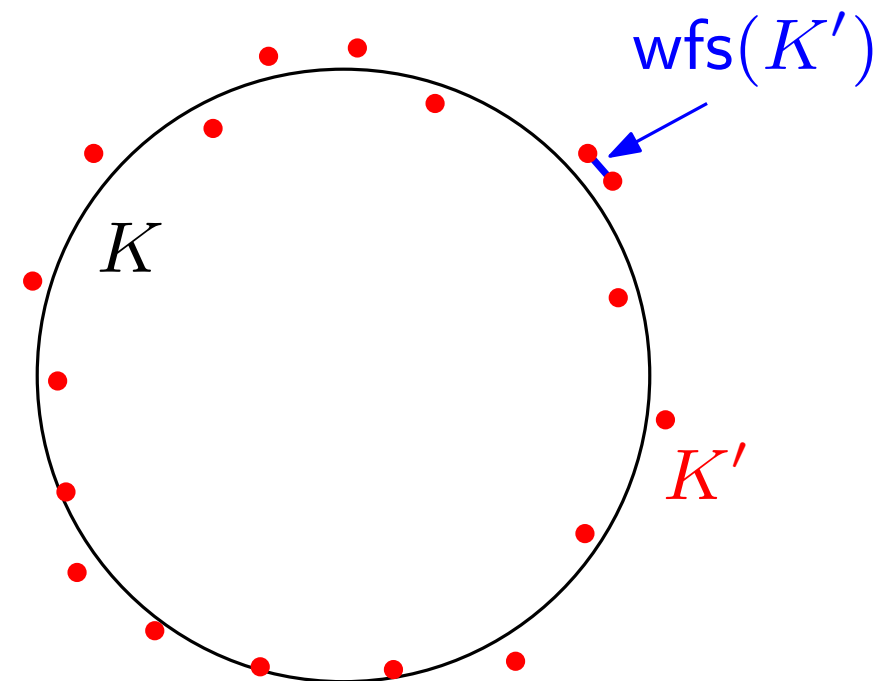
Stability properties



Large class of compact sets (including sub-analytic sets)



$K \rightarrow \text{wfs}(K)$  is not continuous (unstability of critical points).



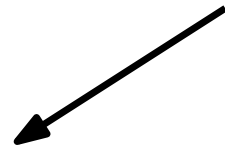
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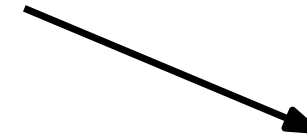
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Restrict to a smaller class of compact sets with some stability properties of the critical points.



## Option 2:

Try to get topological information about  $K$  without any assumption on  $\text{wfs}(K')$ .

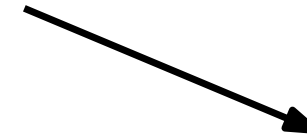
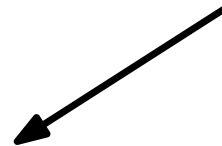
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Notion of  $\mu$ -critical points.  
Strong reconstruction results. (not in this course)

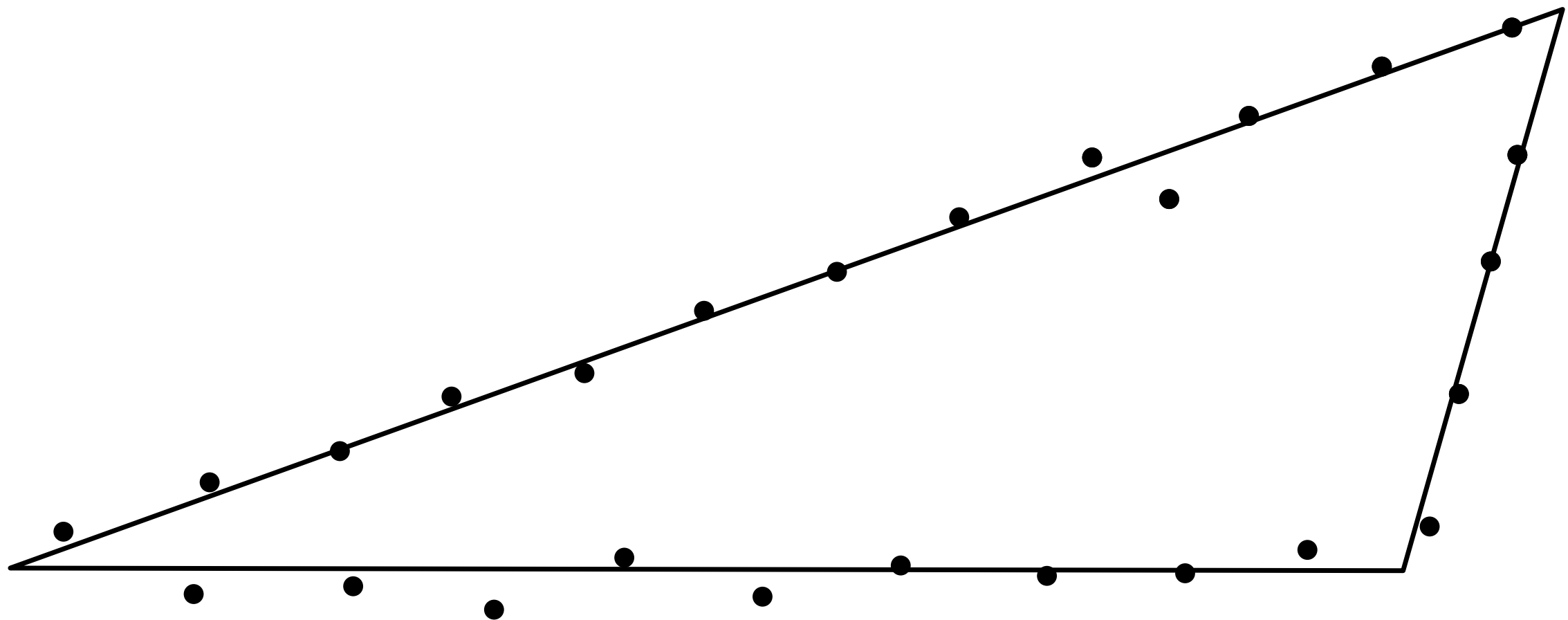
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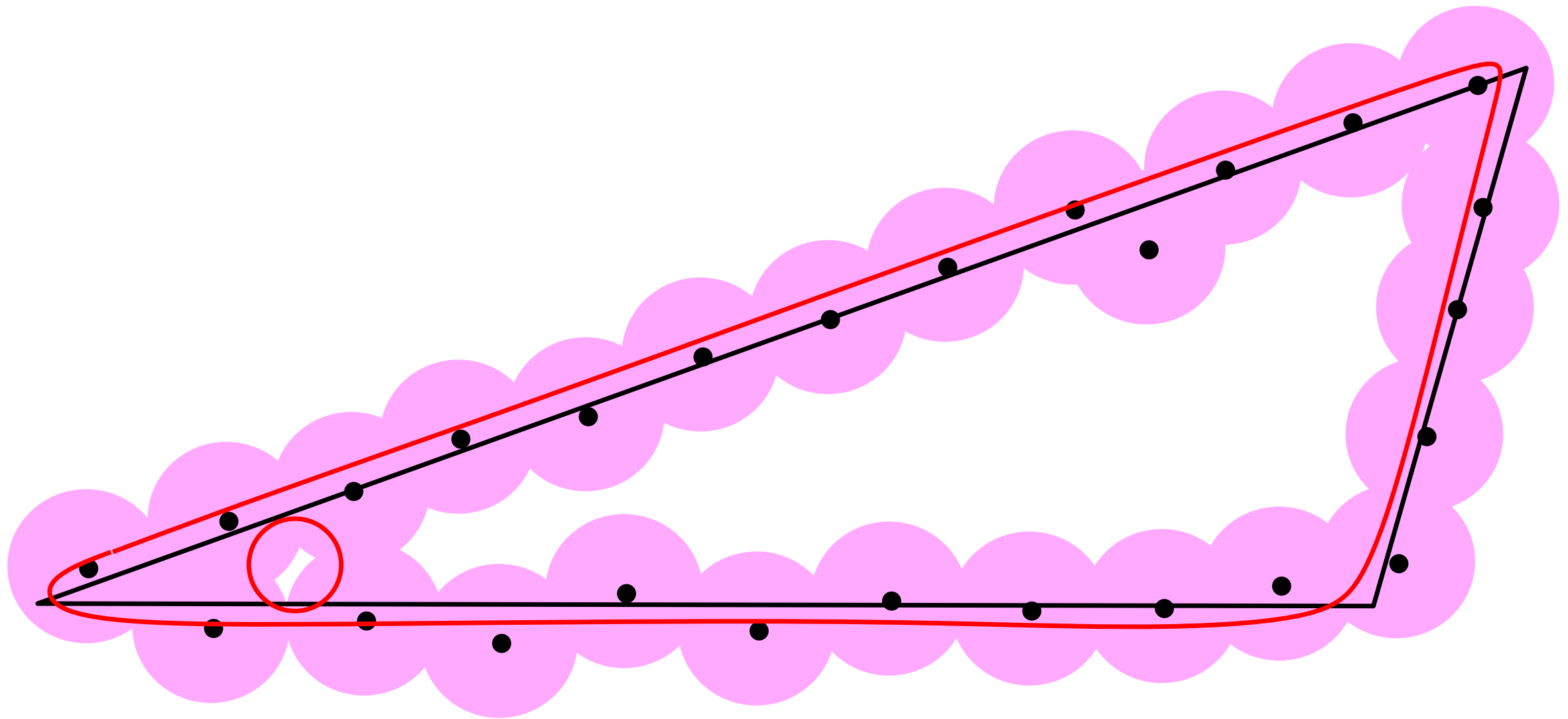
Persistence-based inference

Motivation: getting topological information  
without reconstructing



How to determine the number of “cycles” of the underlying shape from the point cloud approximation?

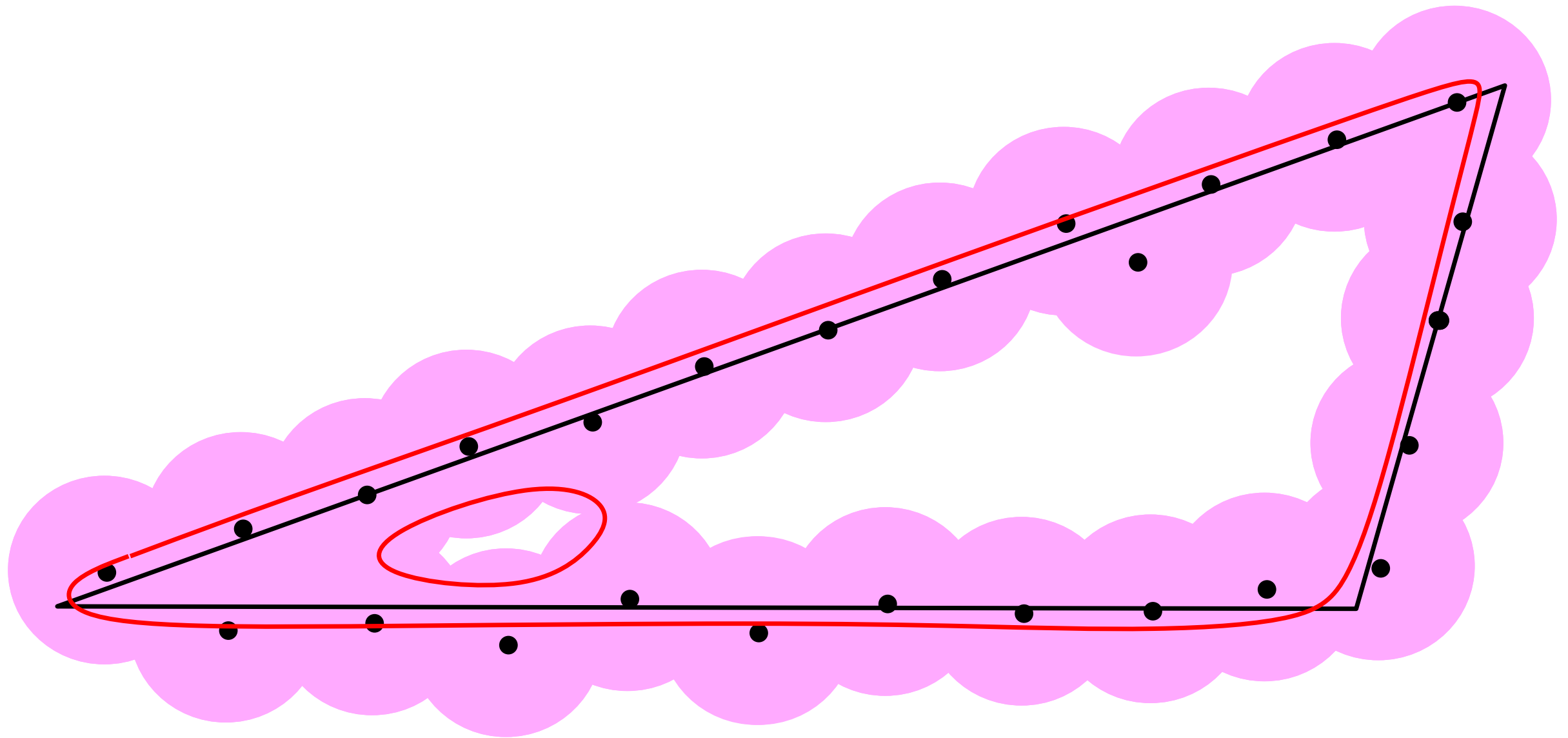
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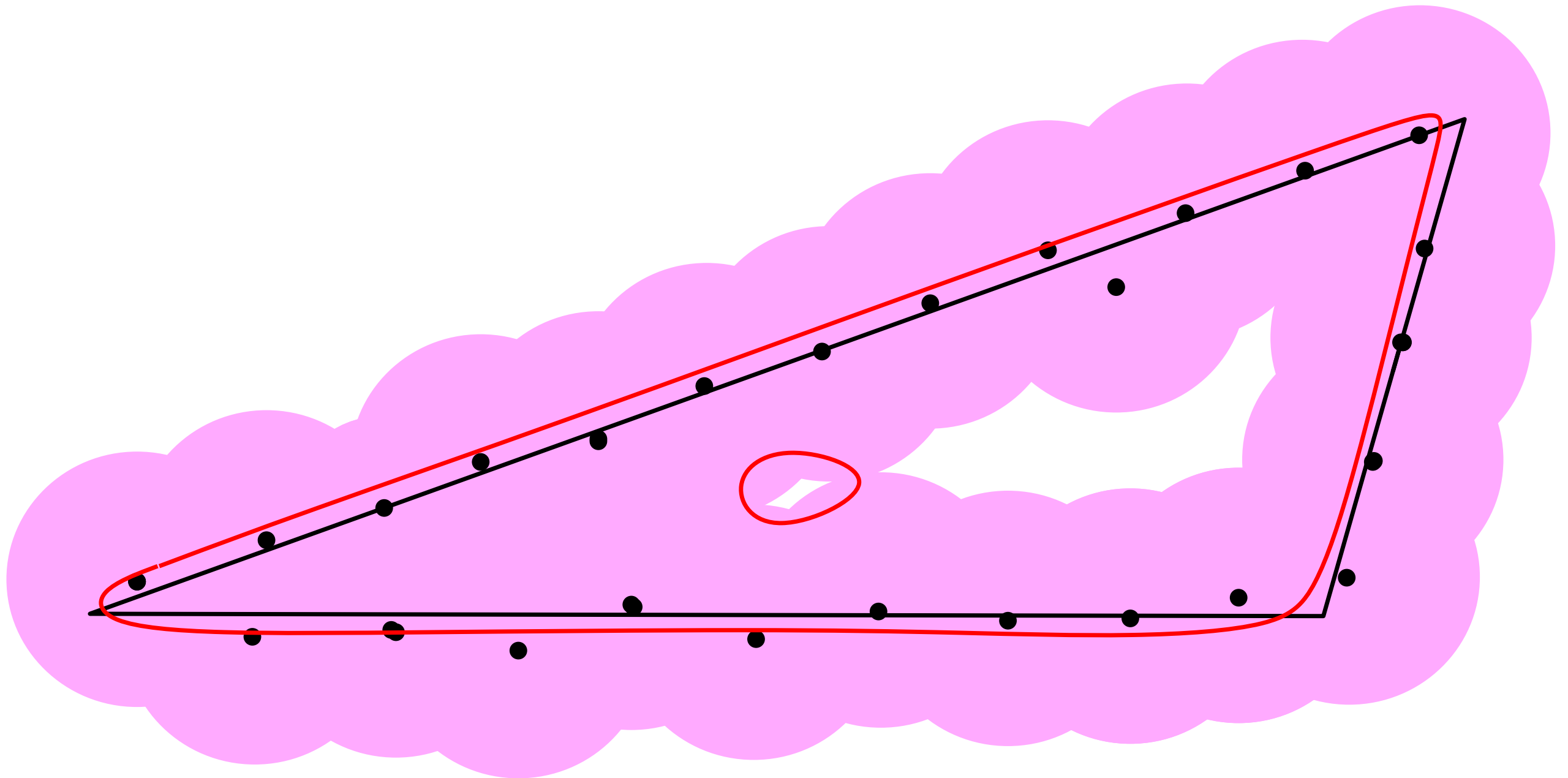


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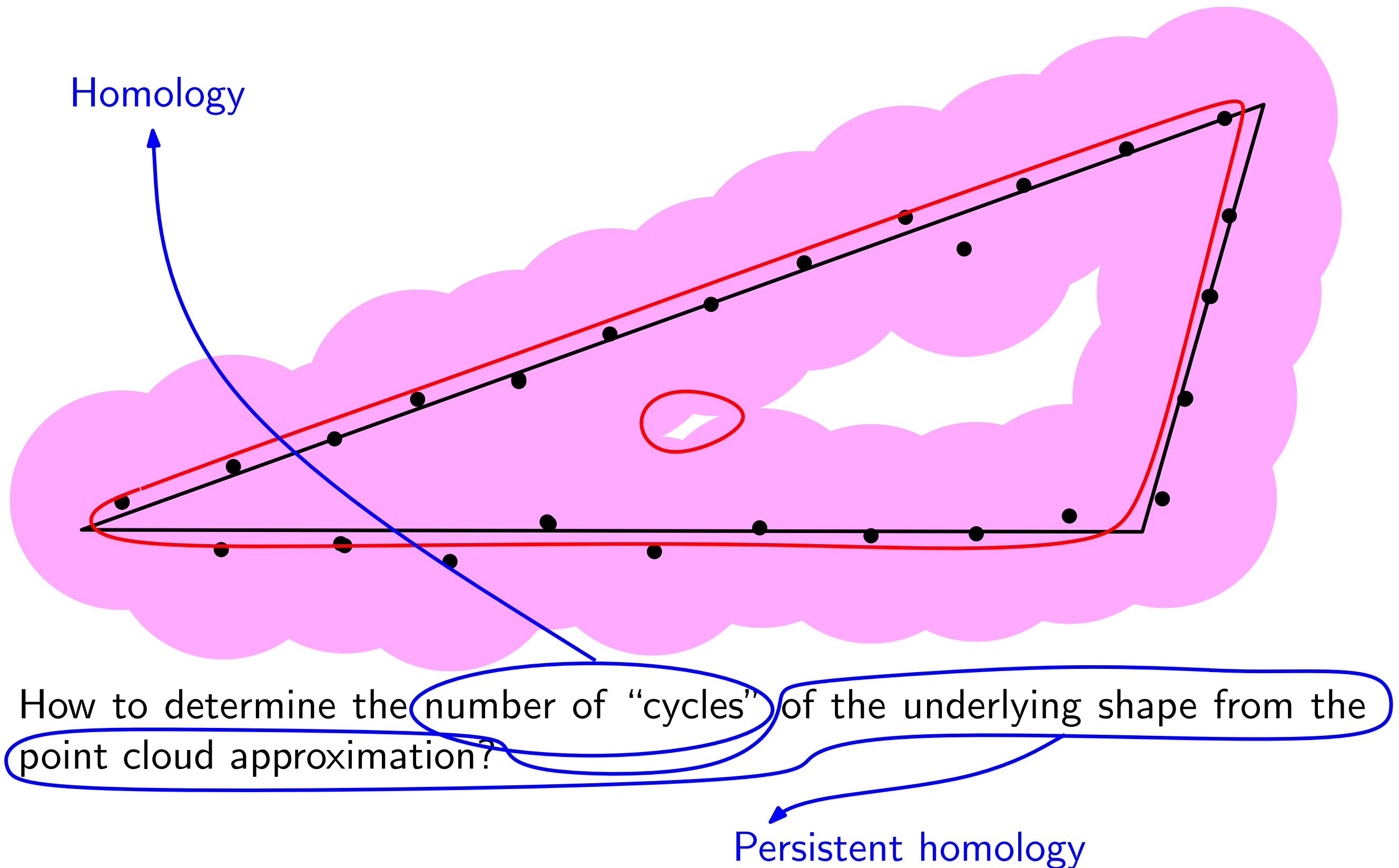
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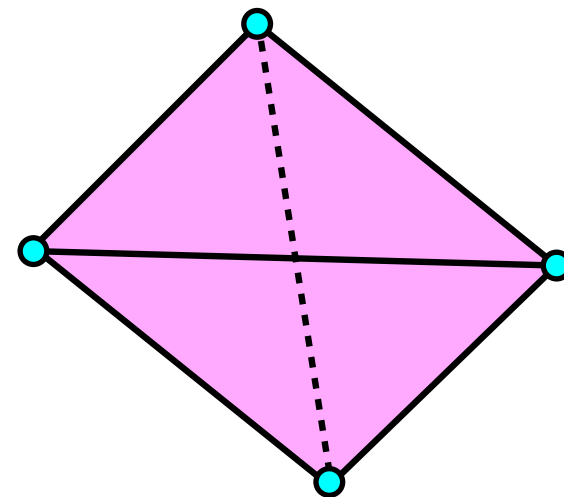
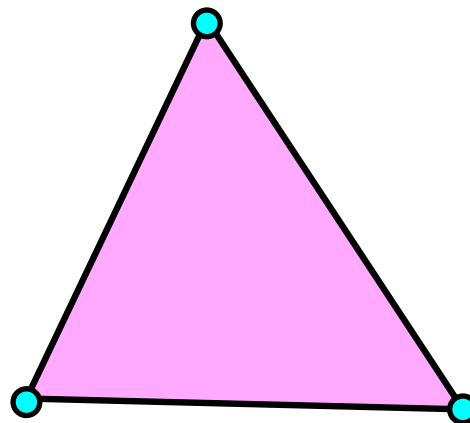
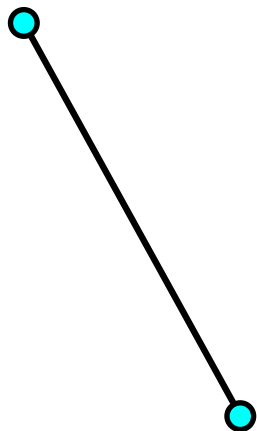


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# Simplices



0-simplex:  
vertex

1-simplex:  
edge

2-simplex:  
triangle

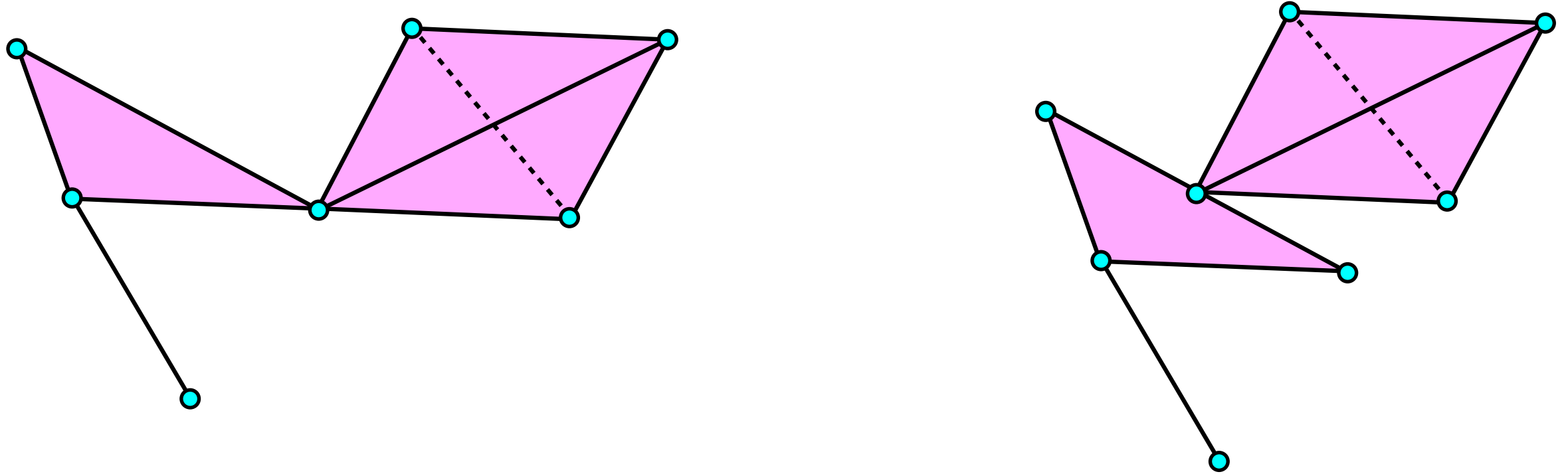
3-simplex:  
tetraedron

$v_0, v_1, \dots, v_k \in \mathbb{R}^d$  are **affinely independent** if

$$\left( \sum_{i=0}^k t_i v_i = 0 \text{ and } \sum_{i=0}^k t_i = 0 \right) \Rightarrow t_0 = t_1 = \dots = t_k = 0$$

In this case  $\sigma = [v_0, v_1, \dots, v_k]$  is a **simplex** of dimension  $d$ . A simplex generated by a subset of the vertices  $v_0, v_1, \dots, v_k$  of  $\sigma$  is a **face** of  $\sigma$ .

# Simplicial complexes

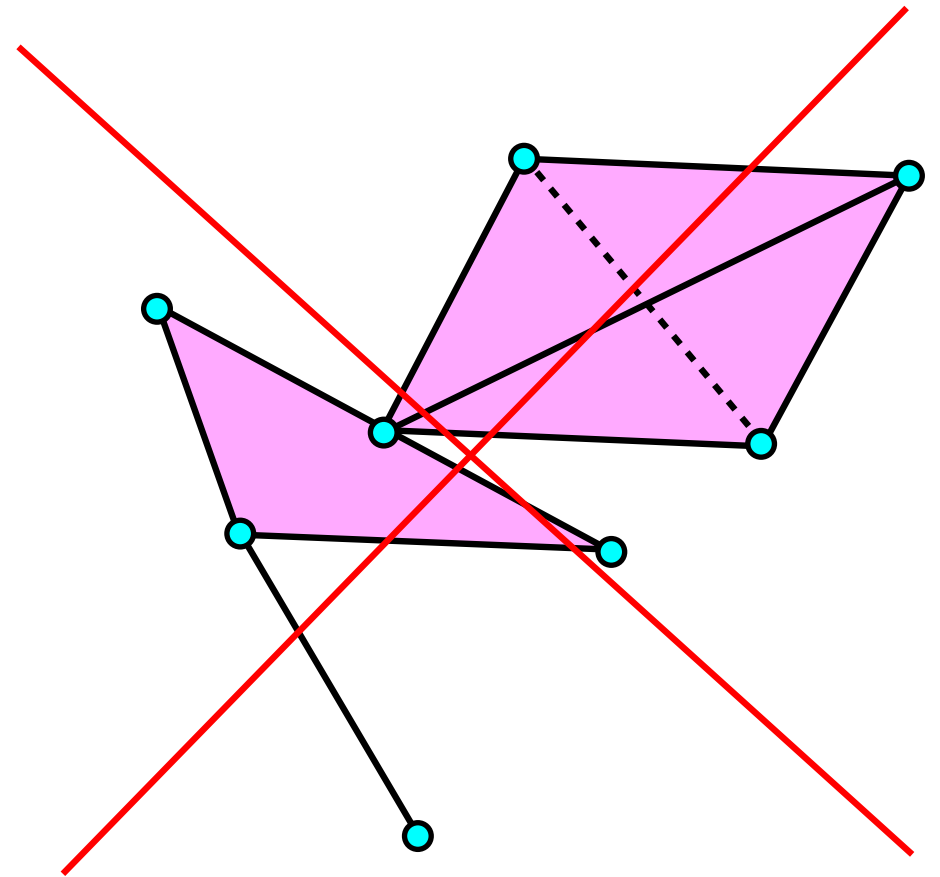
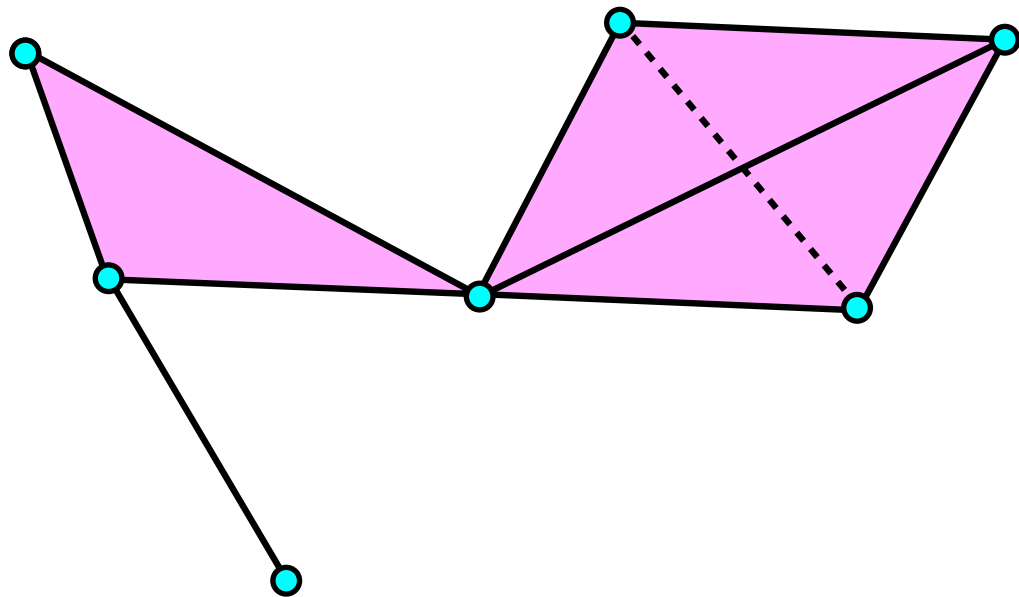


A **(finite) simplicial complex**  $C$  is a (finite) union of simplices s.t.

*i)* for any  $\sigma \in C$ , all the faces of  $\sigma$  are in  $C$ ,

*ii)* the intersection of any two simplices of  $C$  is either empty or a simplex which is their common face of highest dimension.

# Simplicial complexes

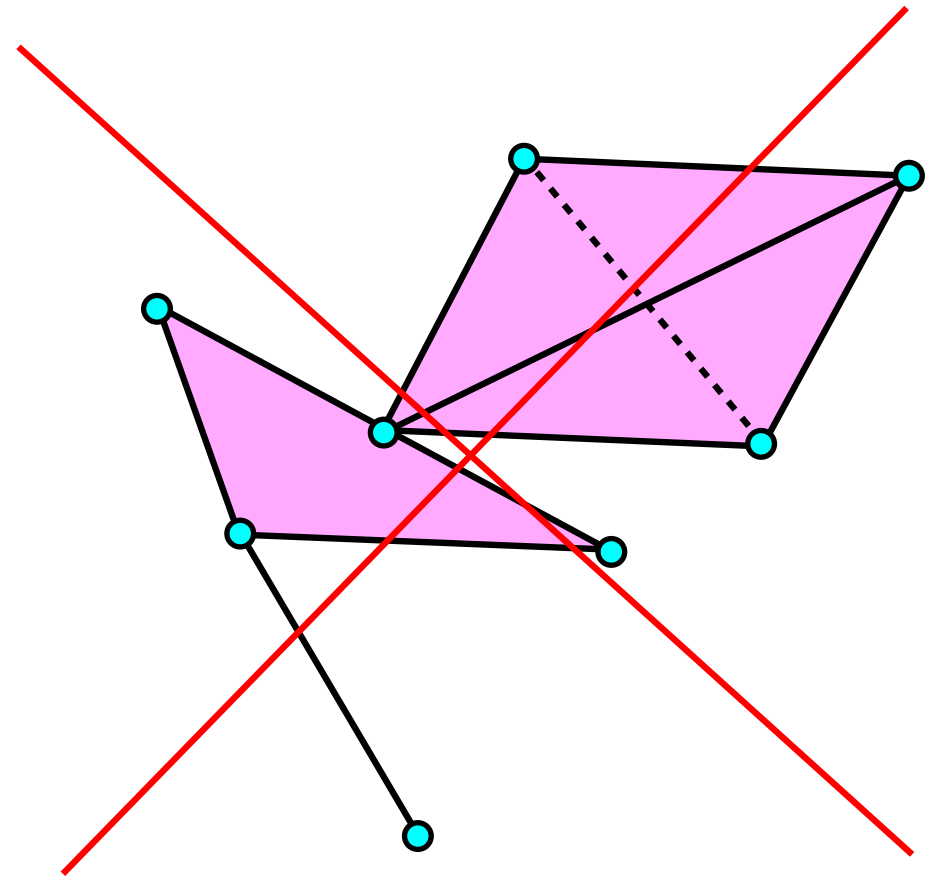
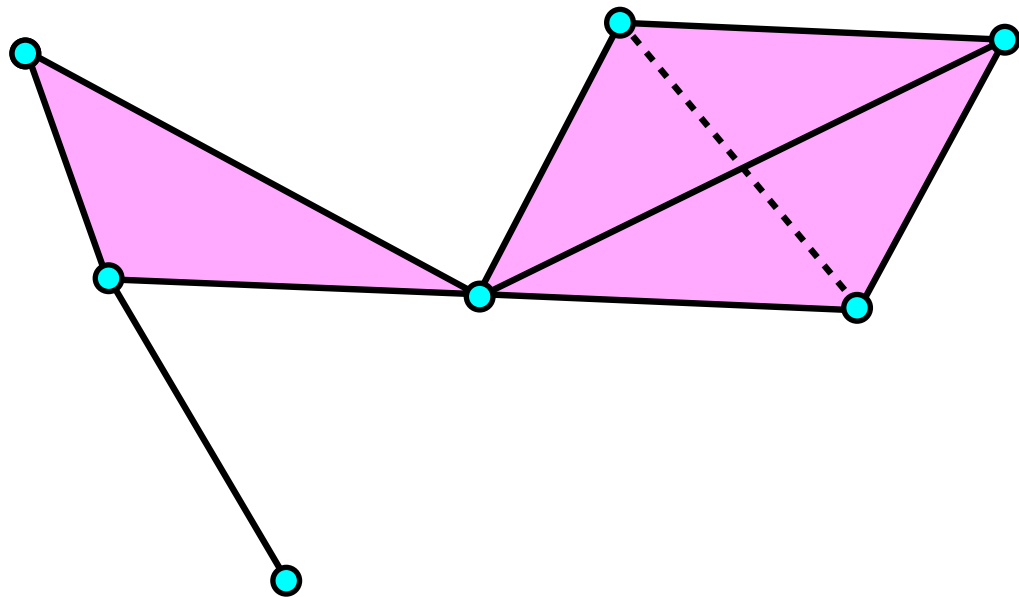


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**Faces:** the simplices of  $C$ .

**$j$ -skeleton:** the subcomplex made of the simplices of dimension at most  $j$ .

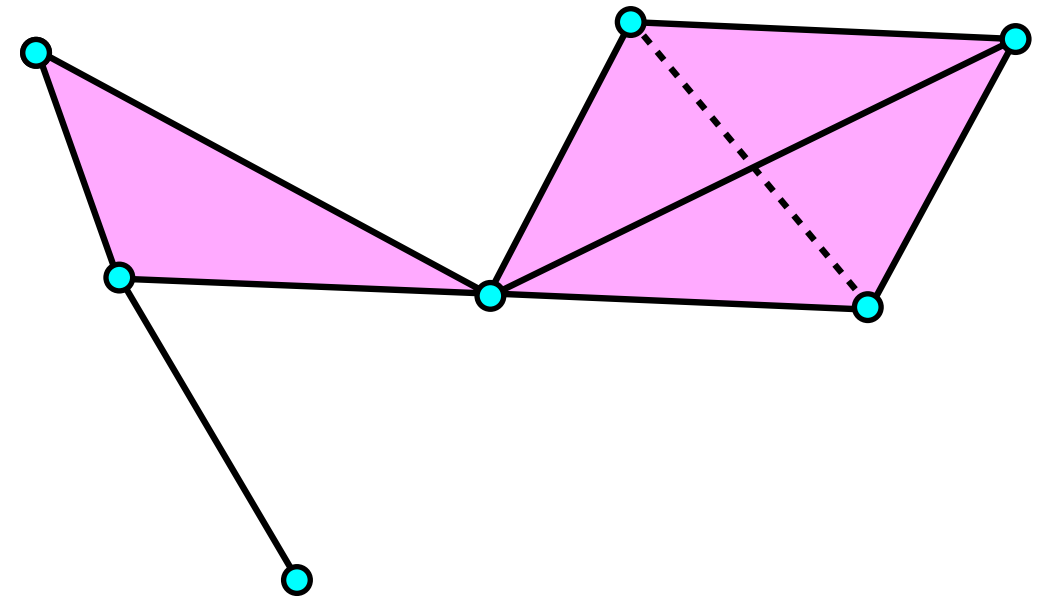
**Dimension** of  $C$ : the maximum of the dimensions of the faces.  $C$  is homogeneous of dimension  $n$  if any of its faces is a face of a  $n$ -dimensional simplex.

# Abstract simplicial complexes

Let  $P = \{p_1, \dots, p_n\}$  be a (finite) set. An **abstract simplicial complex**  $K$  with vertex set  $P$  is a set of subsets of  $P$  satisfying the two conditions :

1. The elements of  $P$  belong to  $K$ .
2. If  $\tau \in K$  and  $\sigma \subseteq \tau$ , then  $\sigma \in K$ .

The elements of  $K$  are the **simplices**.



Let  $\{e_1, \dots, e_n\}$  a basis of  $\mathbb{R}^n$  }. The **geometric realization** of  $K$  is the (geometric) subcomplex  $|K|$  of the simplex spanned by  $e_1, \dots, e_n$  such that:

$$[e_{i_0} \cdots e_{i_k}] \in |K| \text{ iff } \{p_{i_0}, \dots, p_{i_k}\} \in K$$

$|K|$  is a topological space (subspace of an Euclidean space)!

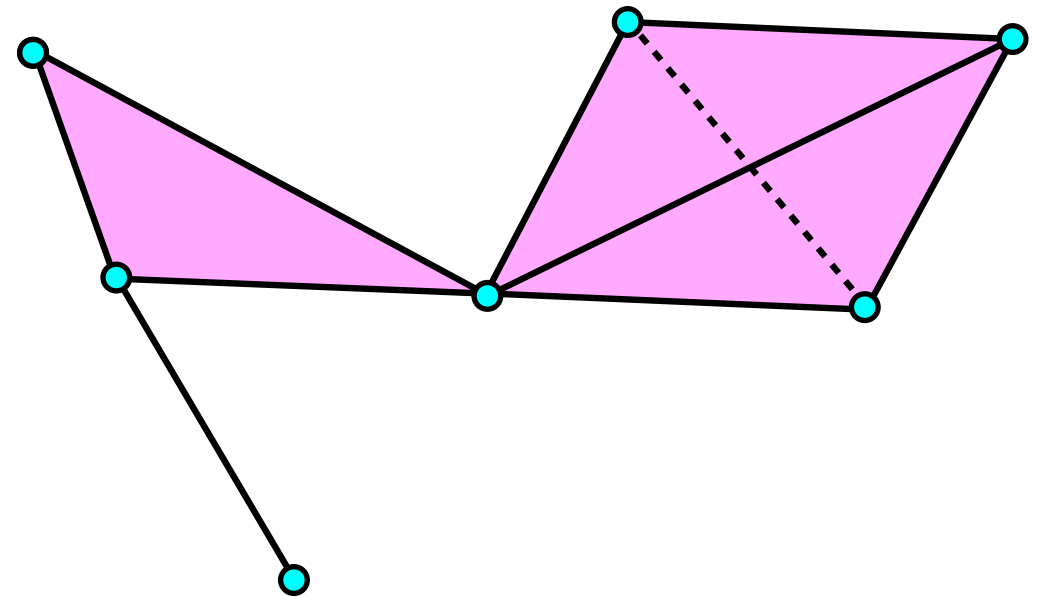


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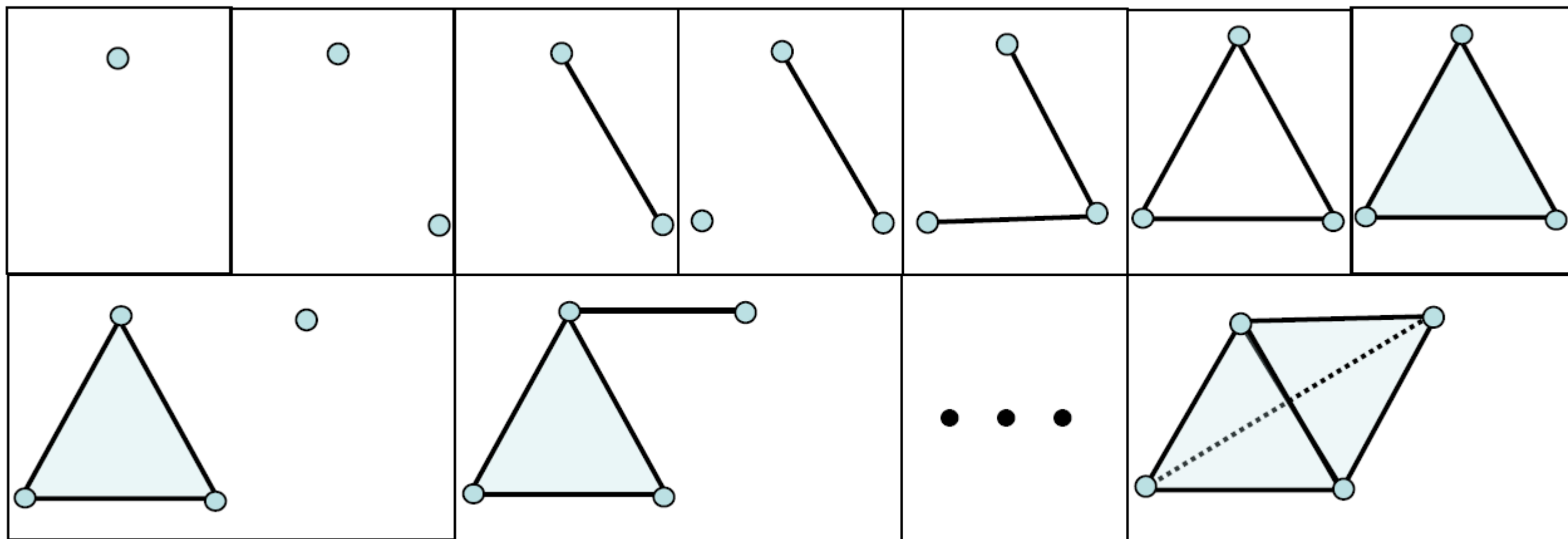
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## IMPORTANT

Simplicial complexes can be seen at the same time as geometric/topological spaces (good for top./geom. inference) and as combinatorial objects (abstract simplicial complexes, good for computations).

# Filtrations of simplicial complexes



A **filtration** of a (finite) simplicial complex  $K$  is a sequence of subcomplexes such that

i)  $\emptyset = K^0 \subset K^1 \subset \dots \subset K^m = K$ ,

ii)  $K^{i+1} = K^i \cup \sigma^{i+1}$  where  $\sigma^{i+1}$  is a simplex of  $K$ .

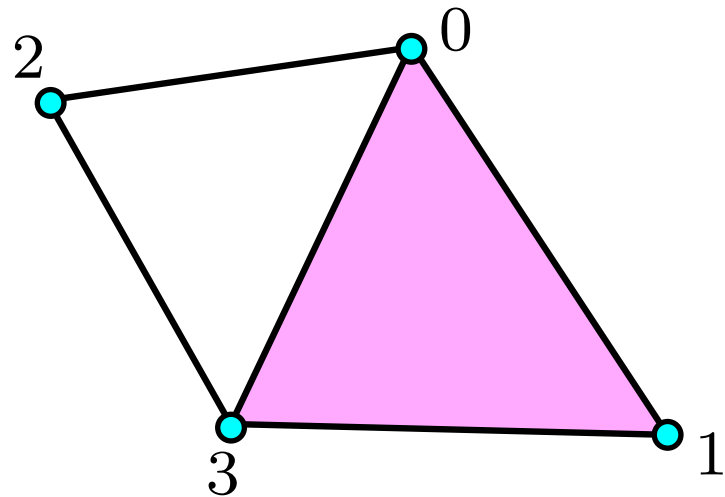
# Example: filtration associated to a function

- $f$  a real valued function defined on the vertices of  $K$
- For  $\sigma = [v_0, \dots, v_k] \in K$ ,  $f(\sigma) = \max_{i=0, \dots, k} f(v_i)$
- The simplices of  $K$  are ordered according increasing  $f$  values (and dimension in case of equal values on different simplices).

$\Rightarrow$  The sublevel sets filtration.

Exercise: show that this is a filtration.

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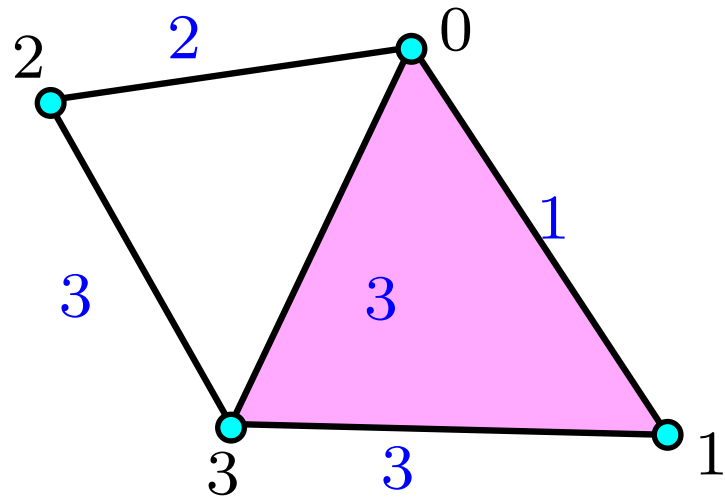


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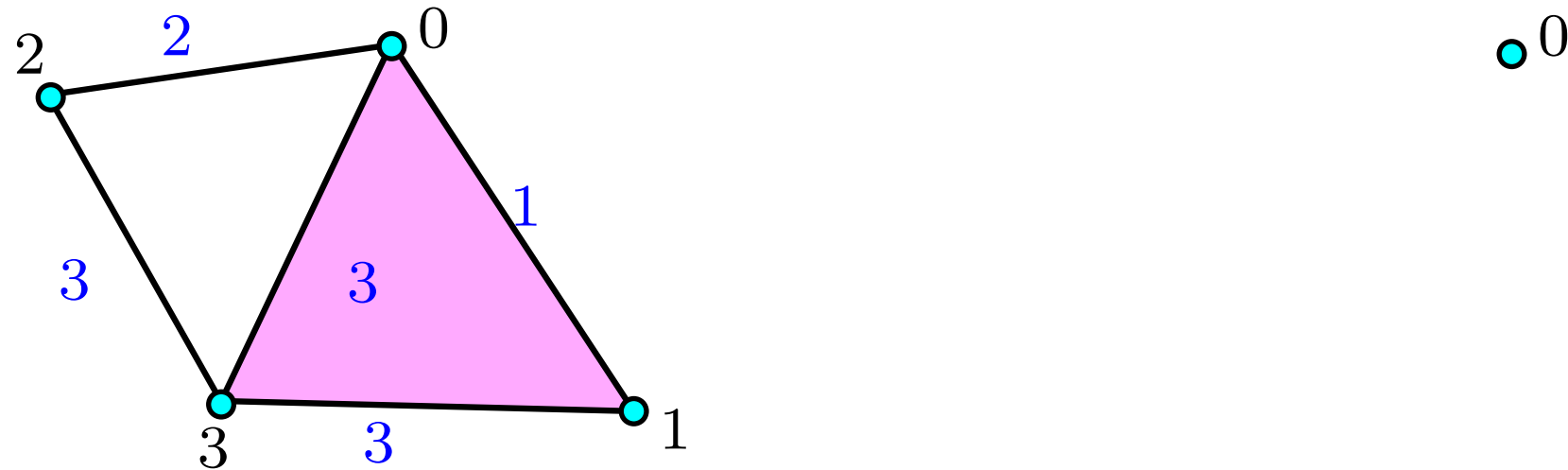


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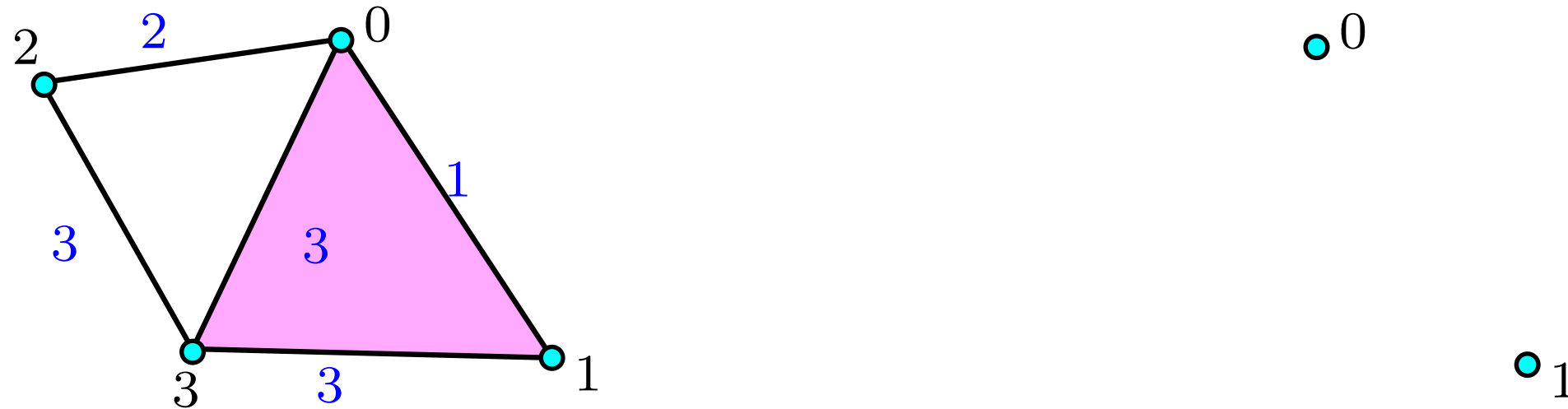


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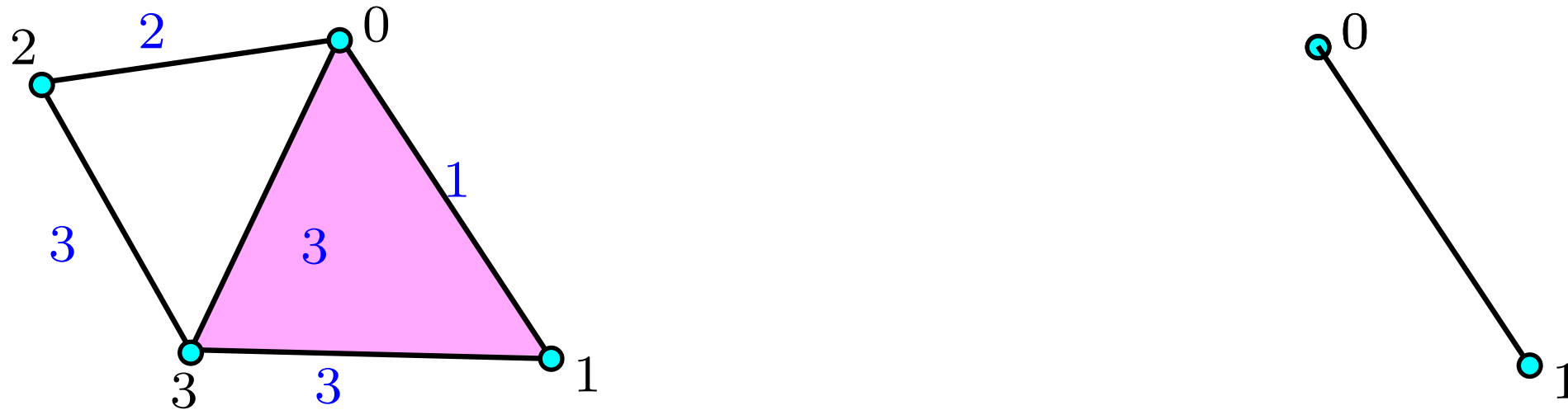


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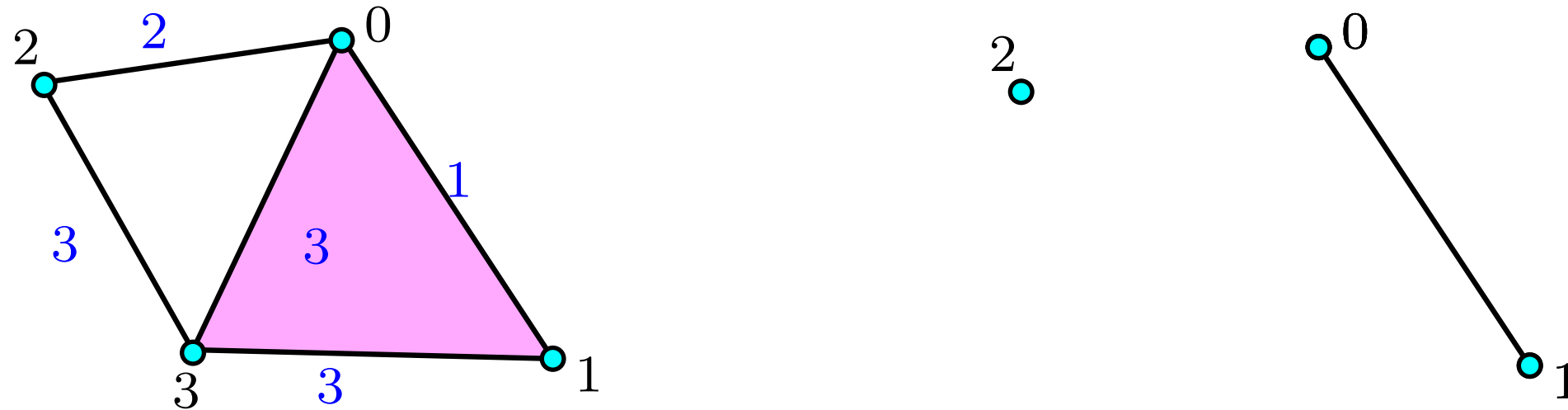
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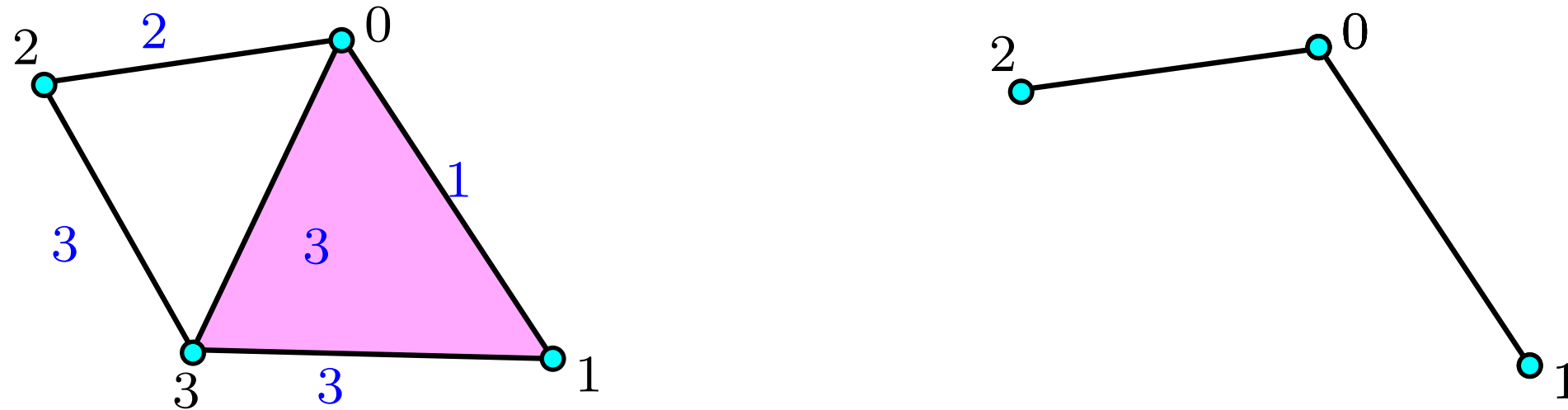


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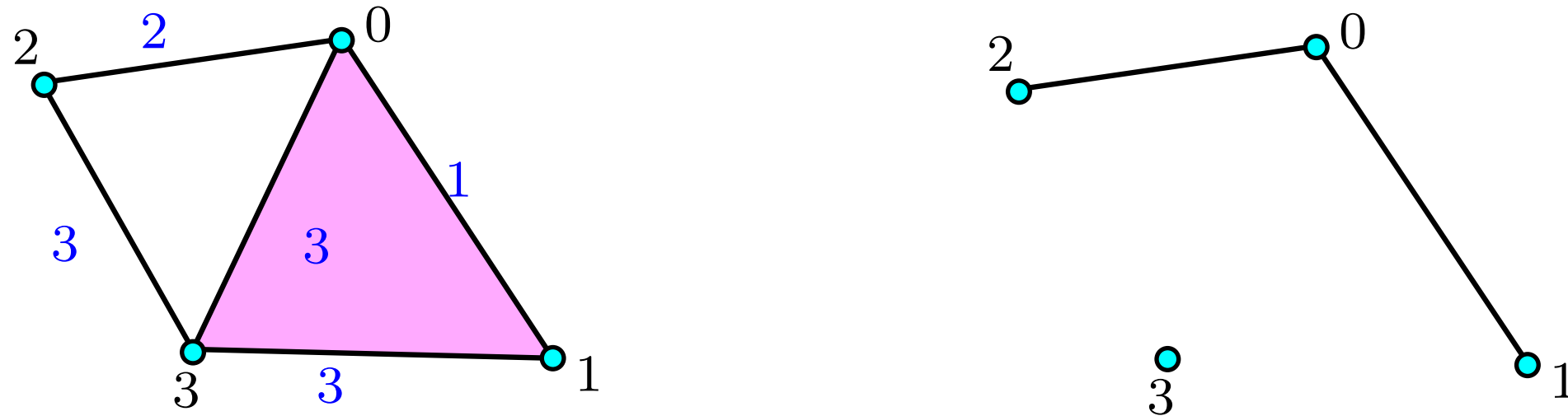


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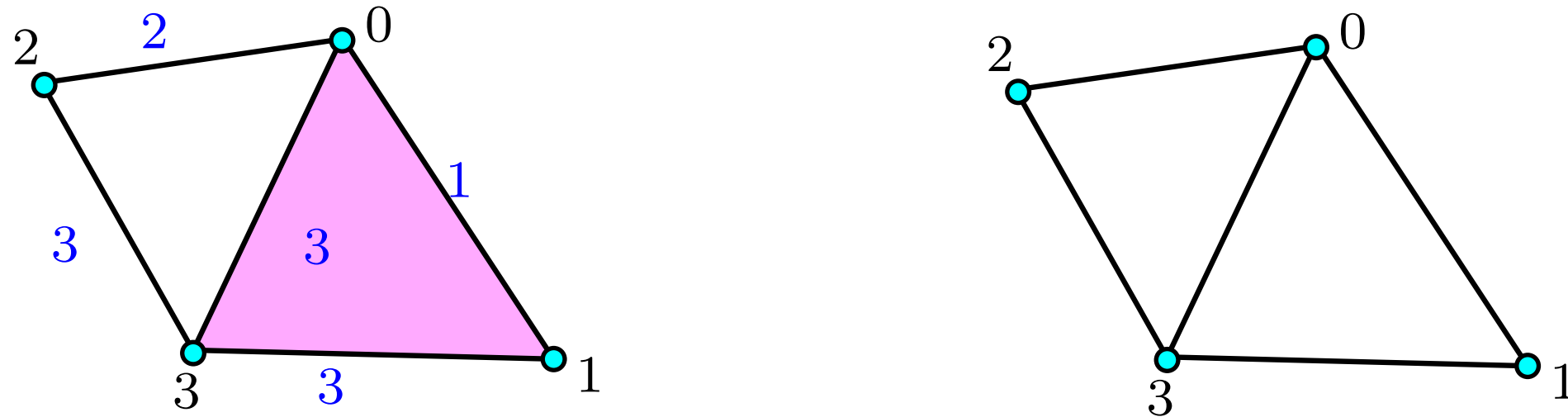


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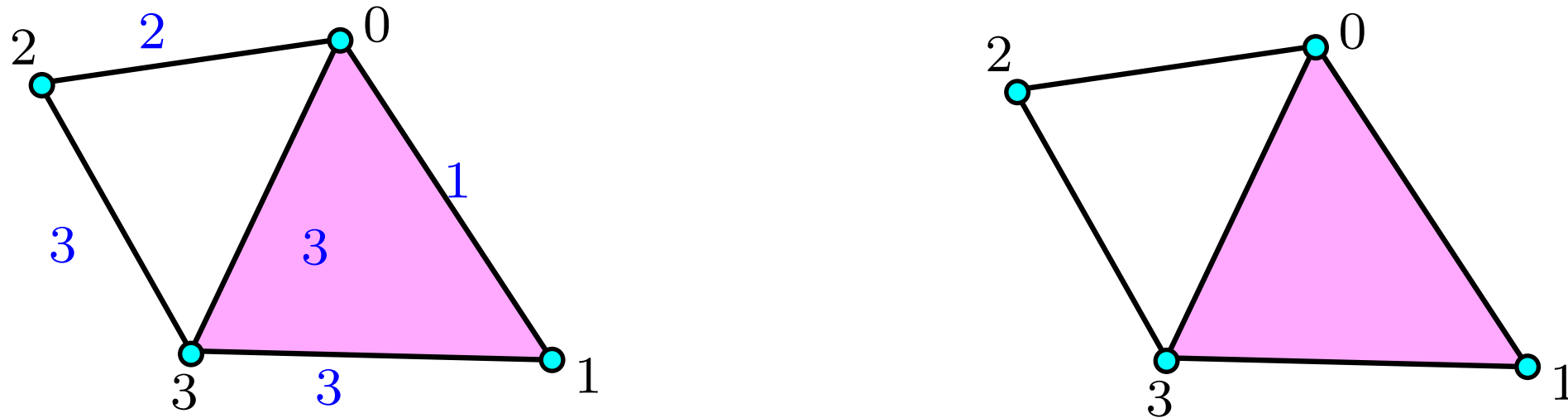


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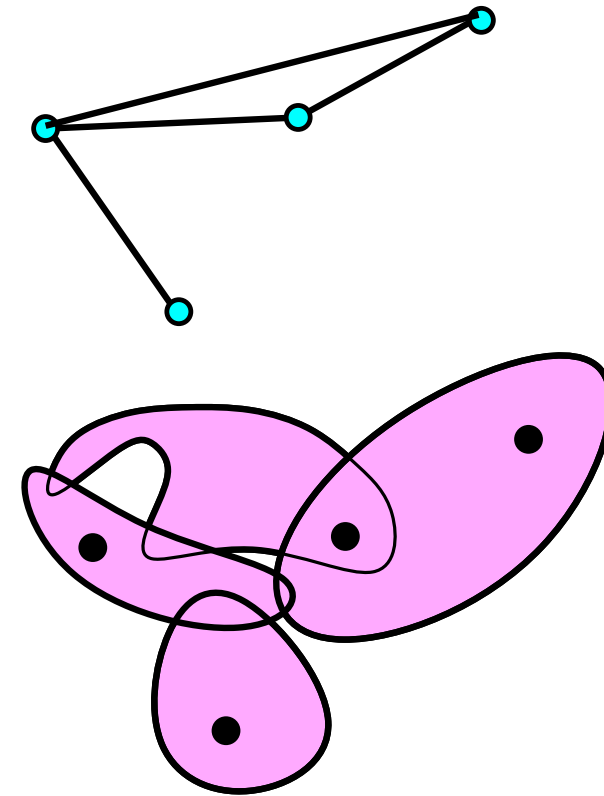
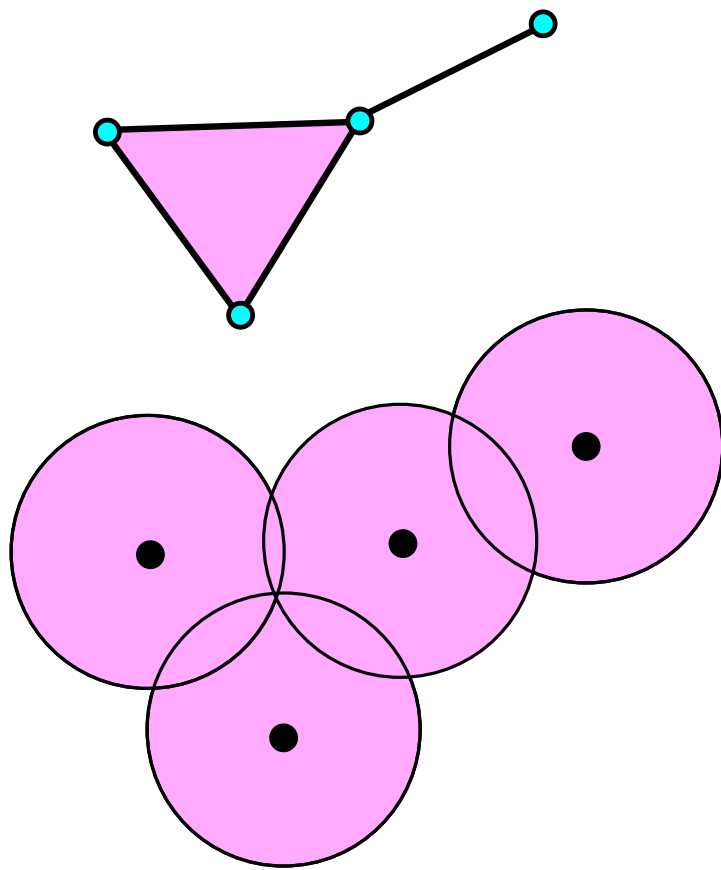


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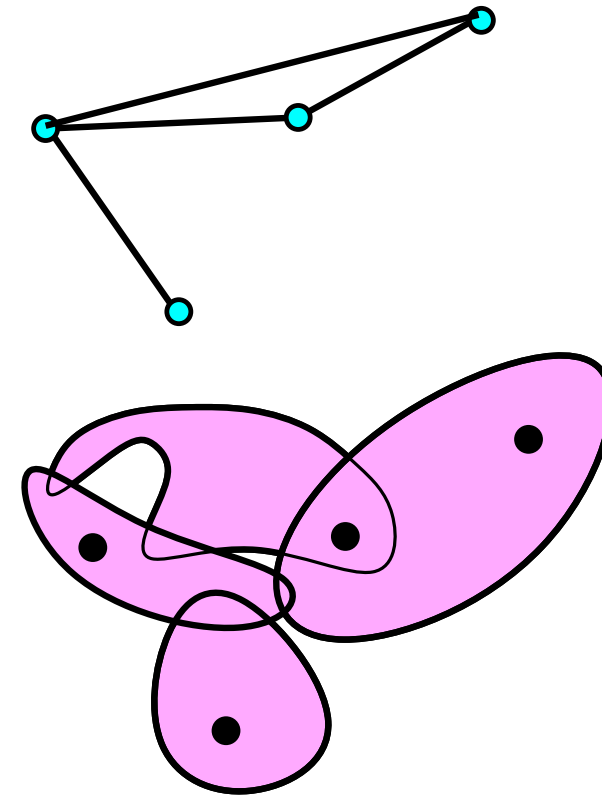
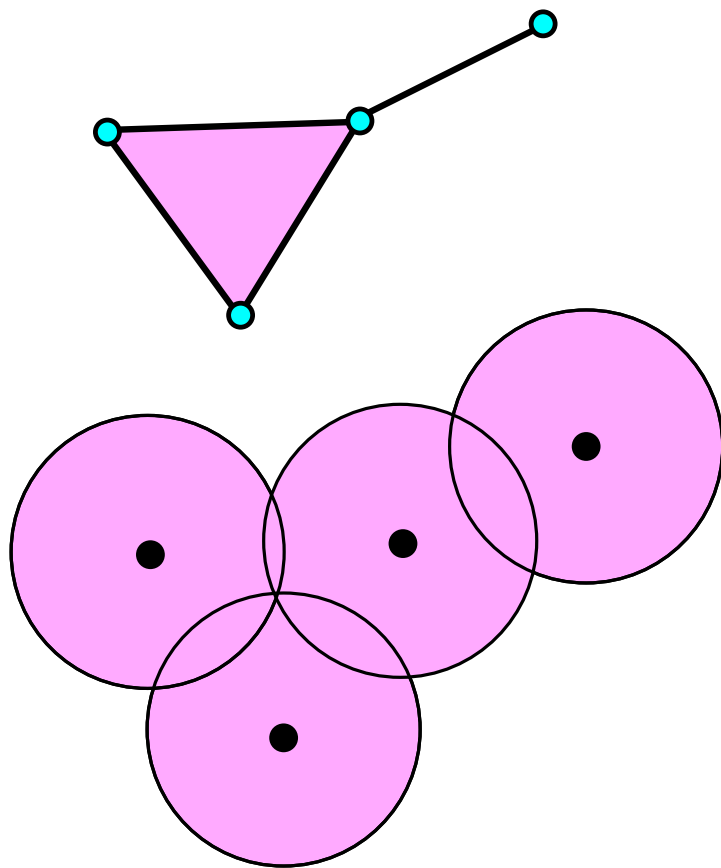
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# Example: The Čech complex



- Let  $\mathcal{U} = (U_i)_{i \in I}$  be a covering of a topological space  $X$  by open sets:  $X = \bigcup_{i \in I} U_i$ .
- The Čech complex  $C(\mathcal{U})$  associated to the covering  $\mathcal{U}$  is the simplicial complex defined by:
  - the vertex set of  $C(\mathcal{U})$  is the set of the open sets  $U_i$
  - $[U_{i_0}, \dots, U_{i_k}]$  is a  $k$ -simplex in  $C(\mathcal{U})$  iff  $\bigcap_{j=0}^k U_{i_j} \neq \emptyset$ .

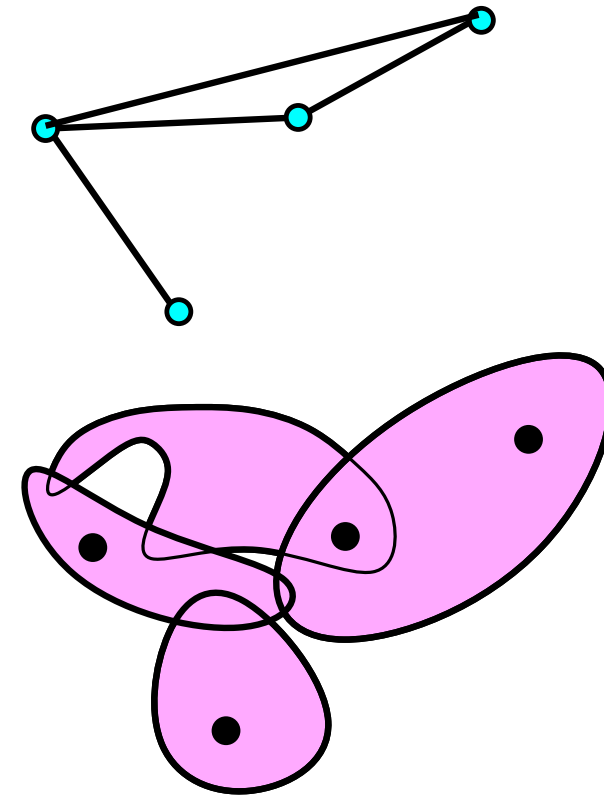
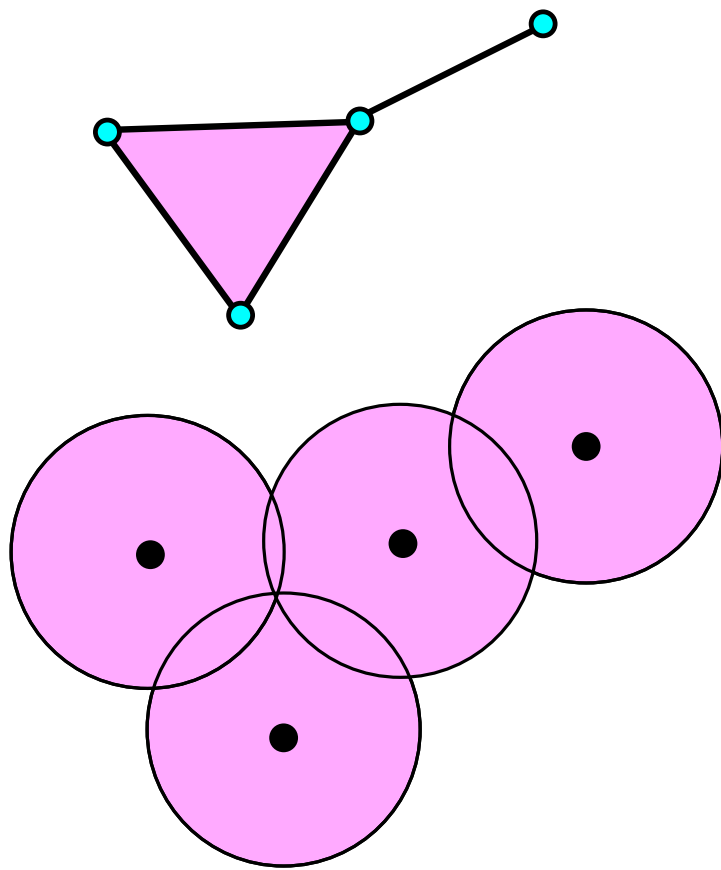
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**Nerve theorem (Leray):** If all the intersections between opens in  $\mathcal{U}$  are either empty or contractible then  $C(\mathcal{U})$  and  $X = \cup_{i \in I} U_i$  are homotopy equivalent.

$\Rightarrow$  The combinatorics of the covering (a simplicial complex) carries the topology of the space.

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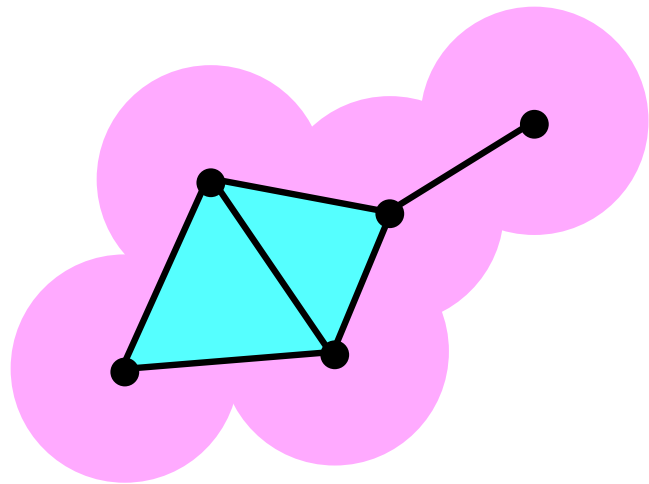
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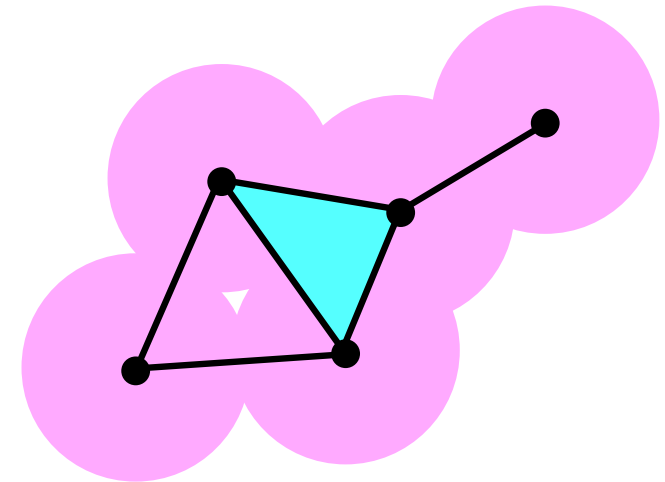
**Warning:** even when the open sets are euclidean balls, the computation of the Čech complex is a very difficult task!



# Example: the Rips complex



Rips vs Čech



Let  $L = \{p_0, \dots, p_n\}$  be a (finite) point cloud (in a metric space).

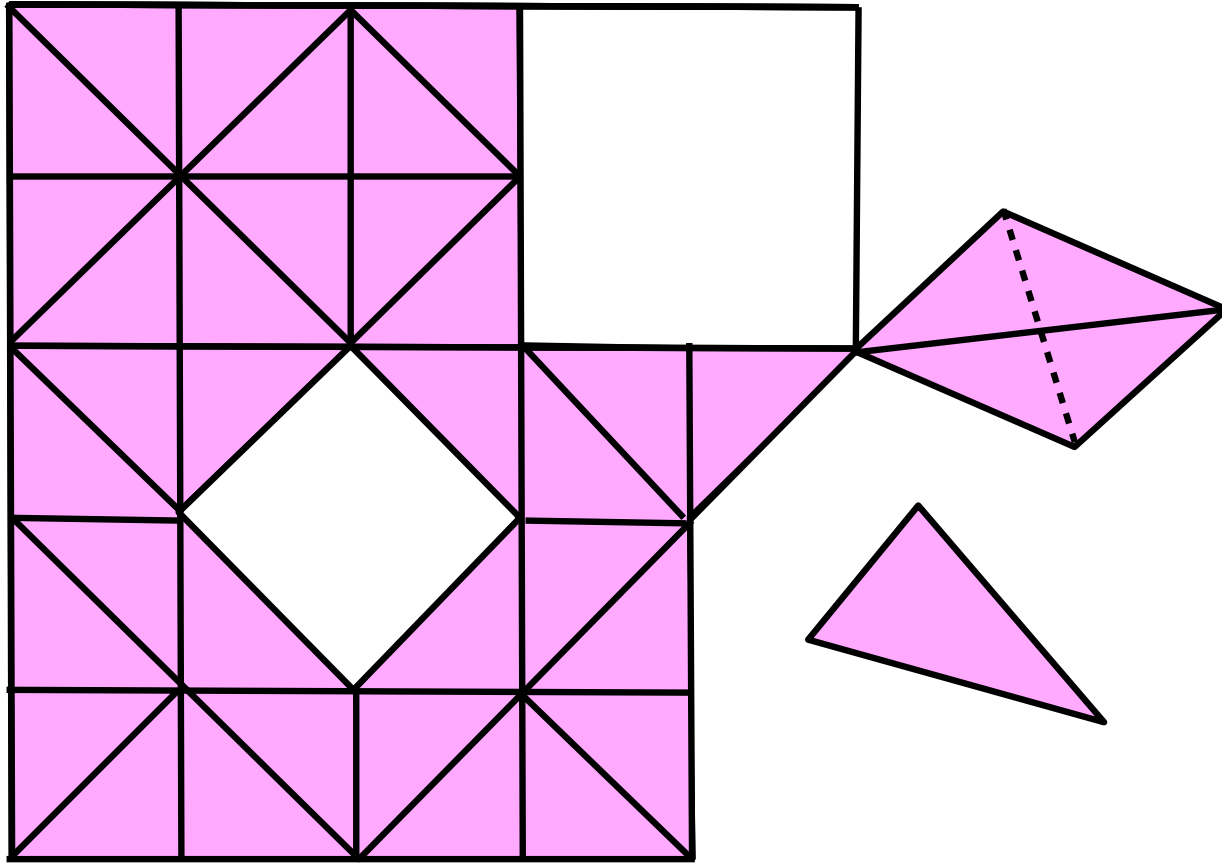
The **Rips complex**  $\mathcal{R}^\alpha(L)$ : for  $p_0, \dots, p_k \in L$ ,

$$\sigma = [p_0 p_1 \cdots p_k] \in \mathcal{R}^\alpha(L) \quad \text{iff} \quad \forall i, j \in \{0, \dots, k\}, \quad d(p_i, p_j) \leq \alpha$$

- Easy to compute and fully determined by its 1-skeleton
- Rips-Čech interleaving: for any  $\alpha > 0$ ,

$$\mathcal{C}^{\frac{\alpha}{2}}(L) \subseteq \mathcal{R}^\alpha(L) \subseteq \mathcal{C}^\alpha(L) \subseteq \mathcal{R}^{2\alpha}(L) \subseteq \dots$$

# Homology of simplicial complexes



- 2 connected components
- Intuitively: 2 cycles

Topological invariants:

- Number of connected components
- Number of cycles: how to define a cycle?
- Number of voids: how to define a void?
- ...

(Simplicial) homology and  
Betti numbers

In the following: homology with coefficient in  $\mathbb{Z}/2$

Refs: J.R. Munkres, *Elements of Algebraic Topology*, Addison-Wesley, 1984.  
A. Hatcher, *Algebraic Topology*, Cambridge University Press 2002.

# The space of $k$ -chains

Let  $K$  be a  $d$ -dimensional simplicial complex. Let  $k \in \{0, 1, \dots, d\}$  and  $\{\sigma_1, \dots, \sigma_p\}$  be the set of  $k$ -simplices of  $K$ .

$k$ -chain:

$$c = \sum_{i=1}^p \varepsilon_i \sigma_i \quad \text{with} \quad \varepsilon_i \in \mathbb{Z}/2\mathbb{Z} = \{0, 1\}$$

Sum of  $k$ -chains:

$$c + c' = \sum_{i=1}^p (\varepsilon_i + \varepsilon'_i) \sigma_i \quad \text{and} \quad \lambda.c = \sum_{i=1}^p (\lambda \varepsilon'_i) \sigma_i$$

where the sums  $\varepsilon_i + \varepsilon'_i$  and the products  $\lambda \varepsilon_i$  are modulo 2.

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The space  $\mathcal{C}_k(K)$  of  $k$ -chains is a  $\mathbb{Z}/2$ -vector space

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Geometric interpretation:

$k$ -chain = union of  $k$ -simplices

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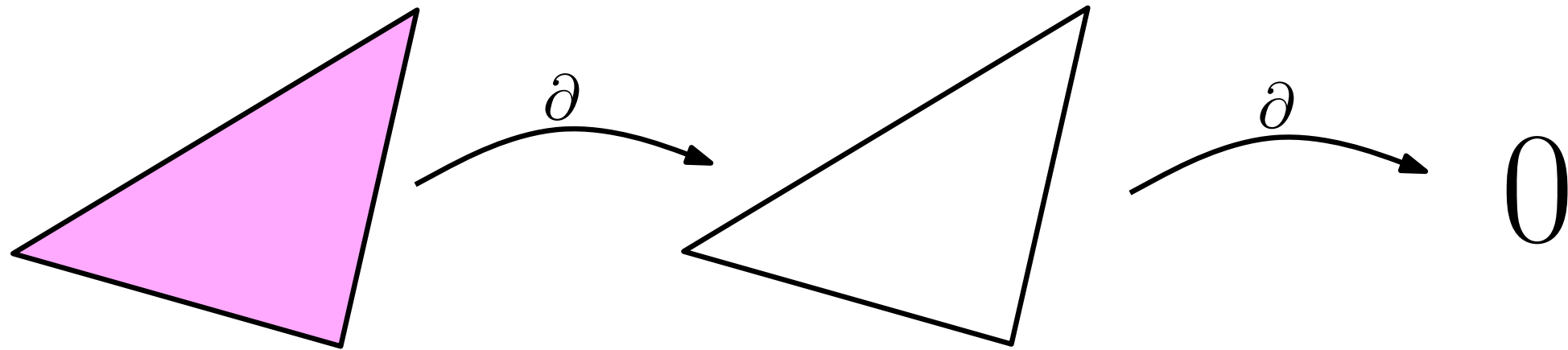
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Geometric interpretation:

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sum  $c + c' =$  symmetric difference

# The boundary operator



The **boundary**  $\partial\sigma$  of a  $k$ -simplex  $\sigma$  is the sum of its  $(k - 1)$ -faces. This is a  $(k - 1)$ -chain.

$$\text{If } \sigma = [v_0, \dots, v_k] \text{ then } \partial\sigma = \sum_{i=0}^k [v_0 \cdots \hat{v}_i \cdots v_k]$$

The boundary operator is the linear map defined by

$$\begin{aligned} \partial : \mathcal{C}_k(K) &\rightarrow \mathcal{C}_{k-1}(K) \\ c &\rightarrow \partial c = \sum_{\sigma \in c} \partial\sigma \end{aligned}$$

# Fundamental property of the boundary operator

$$\partial\partial := \partial \circ \partial = 0$$

**Proof:** by linearity it is just necessary to prove it for a simplex.

$$\begin{aligned}\partial\partial\sigma &= \partial \left( \sum_{i=0}^k [v_0 \cdots \hat{v}_i \cdots v_k] \right) \\ &= \sum_{i=0}^k \partial[v_0 \cdots \hat{v}_i \cdots v_k] \\ &= \sum_{j < i} [v_0 \cdots \hat{v}_j \cdots \hat{v}_i \cdots v_k] + \sum_{j > i} [v_0 \cdots \hat{v}_i \cdots \hat{v}_j \cdots v_k] \\ &= 0\end{aligned}$$

# Cycles and boundaries

The **chain complex** associated to a complex  $K$  of dimension  $d$

$$\emptyset \rightarrow \mathcal{C}_d(K) \xrightarrow{\partial} \mathcal{C}_{d-1}(K) \xrightarrow{\partial} \cdots \mathcal{C}_{k+1}(K) \xrightarrow{\partial} \mathcal{C}_k(K) \xrightarrow{\partial} \cdots \mathcal{C}_1(K) \xrightarrow{\partial} \mathcal{C}_0(K) \xrightarrow{\partial} \emptyset$$

**$k$ -cycles:**

$$Z_k(K) := \ker(\partial : \mathcal{C}_k \rightarrow \mathcal{C}_{k-1}) = \{c \in \mathcal{C}_k : \partial c = \emptyset\}$$

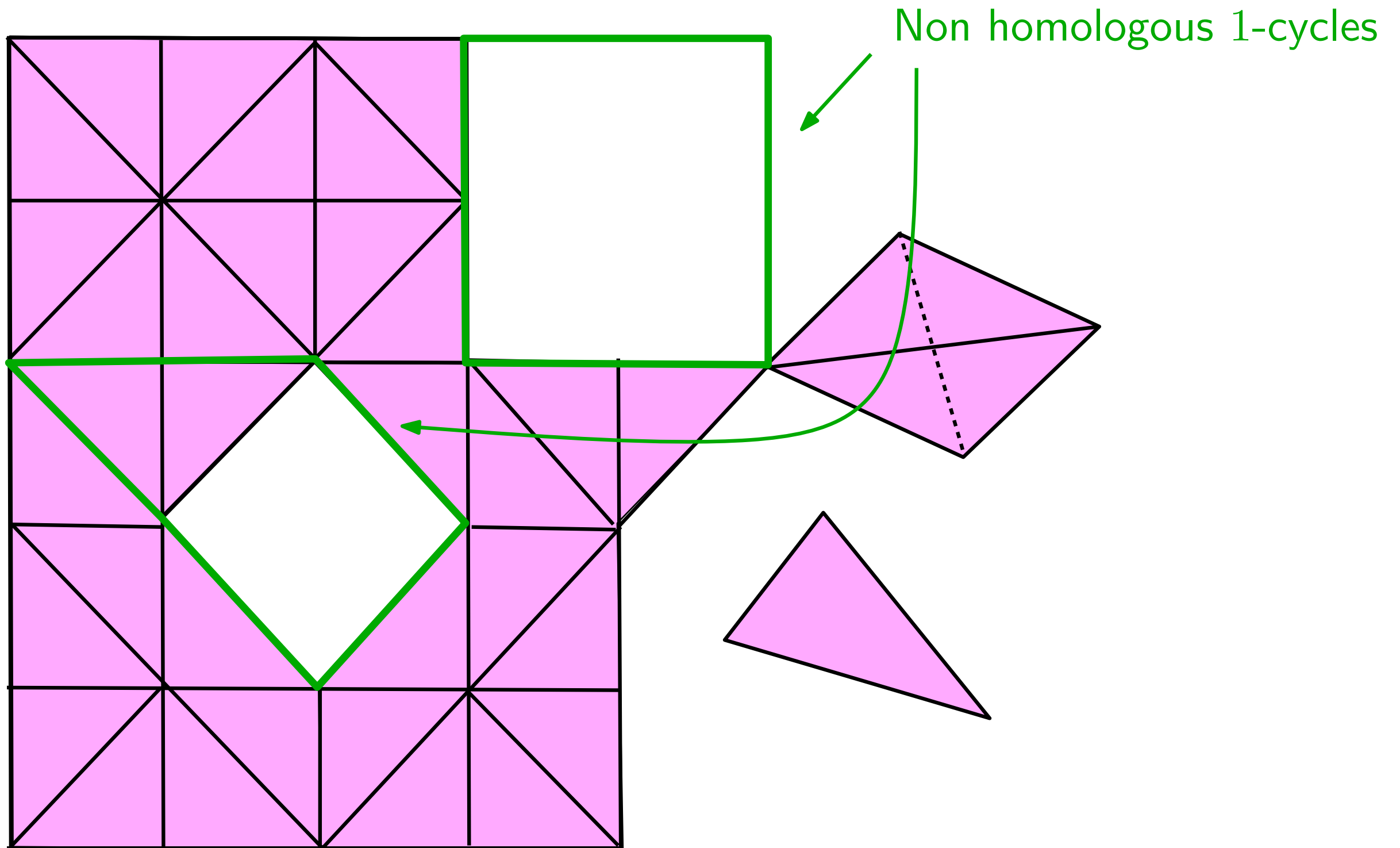
**$k$ -boundaries:**

$$B_k(K) := \operatorname{im}(\partial : \mathcal{C}_{k+1} \rightarrow \mathcal{C}_k) = \{c \in \mathcal{C}_k : \exists c' \in \mathcal{C}_{k+1}, c = \partial c'\}$$

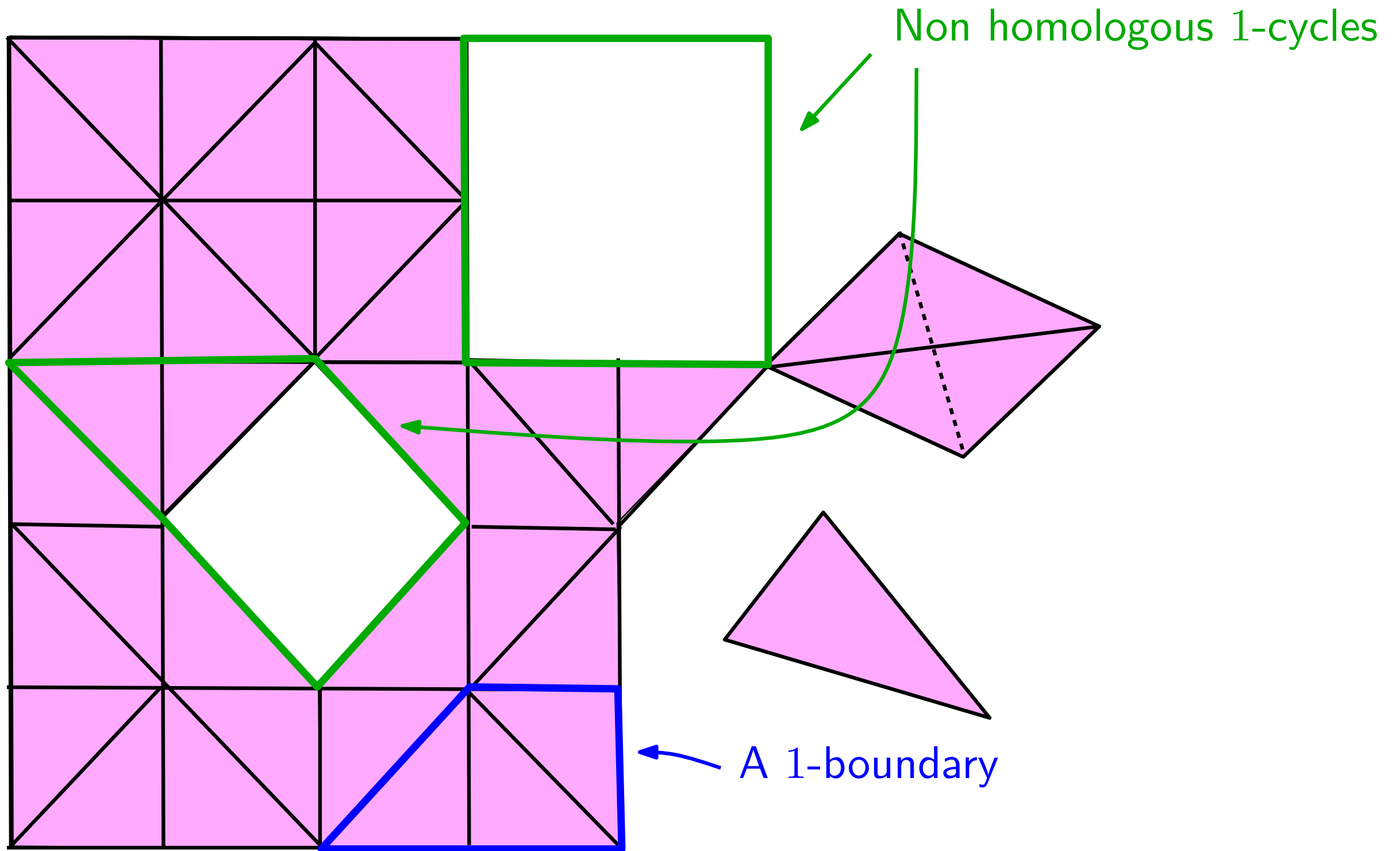
$$B_k(K) \subset Z_k(K) \subset \mathcal{C}_k(K)$$



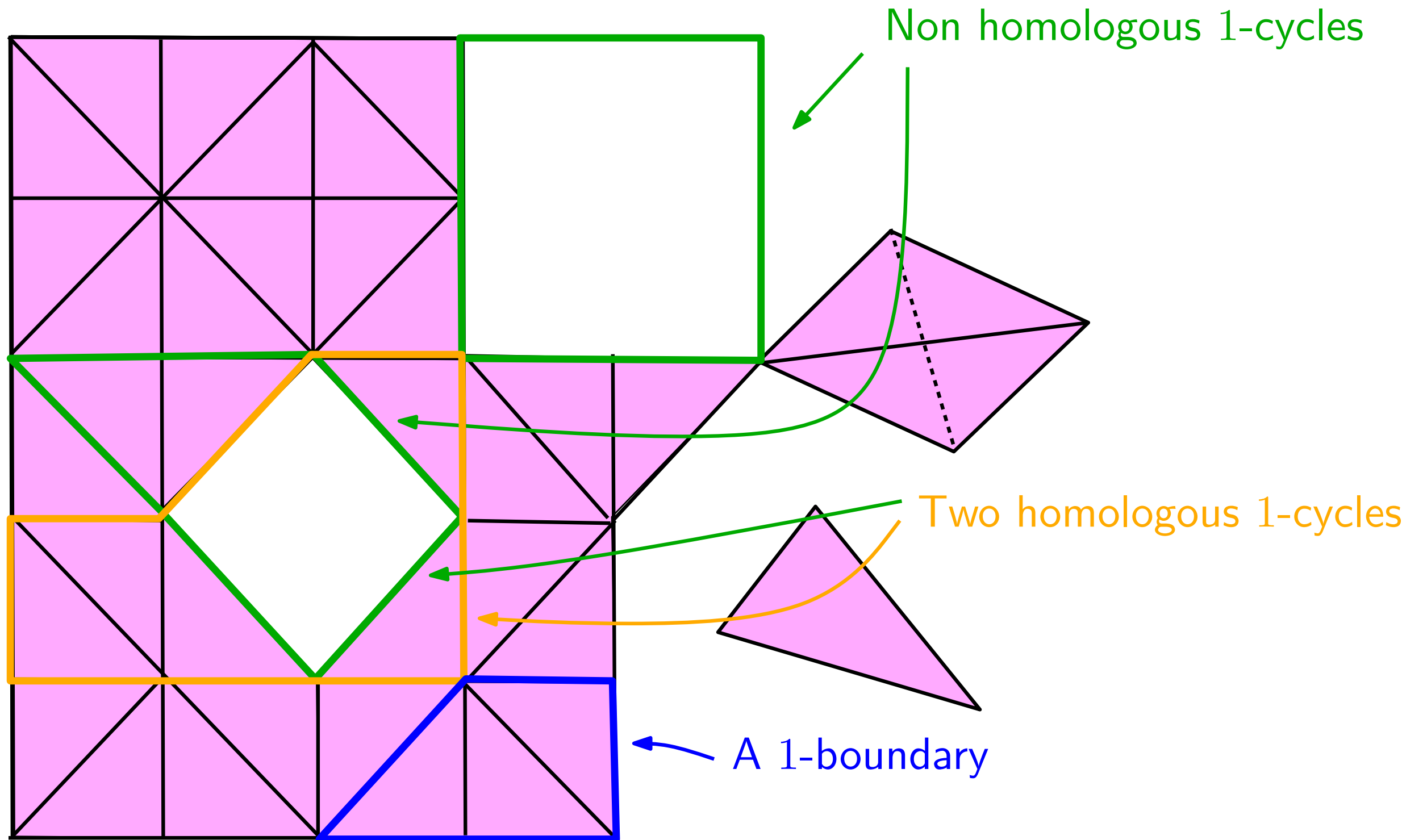
# Cycles and boundaries



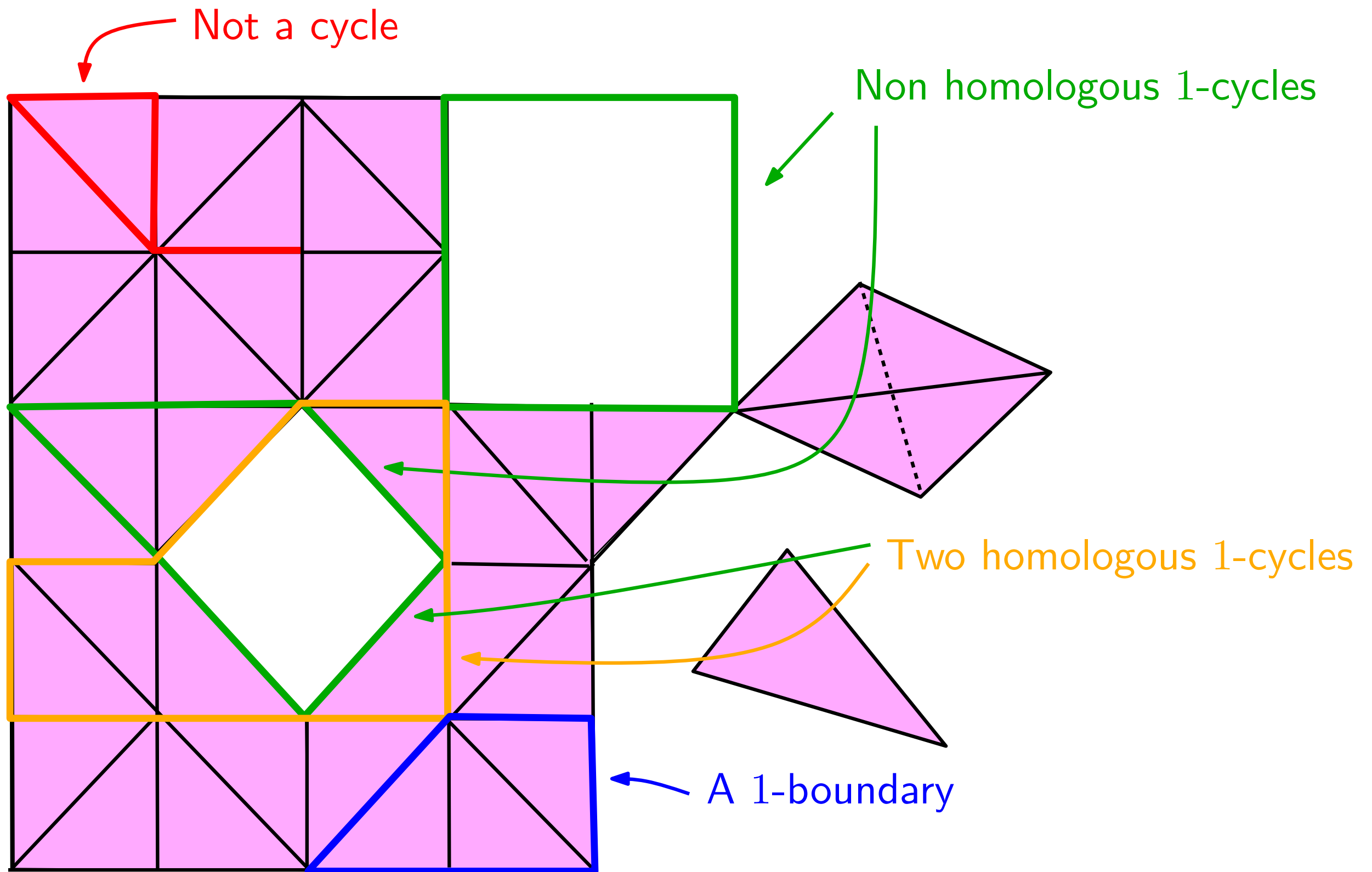
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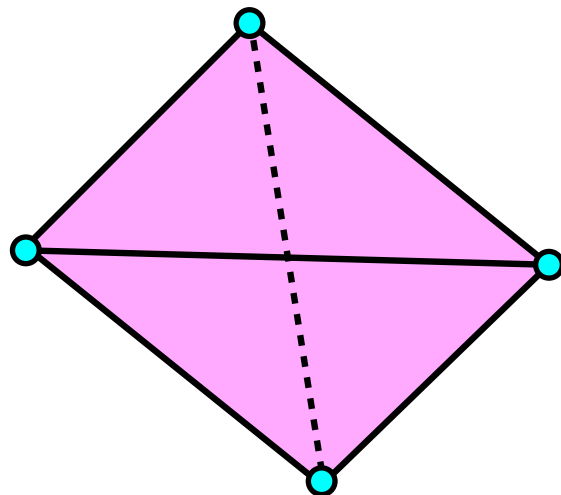
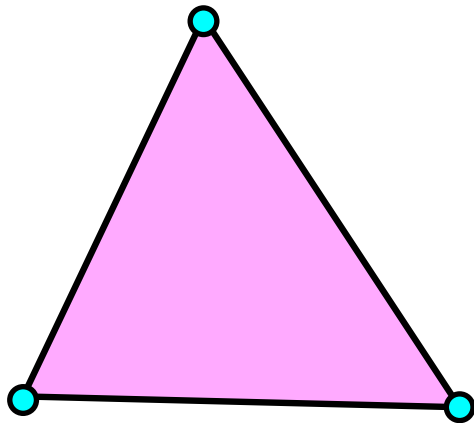
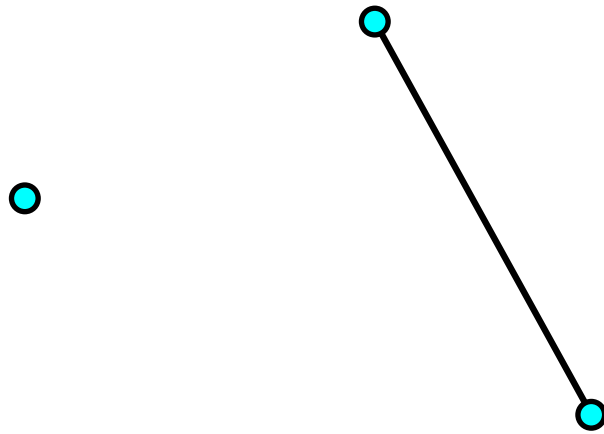
# Homology groups and Betti numbers

$$B_k(K) \subset Z_k(K) \subset \mathcal{C}_k(K)$$

- The  $k^{th}$  **homology group** of  $K$ :  $H_k(K) = Z_k / B_k$
- Tout each cycle  $c \in Z_k(K)$  corresponds its **homology class**  $c + B_k(K) = \{c + b : b \in B_k(K)\}$ .
- Two cycles  $c, c'$  are **homologous** if they are in the same homology class:  $\exists b \in B_k(K)$  s. t.  $b = c' - c (= c' + c)$ .
- The  $k^{th}$  **Betti number** of  $K$ :  $\beta_k(K) = \dim(H_k(K))$ .

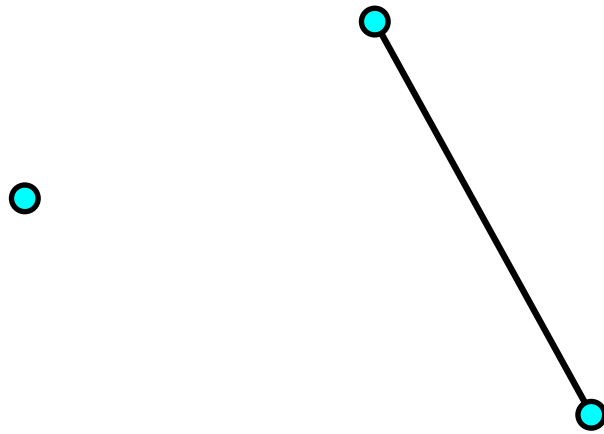
# Elementary examples

**Remark:**  $\beta_0 =$  number of connected components of  $K$



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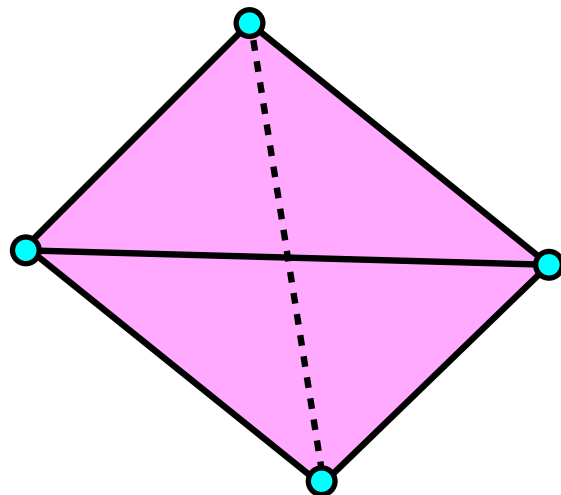
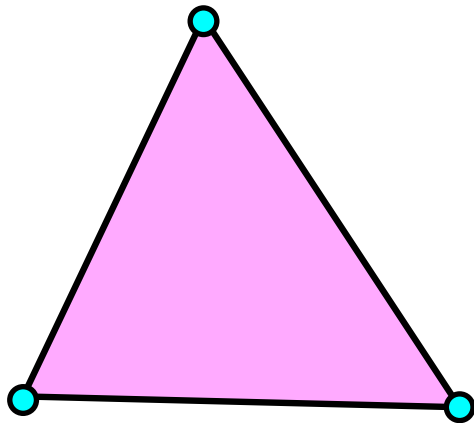
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$$\beta_0 = 2$$

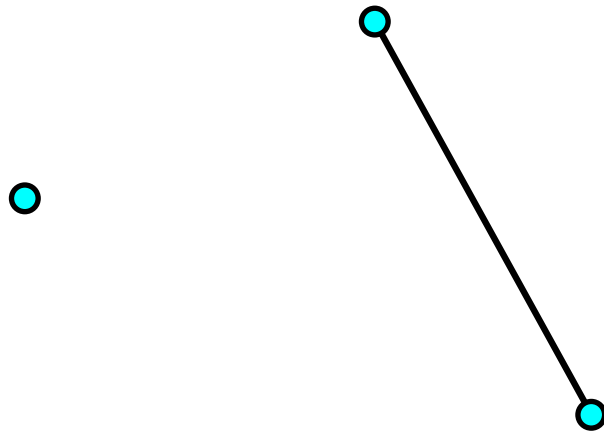
$$\beta_1 = 0$$

$$\beta_2 = 0$$



# Elementary examples

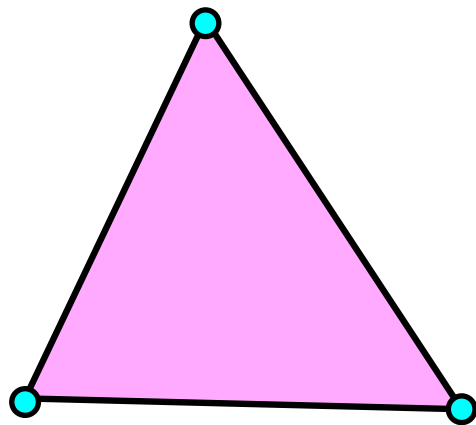
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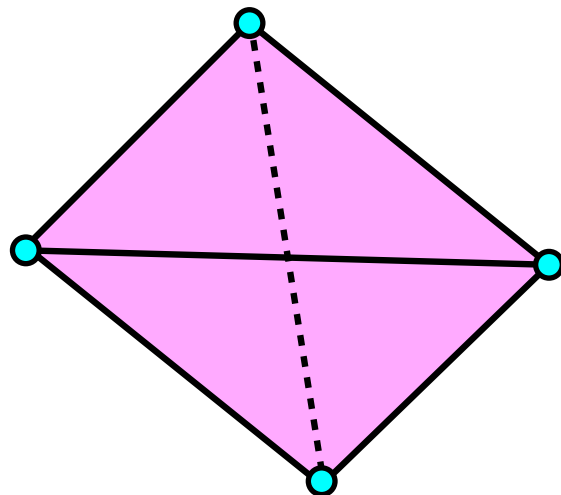
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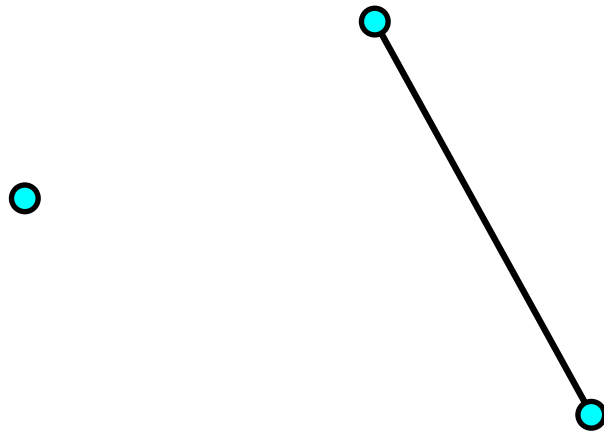
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# Elementary examples

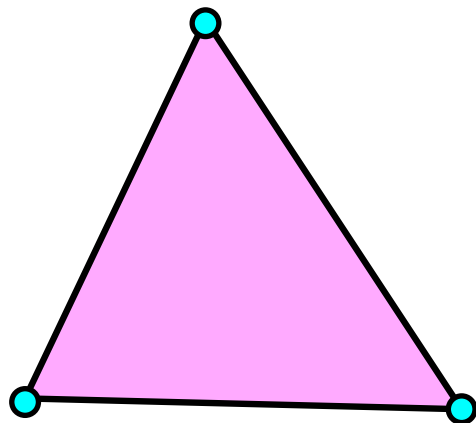
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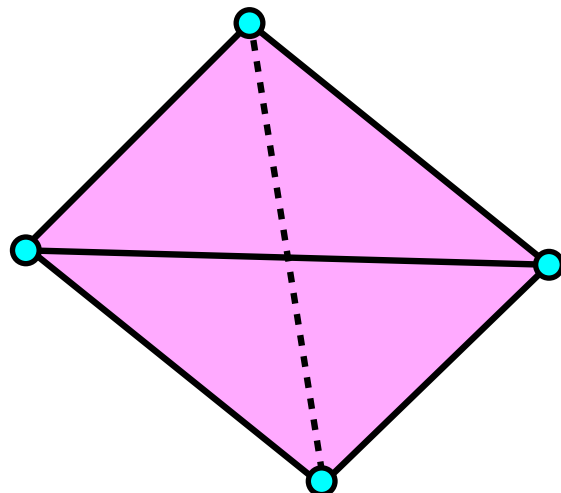
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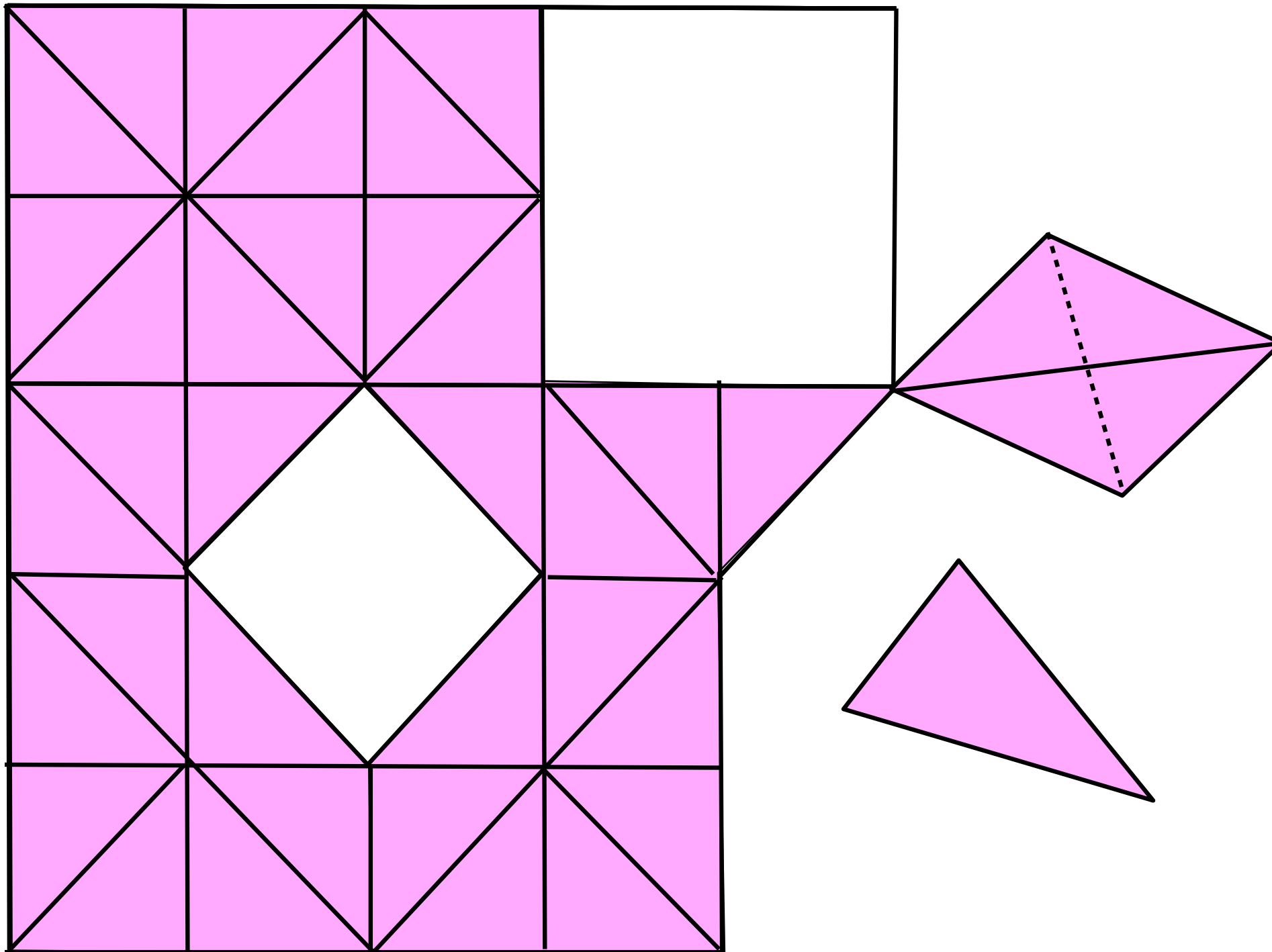
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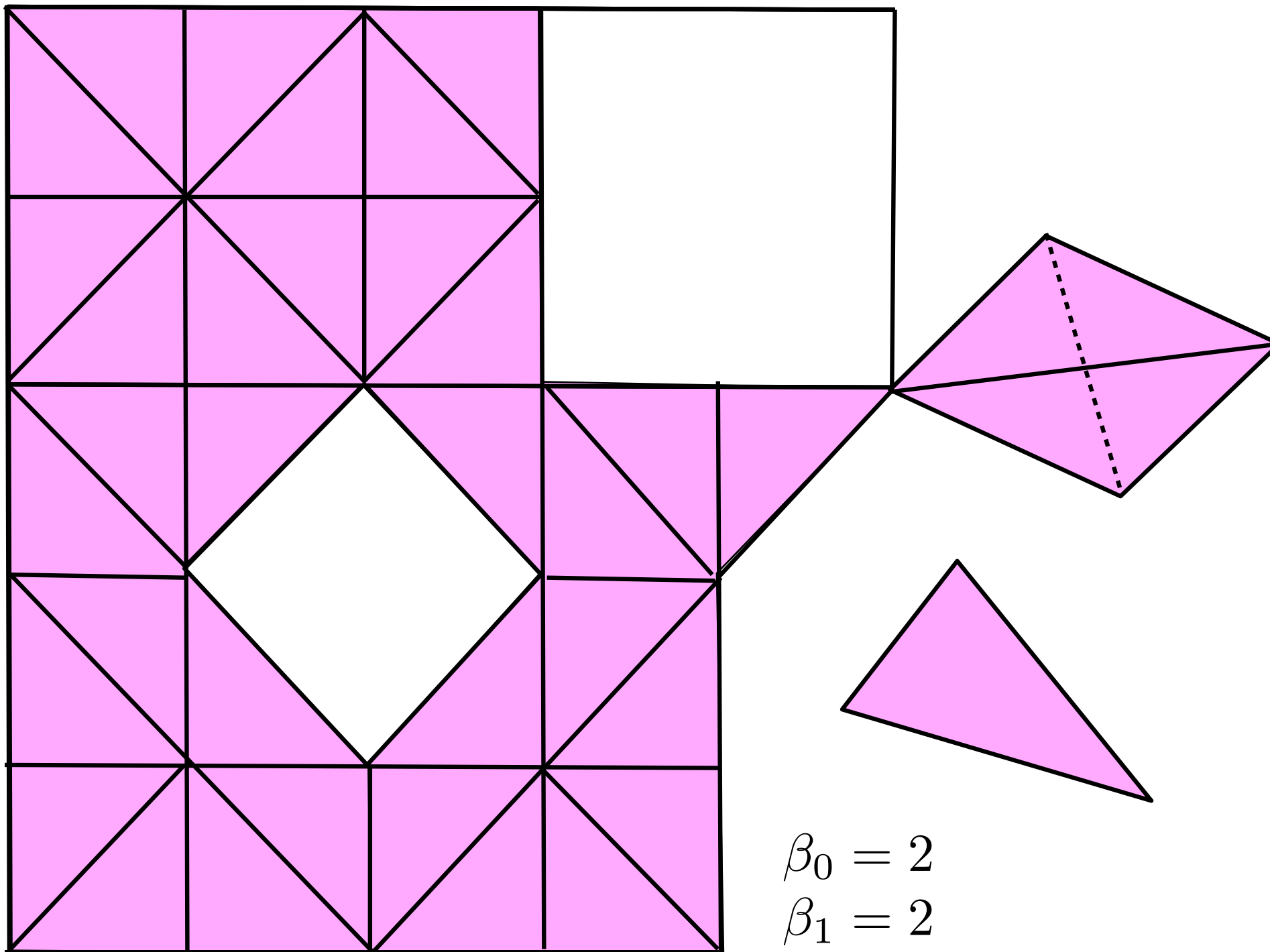
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# Elementary examples



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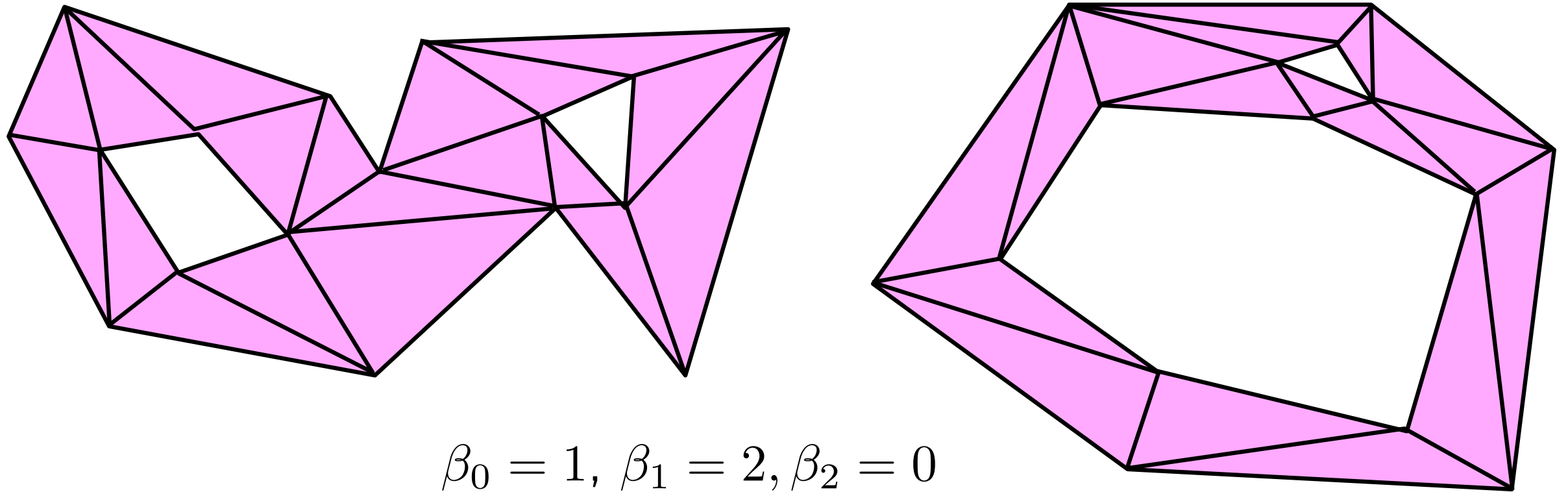
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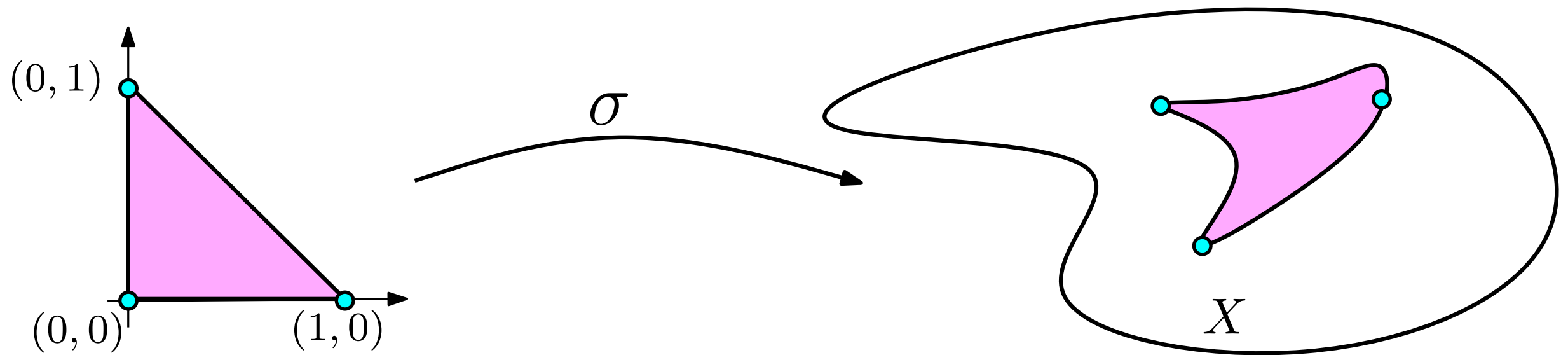
# Topological invariance and singular homology



**Theorem:** If  $K$  and  $K'$  are two simplicial complexes with homeomorphic supports then their homology groups are isomorphic and their Betti numbers are equal.

- This is a classical result in algebraic topology but the proof is not obvious.
- Rely on the notion of singular homology  $\rightarrow$  defined for any topological space.

# Topological invariance and singular homology



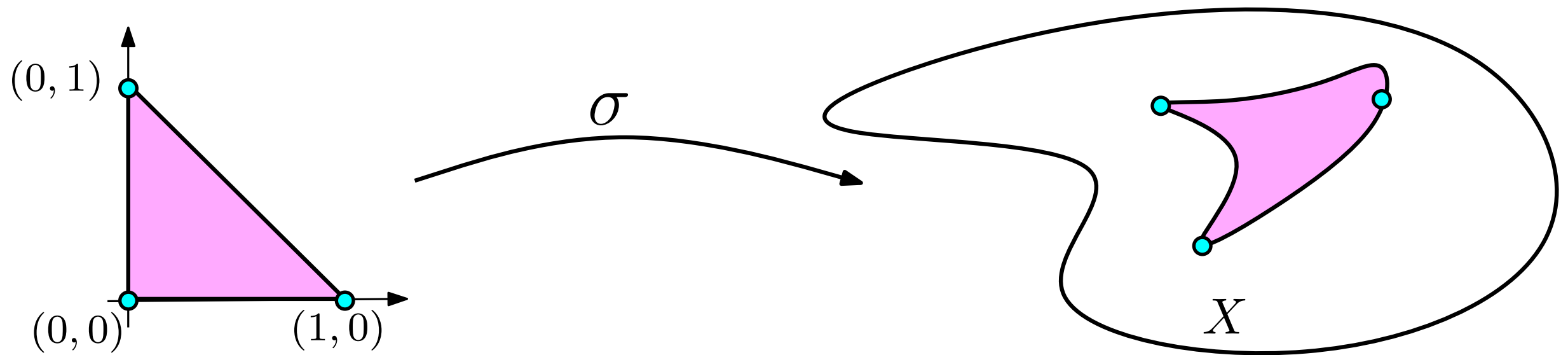
Let  $\Delta_k$  be the standard simplex in  $\mathbb{R}^k$ . A singular  $k$ -simplex in a topological space  $X$  is a continuous map  $\sigma : \Delta_k \rightarrow X$ .

The same construction as for simplicial homology can be done with singular complexes  $\rightarrow$  **Singular homology**

Important properties:

- Singular homology is defined for any topological space  $X$ .
- If  $X$  is homotopy equivalent to the support of a simplicial complex, then the singular and simplicial homology coincide!

# Topological invariance and singular homology



Let  $\Delta_k$  be the standard simplex in  $\mathbb{R}^k$ . A singular  $k$ -simplex in a topological space  $X$  is a continuous map  $\sigma : \Delta_k \rightarrow X$ .

Homology and continuous maps:

- if  $f : X \rightarrow Y$  is a continuous map and  $\sigma : \Delta_k \rightarrow X$  a simplex in  $X$ , then  $f \circ \sigma : \Delta_k \rightarrow Y$  is a simplex in  $Y \Rightarrow f$  induces a linear maps between homology groups:

$$f_{\#} : H_k(X) \rightarrow H_k(Y)$$

- if  $f : X \rightarrow Y$  is an homeomorphism or an homotopy equivalence then  $f_{\#}$  is an isomorphism.

# An algorithm for geometric inference

- $X \subset \mathbb{R}^d$  be a compact set such that  $\text{wfs}(X) > 0$ .
- $L \subset \mathbb{R}^d$  be a finite set such that  $d_H(X, L) < \varepsilon$  for some  $\varepsilon > 0$ .

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**Goal:** Compute the Betti numbers of  $X^r$  for  $0 < r < \text{wfs}(X)$  from  $L$ .



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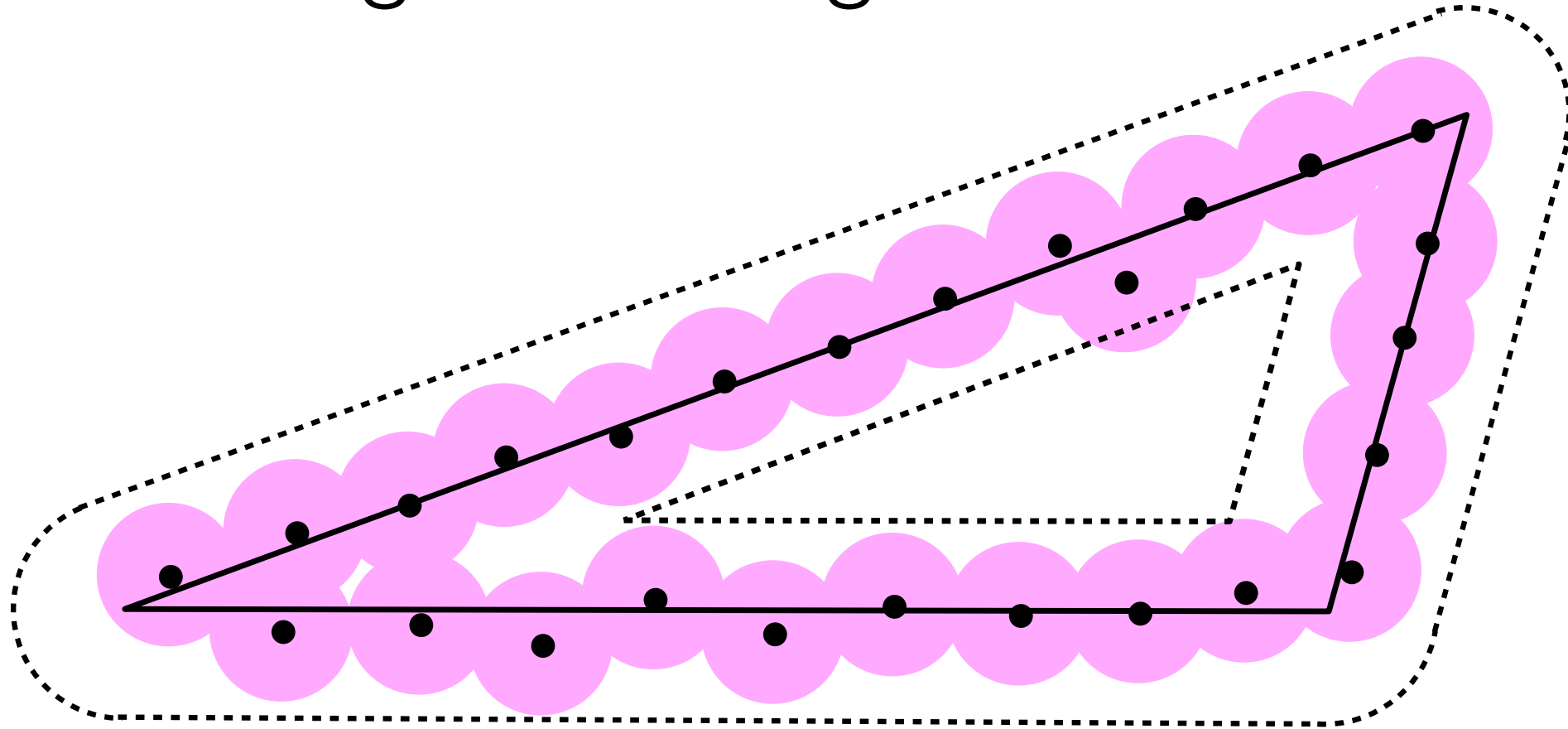
**Theorem:** [CL'05 - CSEH'05]

Assume that  $\text{wfs}(X) > 4\varepsilon$ . For  $\alpha > 0$  s.t.  $\alpha + 4\varepsilon < \text{wfs}(X)$ , let  $i : L^{\alpha+\varepsilon} \hookrightarrow L^{\alpha+3\varepsilon}$  be the canonical inclusion. For any  $0 < r < \text{wfs}(X)$ ,

$$H_k(X^r) \cong \text{im} \left( i_* : H_k(L^{\alpha+\varepsilon}) \rightarrow H_k(L^{\alpha+3\varepsilon}) \right)$$

# An algorithm for geometric inference

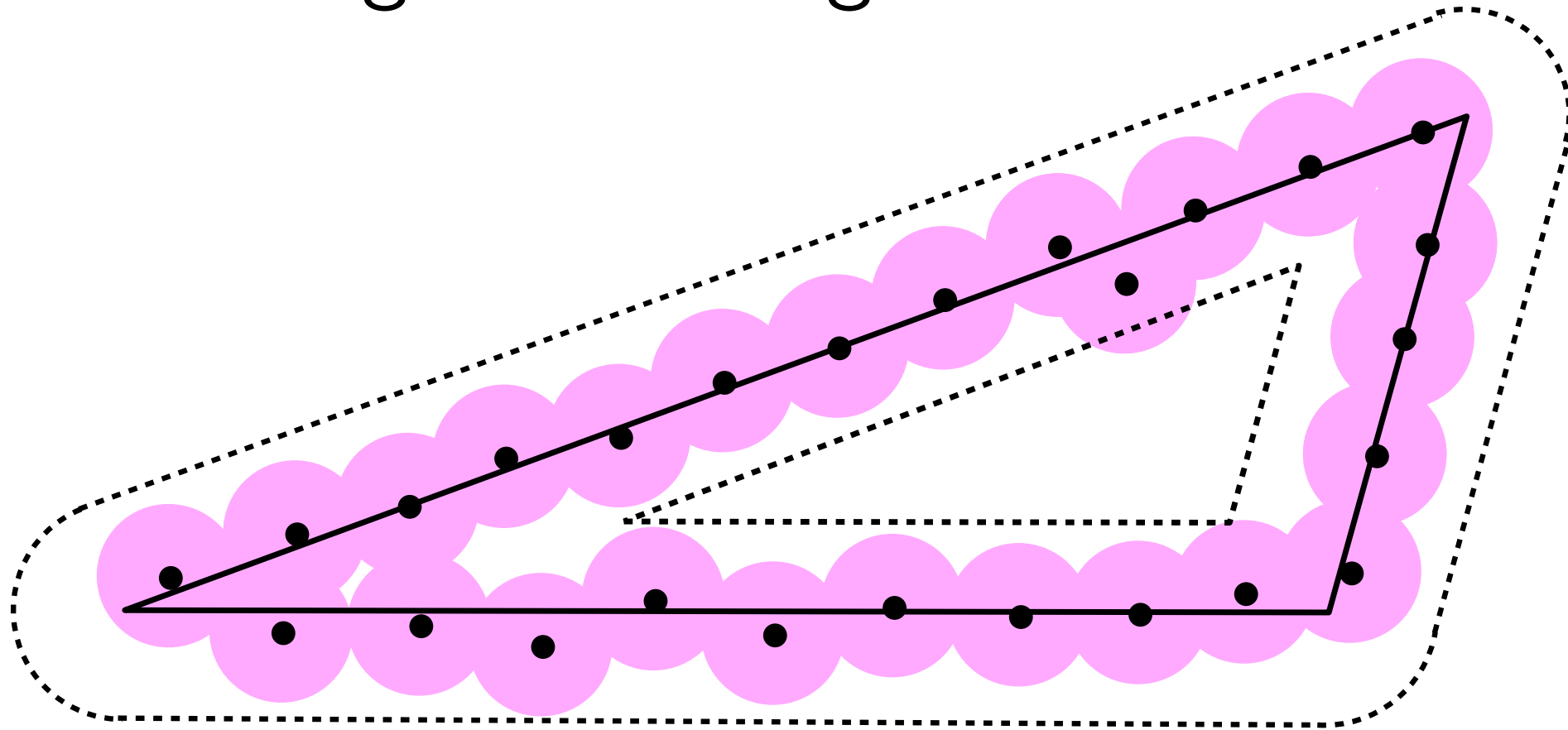
**Proof:**



For any  $\alpha > 0$ ,  $X^\alpha \subseteq L^{\alpha+\varepsilon} \subseteq X^{\alpha+2\varepsilon} \subseteq L^{\alpha+3\varepsilon} \subseteq X^{\alpha+4\varepsilon} \subseteq \dots$

# An algorithm for geometric inference

**Proof:**



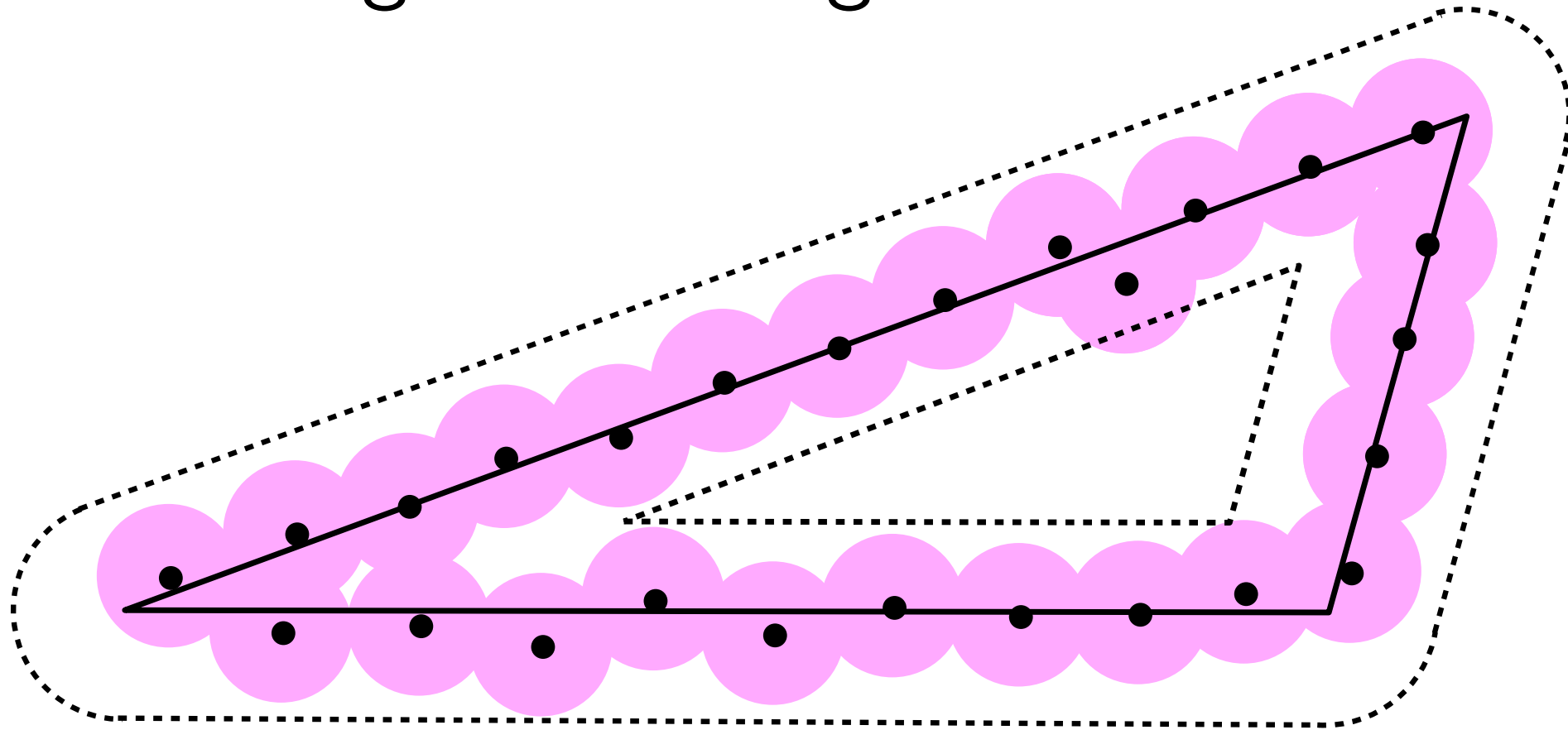
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At homology level:

$$H_k(X^\alpha) \rightarrow H_k(L^{\alpha+\varepsilon}) \rightarrow H_k(X^{\alpha+2\varepsilon}) \rightarrow H_k(L^{\alpha+3\varepsilon}) \rightarrow H_k(X^{\alpha+4\varepsilon}) \rightarrow \dots$$

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**Proof:**



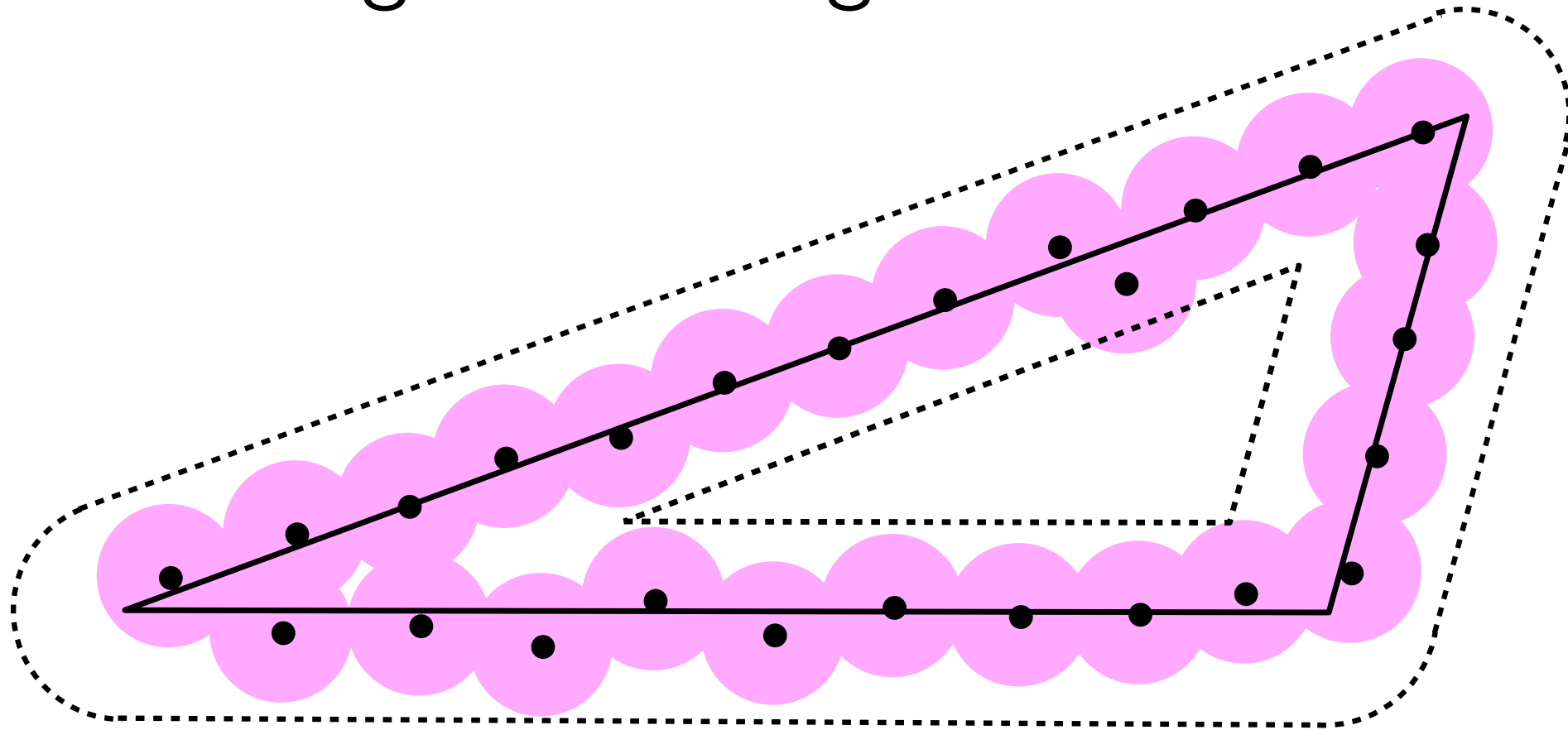
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At homology level:

$$\begin{array}{c}
 \text{rank} = \dim H_k(X^\alpha) \\
 \xrightarrow{\hspace{10em}} \\
 H_k(X^\alpha) \rightarrow H_k(L^{\alpha+\varepsilon}) \rightarrow H_k(X^{\alpha+2\varepsilon}) \rightarrow H_k(L^{\alpha+3\varepsilon}) \rightarrow H_k(X^{\alpha+4\varepsilon}) \rightarrow \dots \\
 \xrightarrow{\text{isomorphism}} \hspace{10em} \xrightarrow{\text{isomorphism}}
 \end{array}$$

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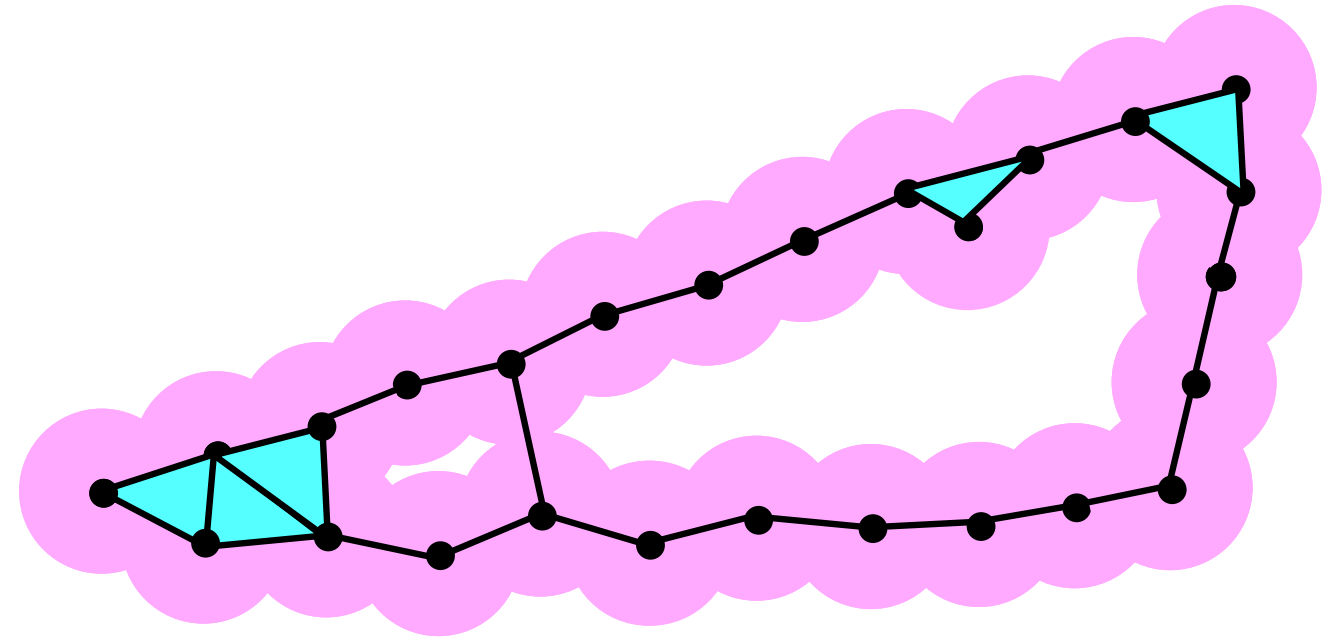
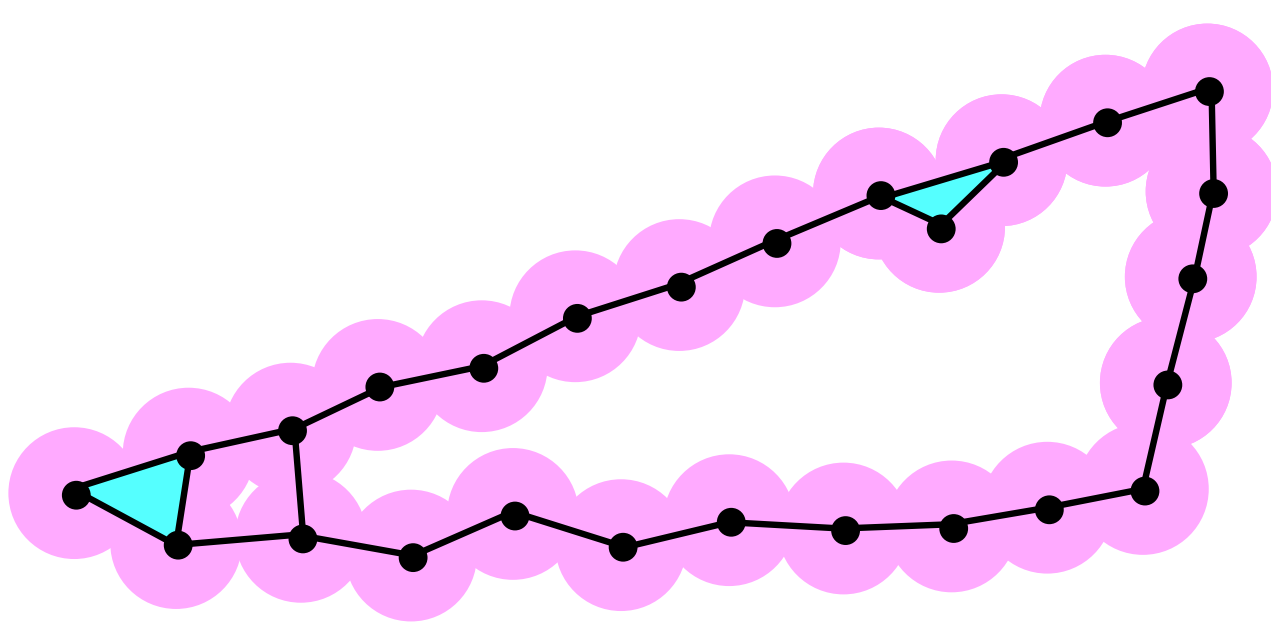
Cannot be directly computed !

$$H_k(X^\alpha) \rightarrow H_k(L^{\alpha+\varepsilon}) \rightarrow H_k(X^{\alpha+2\varepsilon}) \rightarrow H_k(L^{\alpha+3\varepsilon}) \rightarrow H_k(X^{\alpha+4\varepsilon}) \rightarrow \dots$$

isomorphism

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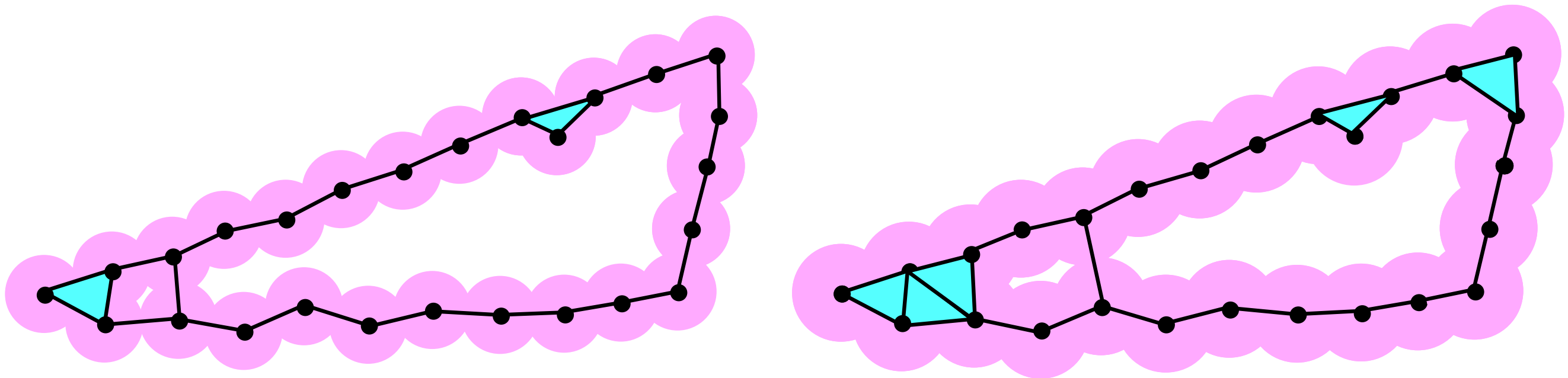
# Using the Čech complex



The Čech complex  $\mathcal{C}^\alpha(L)$ :

for  $p_0, \dots, p_k \in L$ ,  $\sigma = [p_0 p_1 \dots p_k] \in \mathcal{C}^\alpha(L)$  iff  $\bigcap_{i=0}^k B(p_i, \alpha) \neq \emptyset$

# Using the Čech complex

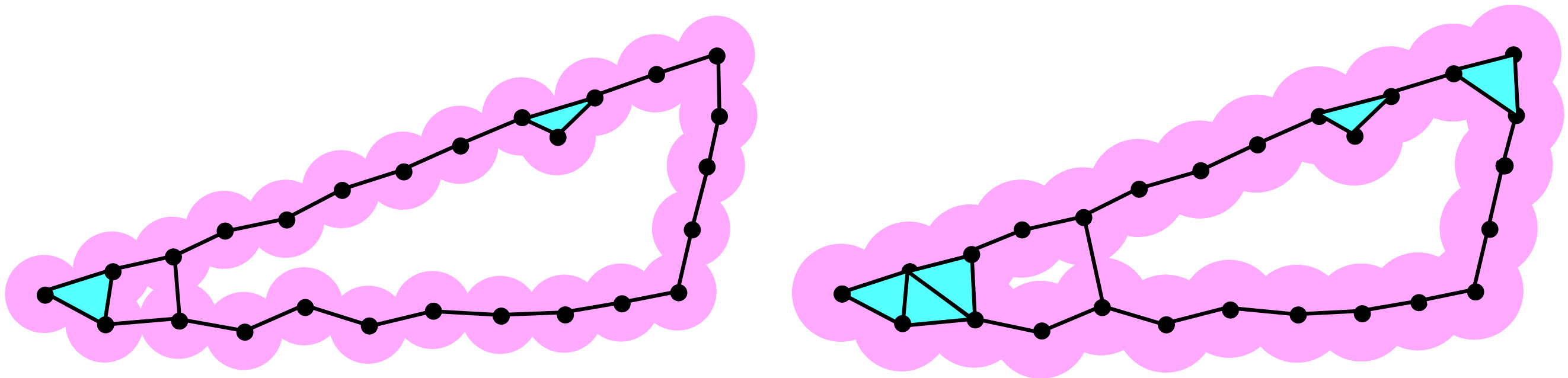


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**Nerve theorem:** For any  $\alpha > 0$ ,  $L^\alpha$  and  $\mathcal{C}^\alpha(L)$  are homotopy equivalent and the homotopy equivalences can be chosen to commute with inclusions.

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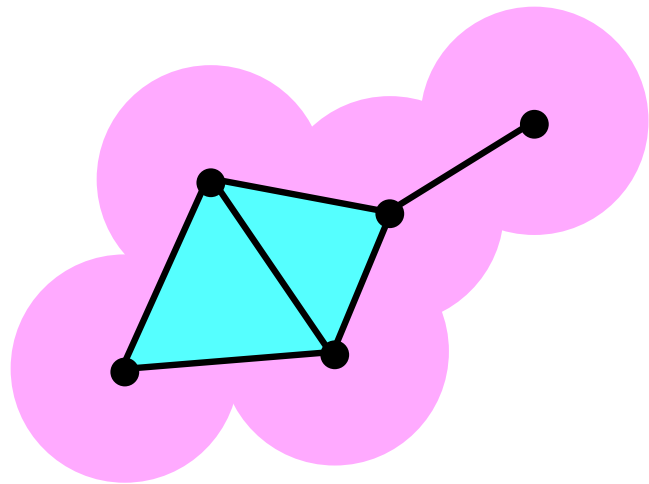
**Nerve theorem:** For any  $\alpha > 0$ ,  $L^\alpha$  and  $\mathcal{C}^\alpha(L)$  are homotopy equivalent and the homotopy equivalences can be chosen to commute with inclusions.

$$\begin{array}{ccccccc} \dots & \rightarrow & H_k(L^{\alpha+\varepsilon}) & \rightarrow & H_k(L^{\alpha+3\varepsilon}) & \rightarrow & \dots \\ & & \downarrow & & \downarrow & & \\ \dots & \rightarrow & H_k(\mathcal{C}^{\alpha+\varepsilon}(L)) & \rightarrow & H_k(\mathcal{C}^{\alpha+3\varepsilon}(L)) & \rightarrow & \dots \end{array}$$

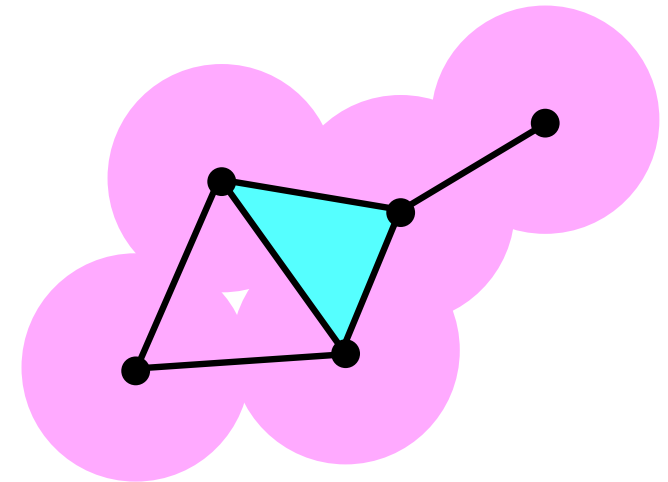
Allow to work with simplicial complexes but... still too difficult to compute



# Using the Rips complex



Rips vs Čech



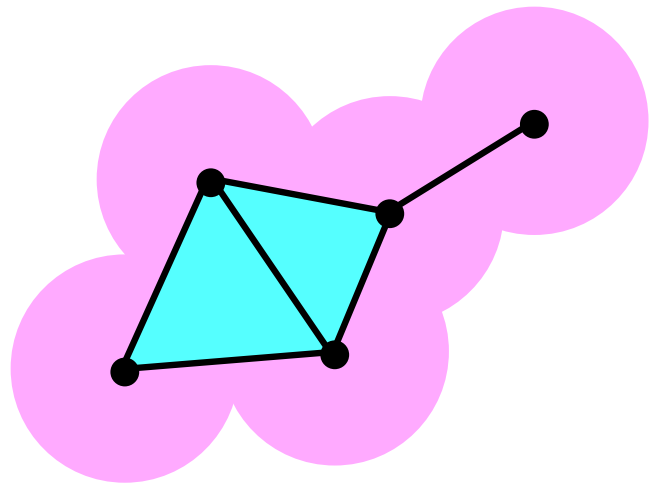
The Rips complex  $\mathcal{R}^\alpha(L)$ : for  $p_0, \dots, p_k \in L$ ,

$$\sigma = [p_0 p_1 \dots p_k] \in \mathcal{R}^\alpha(L) \text{ iff } \forall i, j \in \{0, \dots, k\}, d(p_i, p_j) \leq \alpha$$

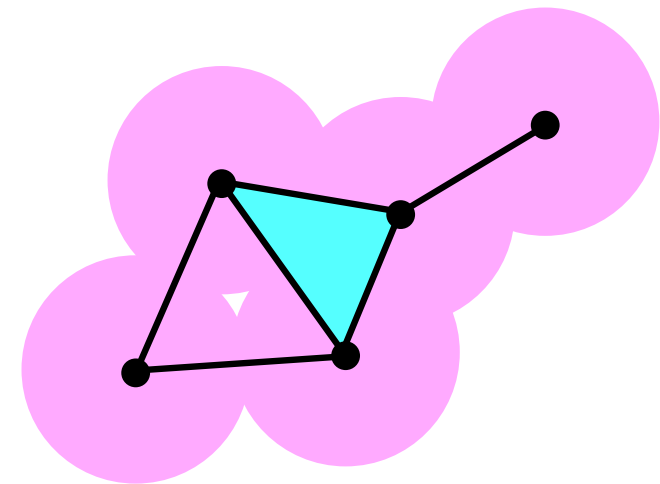
- Easy to compute and fully determined by its 1-skeleton
- Rips-Čech interleaving: for any  $\alpha > 0$ ,

$$\mathcal{C}^{\frac{\alpha}{2}}(L) \subseteq \mathcal{R}^\alpha(L) \subseteq \mathcal{C}^\alpha(L) \subseteq \mathcal{R}^{2\alpha}(L) \subseteq \dots$$

# Using the Rips complex



Rips vs Čech



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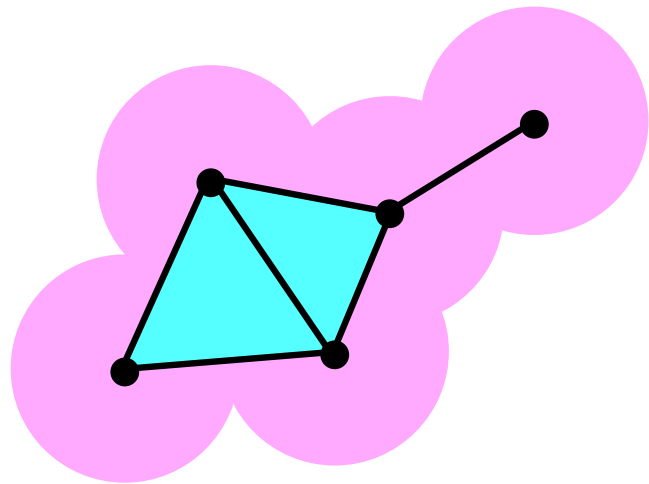
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**Theorem:** [C-Oudot'08]

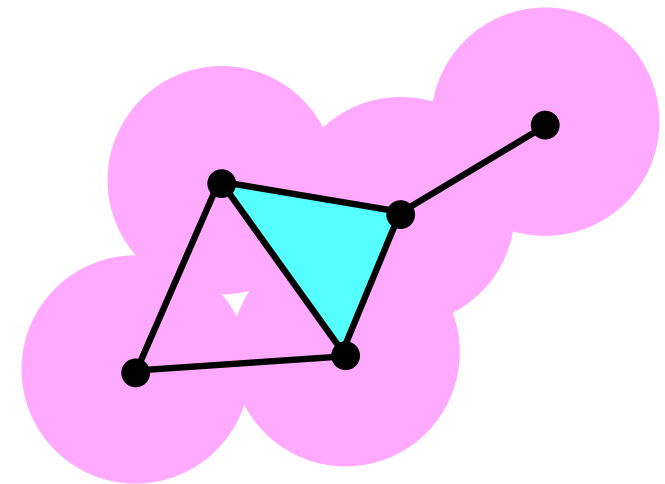
Let  $X \subset \mathbb{R}^d$  be a compact set and  $L \subset \mathbb{R}^d$  a finite set such that  $d_H(X, L) < \varepsilon$  for some  $\varepsilon < \frac{1}{9} \text{wfs}(X)$ . Then for all  $\alpha \in [2\varepsilon, \frac{1}{4}(\text{wfs}(X) - \varepsilon)]$  and all  $\lambda \in (0, \text{wfs}(X))$ , one has:  $\forall k \in \mathbb{N}$

$$\beta_k(X^\lambda) = \dim(H_k(X^\lambda)) = \text{rk}(\mathcal{R}^\alpha(L) \rightarrow \mathcal{R}^{4\alpha}(L))$$

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Rips vs Čech



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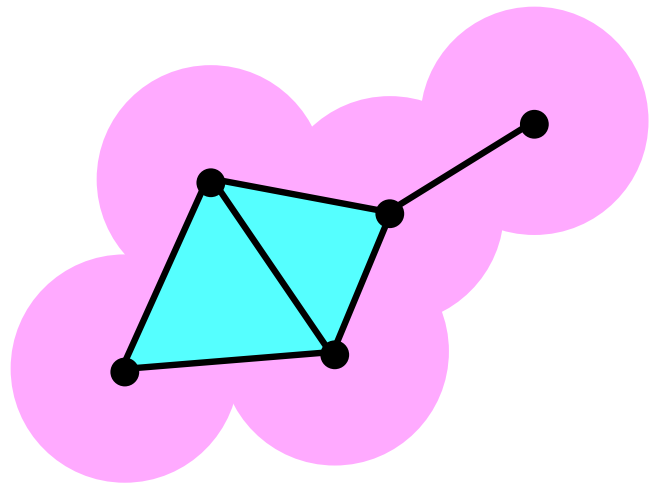
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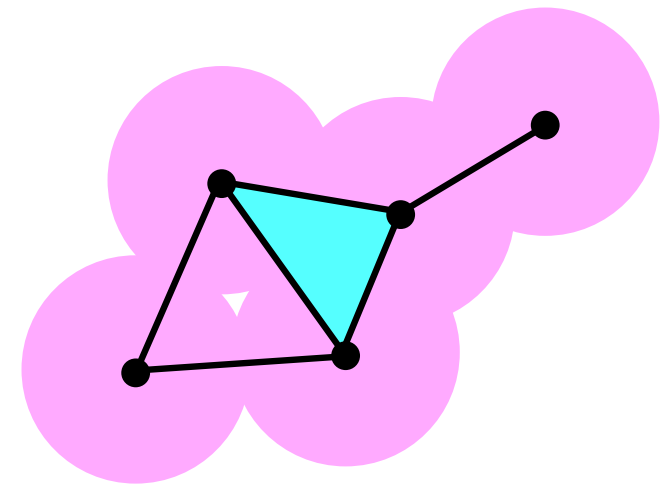
$$\beta_k(X^\lambda) = \dim(H_k(X^\lambda)) = \text{rk}(\mathcal{R}^\alpha(L) \rightarrow \mathcal{R}^{4\alpha}(L))$$

Easy to compute using persistence algo.

# Using the Rips complex



Rips vs Čech



The Rips complex  $\mathcal{R}^\alpha(L)$ : for  $p_0, \dots, p_k \in L$ ,

$$\sigma = [p_0 p_1 \cdots p_k] \in \mathcal{R}^\alpha(L) \text{ iff } \forall i, j \in \{0, \dots, k\}, d(p_i, p_j) \leq \alpha$$

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$$\beta_k(X^\lambda) = \dim(H_k(X^\lambda)) = \text{rk}(\mathcal{R}^\alpha(L) \rightarrow \mathcal{R}^{4\alpha}(L))$$

➡ **Pb:** Choice of  $\alpha$  when  $\text{wfs}(X)$  is unknown?

# Multiscale inference

**Input:** A point cloud  $W$  and its pairwise distances  $\{d(w, w')\}_{w, w' \in W}$ .  
→ Maintain a nested pair  $\mathcal{R}^{4\varepsilon}(L) \hookrightarrow \mathcal{R}^{16\varepsilon}(L)$  where  $L = L(\varepsilon)$ .

Init.:  $L = \emptyset$ ;  $\varepsilon = +\infty$

WHILE  $L \subset W$

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

update  $\varepsilon = \max_{w \in W} d(w, L)$

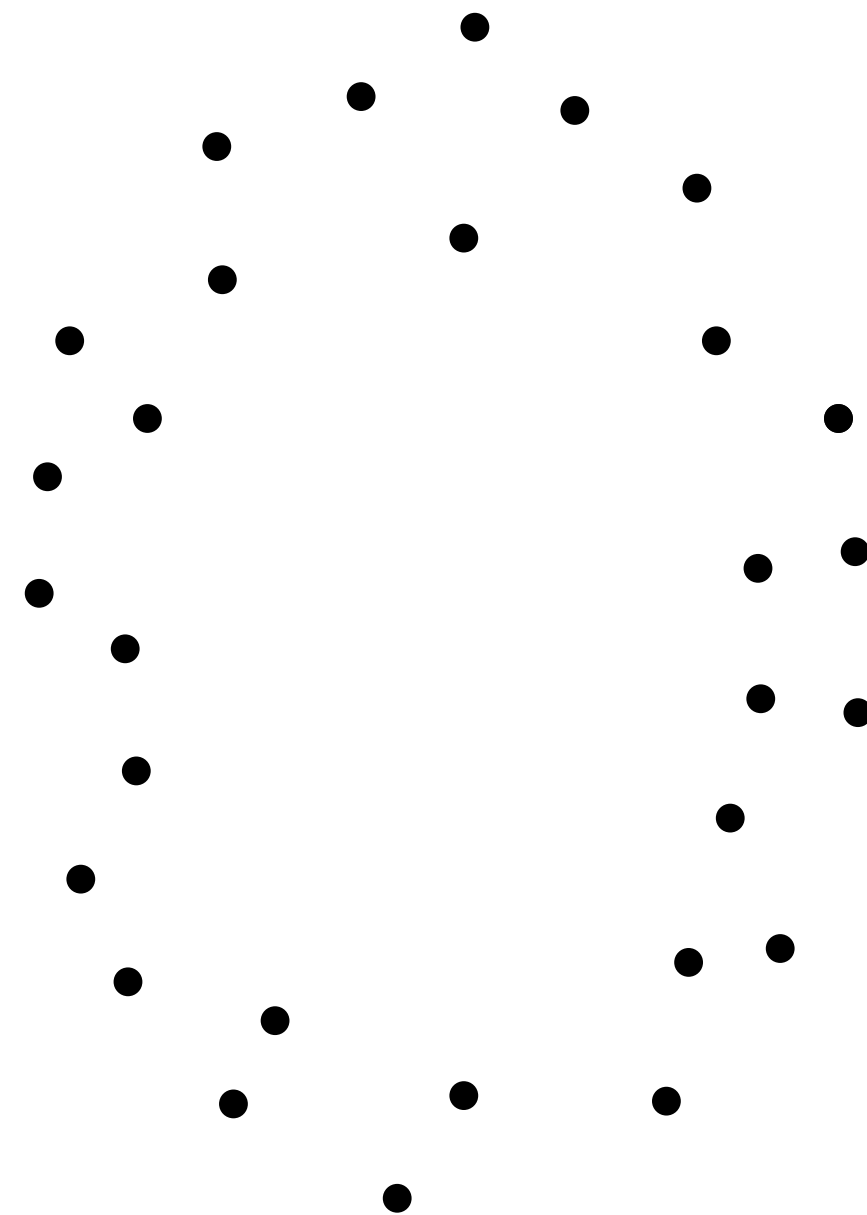
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Persistence( $\mathcal{R}^{4\varepsilon}(L) \hookrightarrow \mathcal{R}^{16\varepsilon}(L)$ )

END\_WHILE

Output: Sequence of persistent Betti numbers  
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Rank of the map induced at homology level



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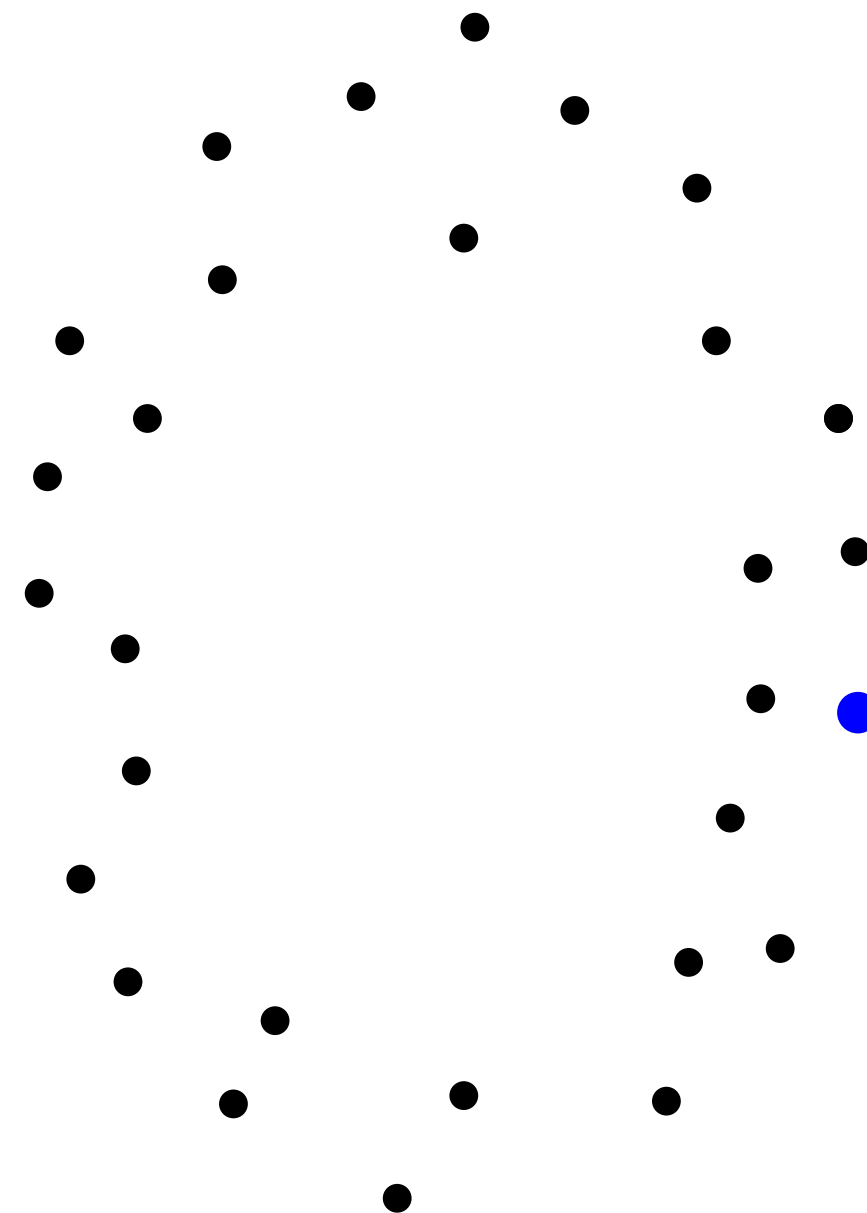
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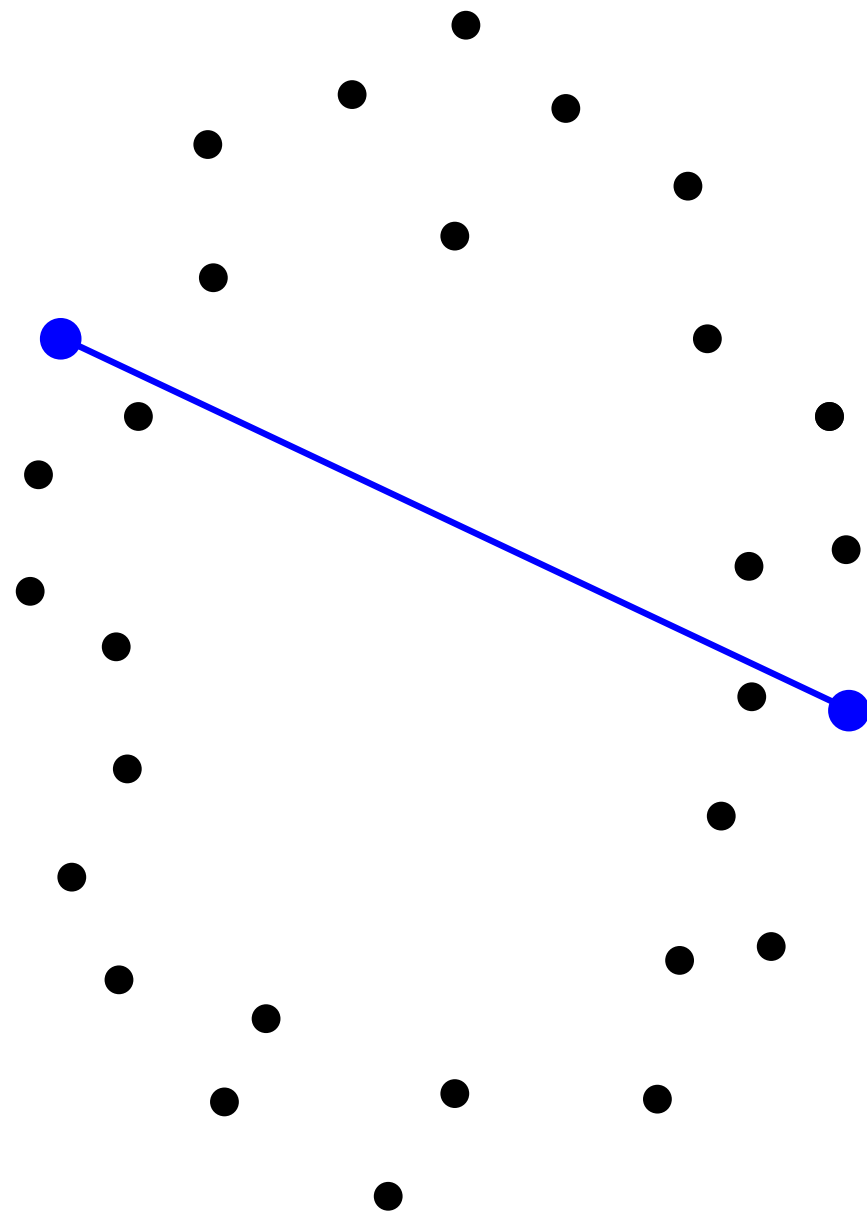
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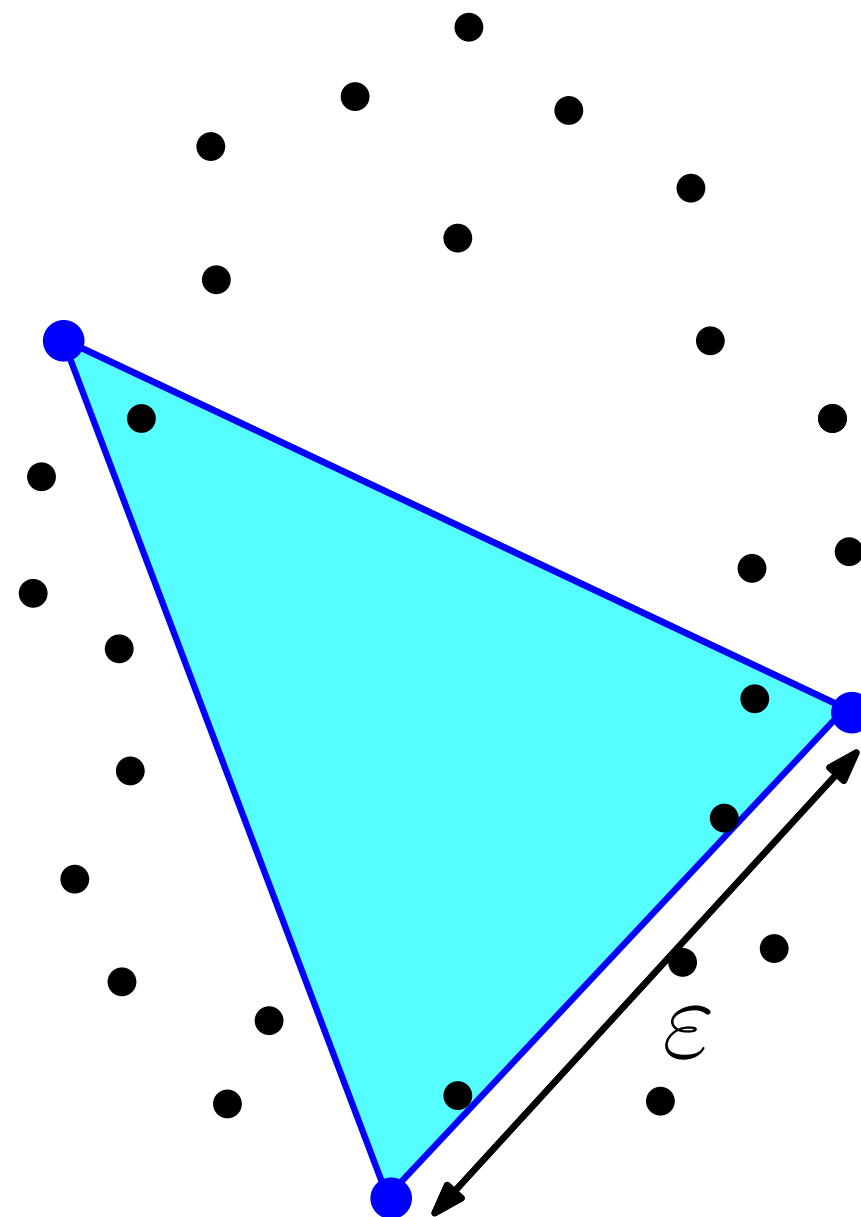
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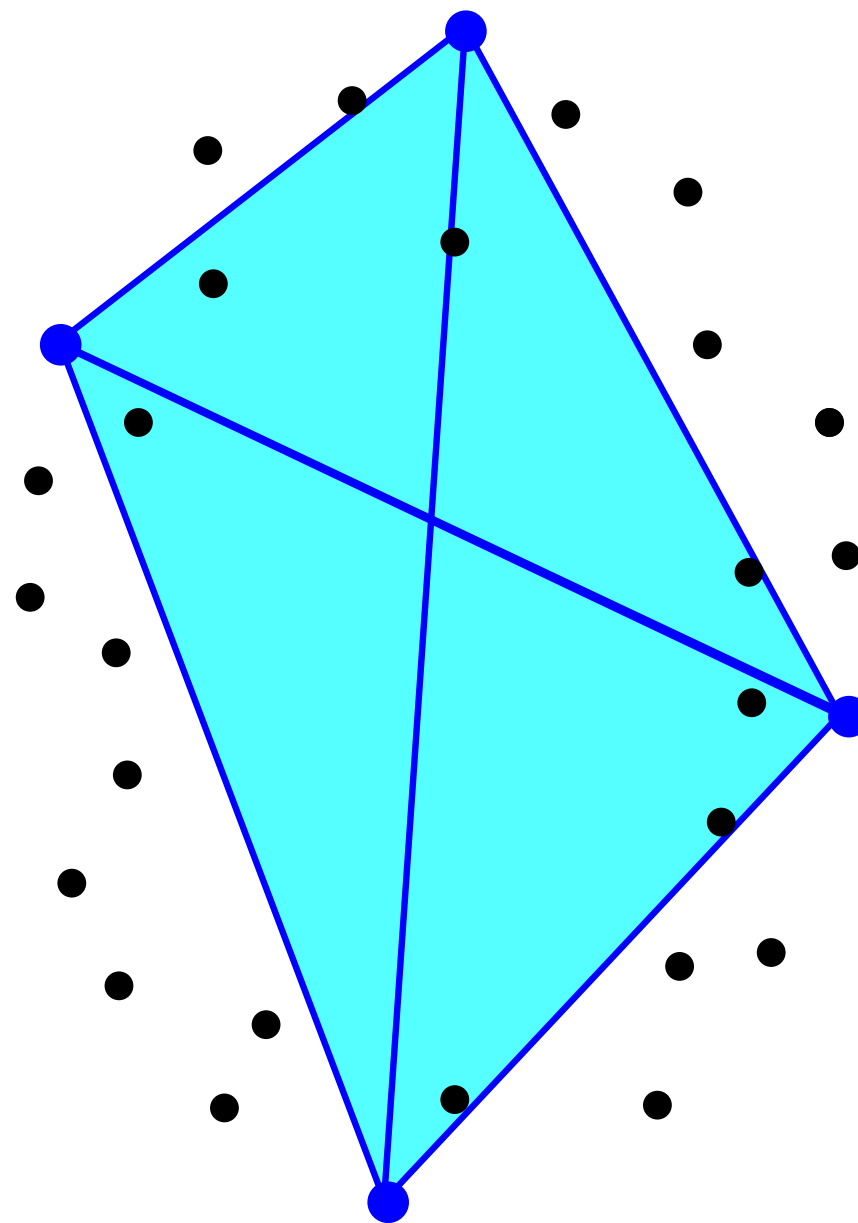
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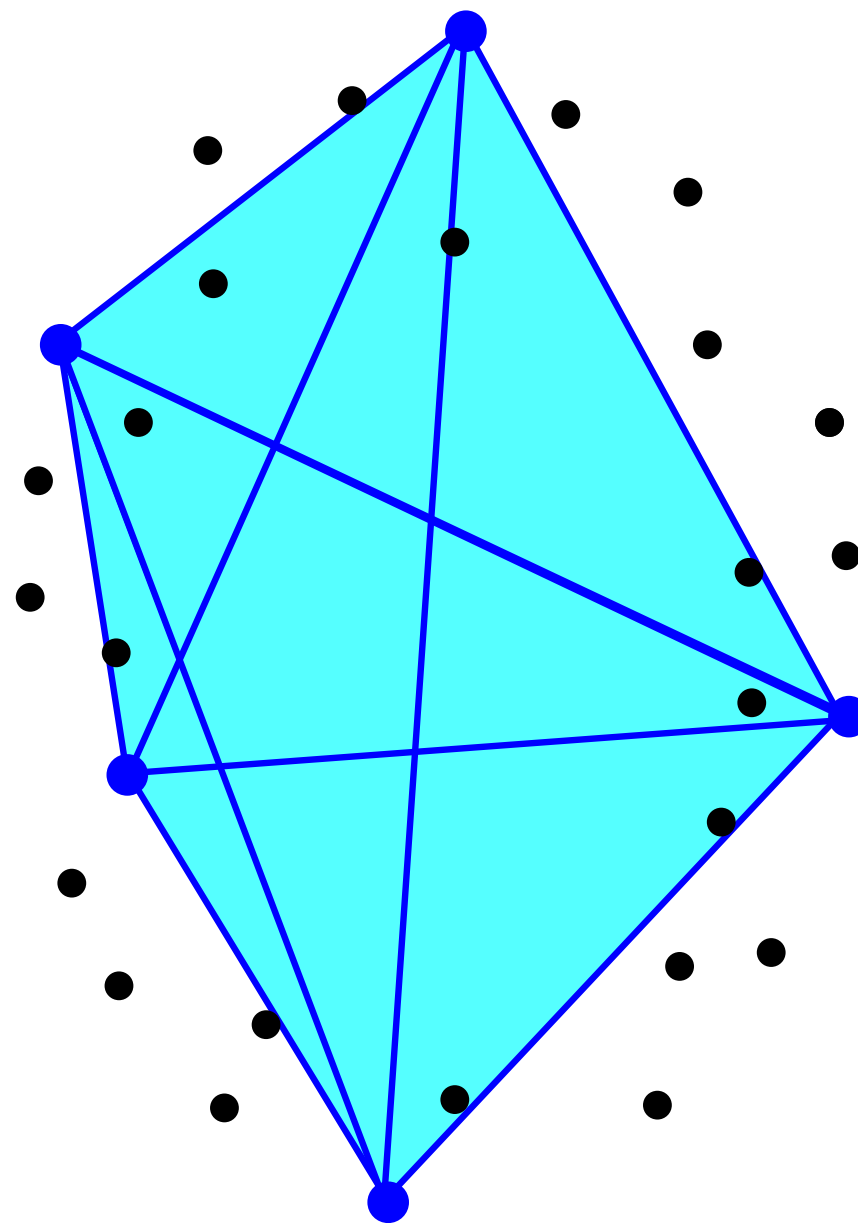
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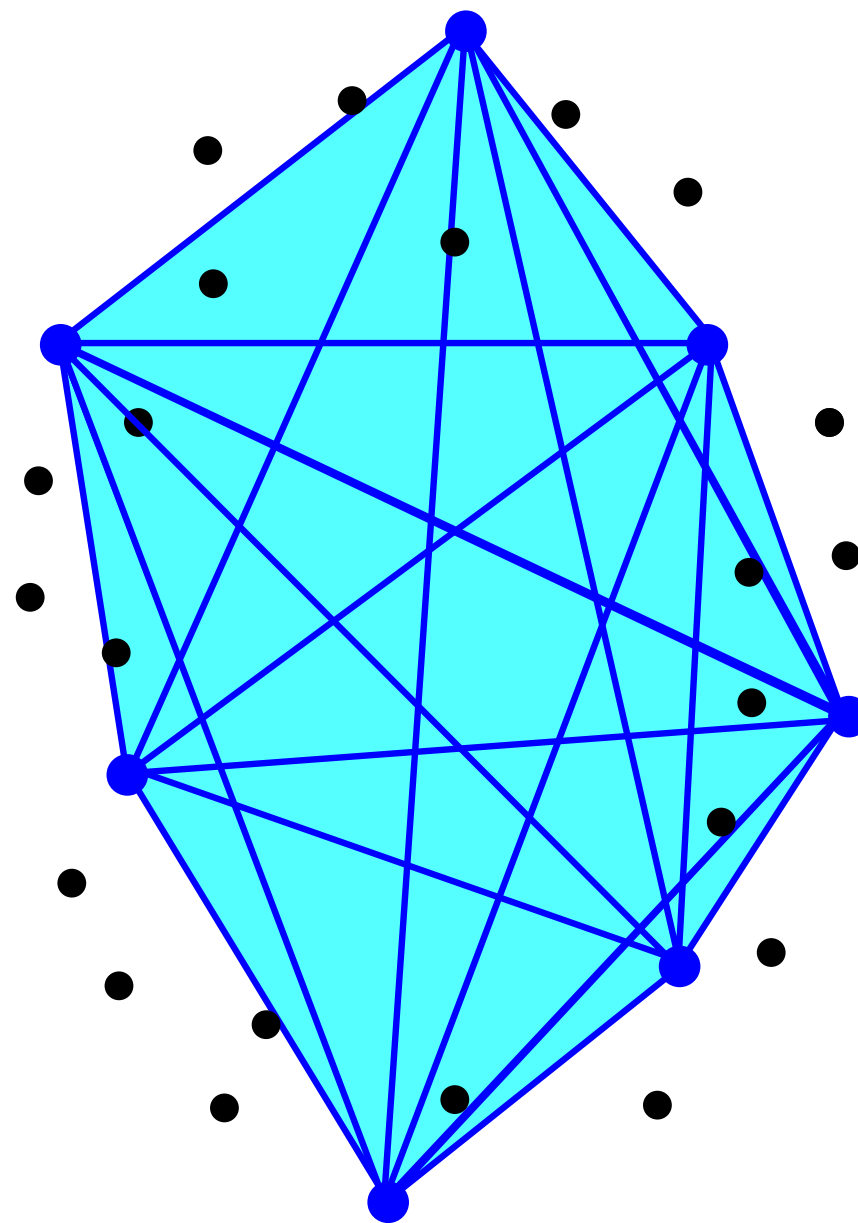
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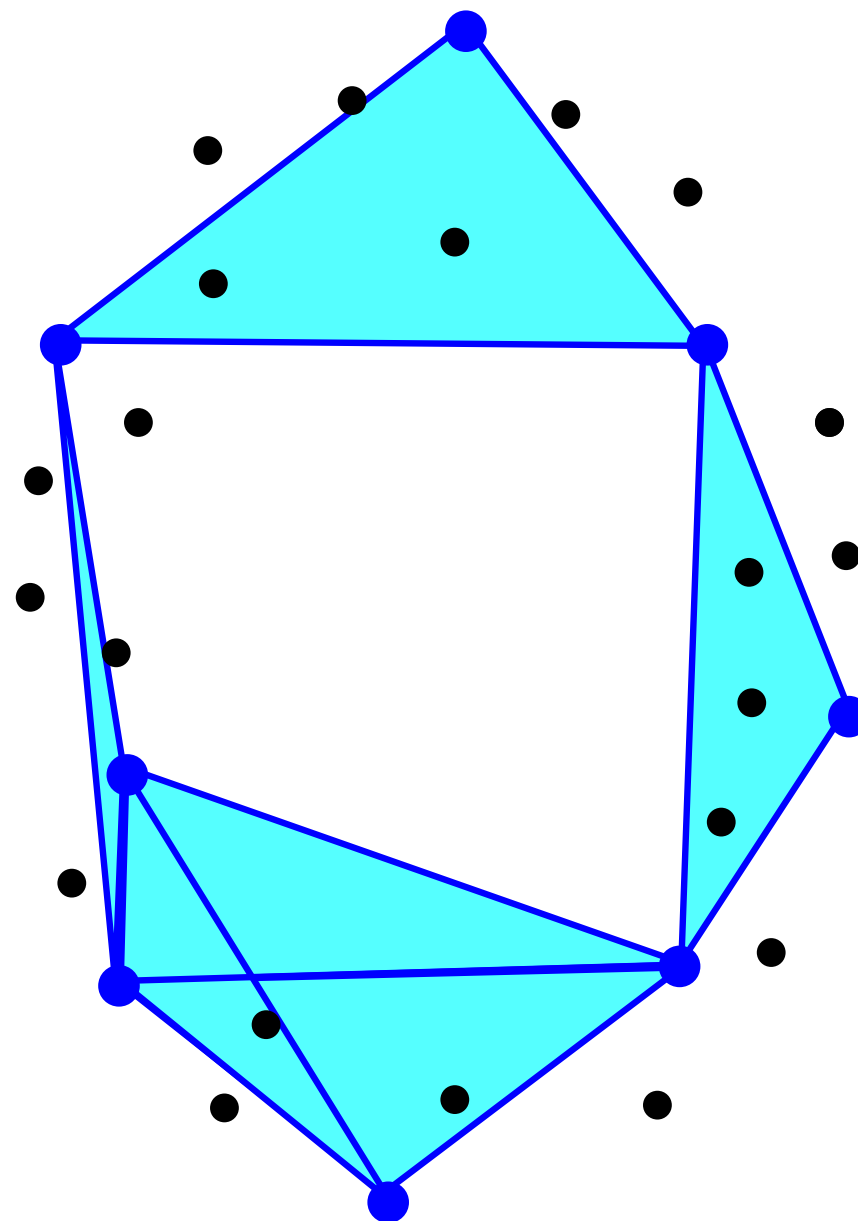
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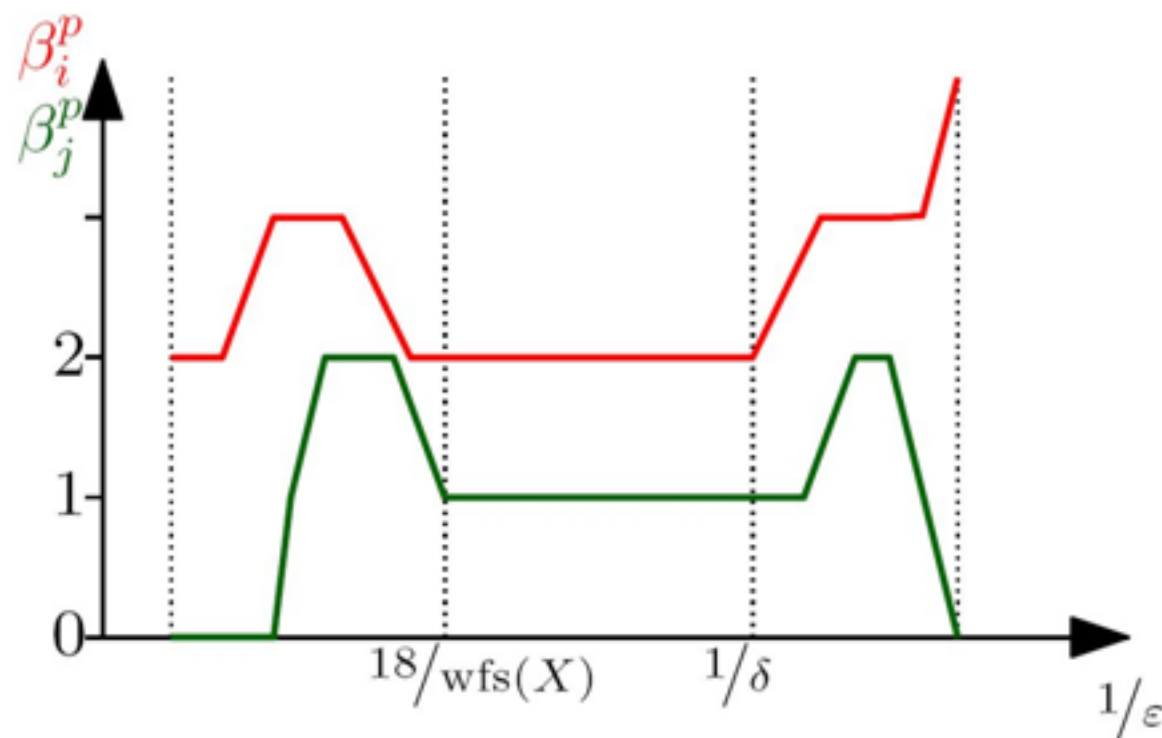
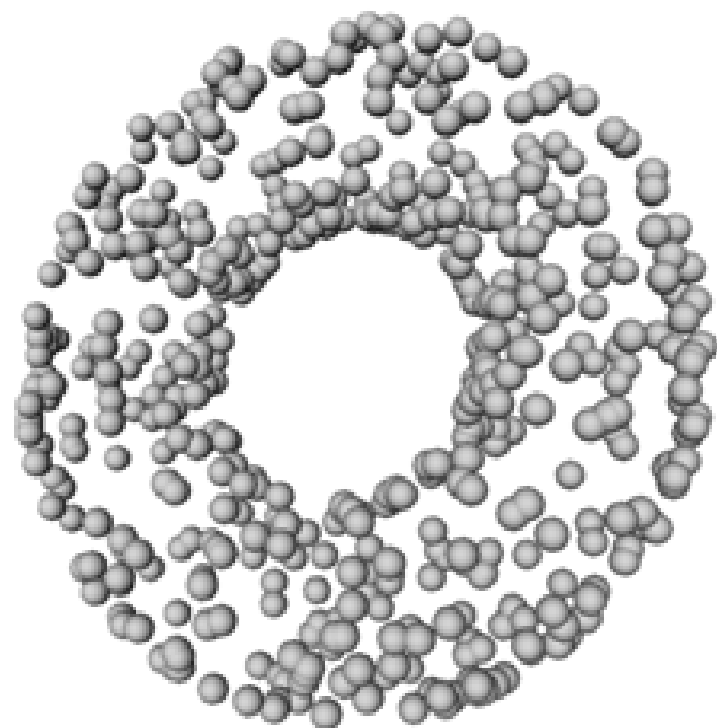
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# Multiscale inference



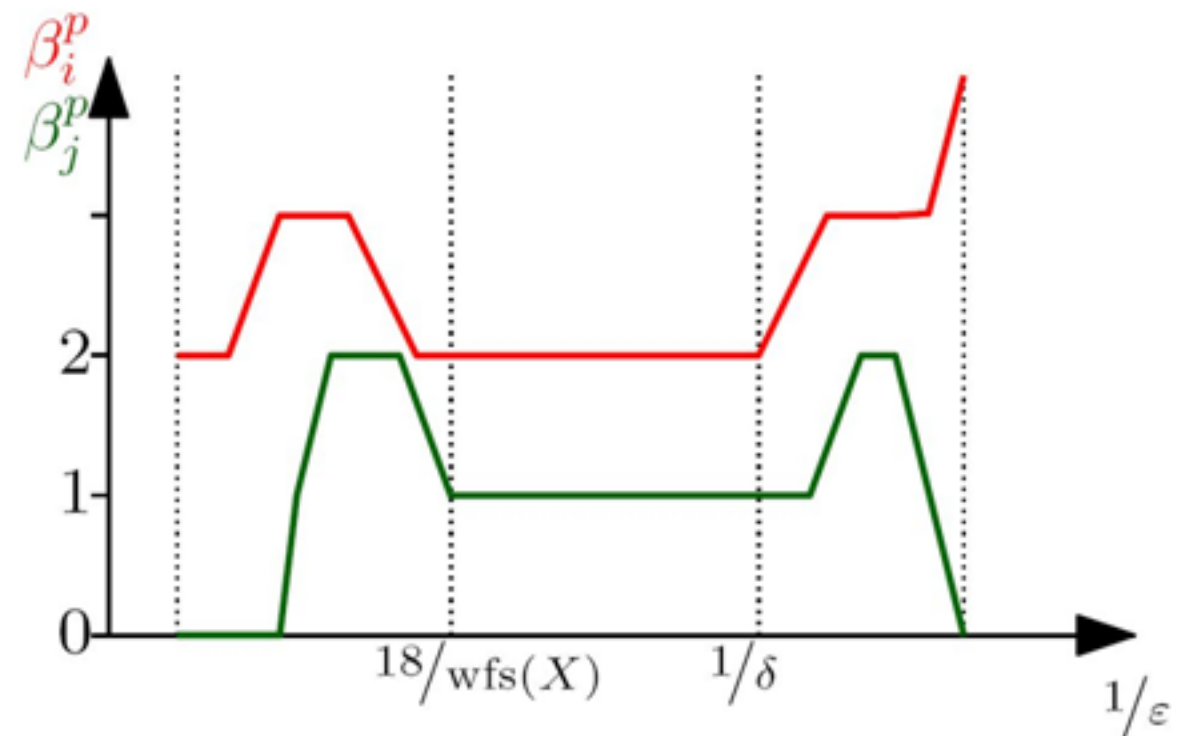
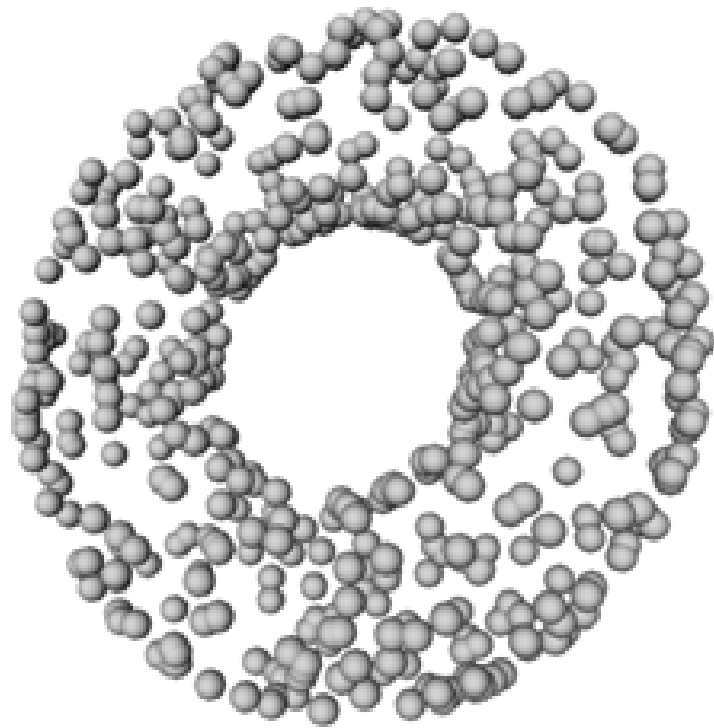
**Theorem:** [C-Oudot'08]

If  $d_H(W, X) < \delta$  for  $\delta < \frac{1}{18} \text{wfs}(X)$ , then at every iteration of the algorithm such that  $\delta < \epsilon < \frac{1}{18} \text{wfs}(X)$ ,

$$\beta_k(X^\lambda) = \dim H_k(X^\lambda) = rk(H_k(\mathcal{R}^{4\epsilon}(L)) \rightarrow H_k(\mathcal{R}^{4\epsilon}(L)))$$

for any  $\lambda \in (0, \text{wfs}(X))$  and any  $k \in \mathbb{N}$ .

# Multiscale inference



## Complexity of the algorithm:

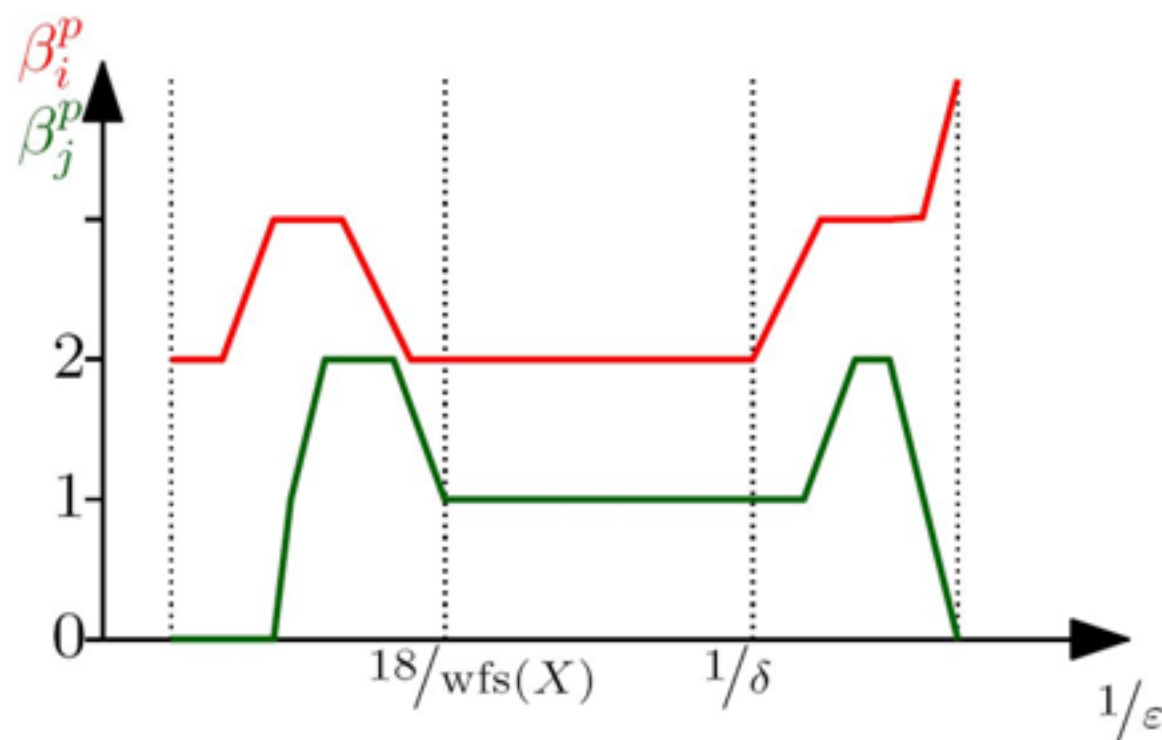
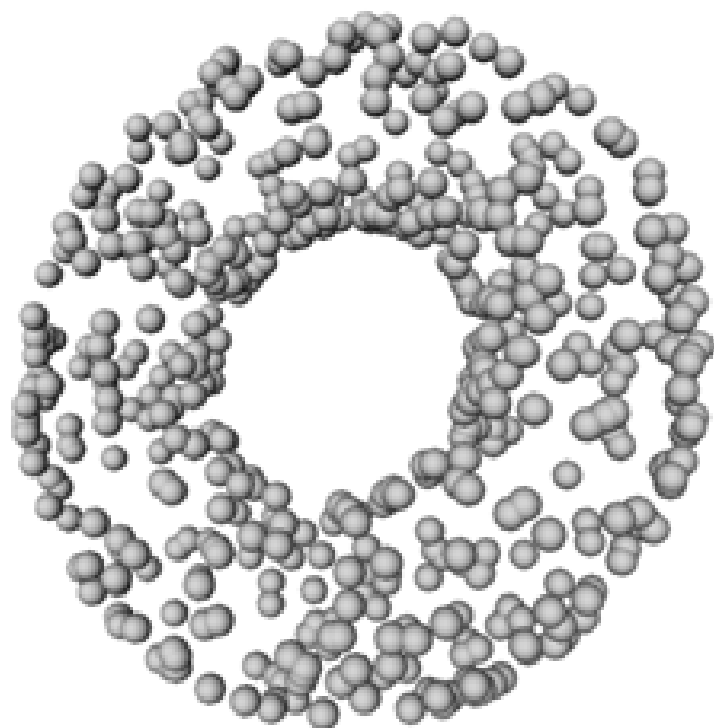
- If  $X \subset \mathbb{R}^d$  is non smooth the running time of the algorithm is

$$O(8^{33^d} |W|^5)$$

- If  $X$  is a smooth submanifold of  $\mathbb{R}^d$  dimension  $m$  the running time is

$$O(8^{35^m} |W|)$$

# Multiscale inference

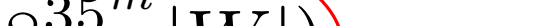


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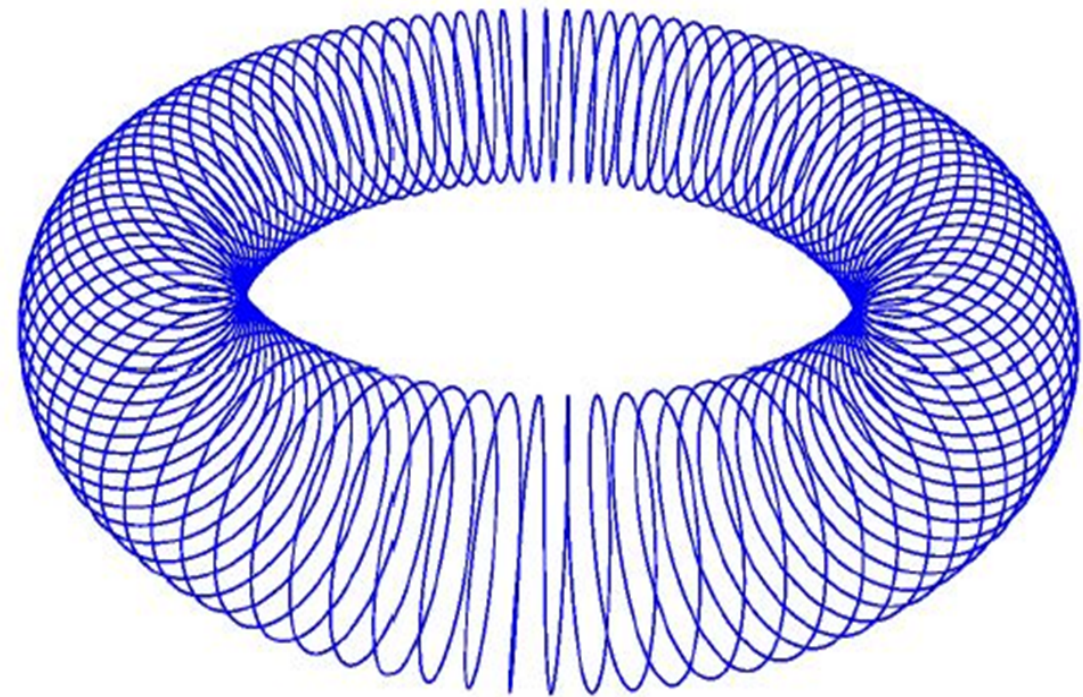
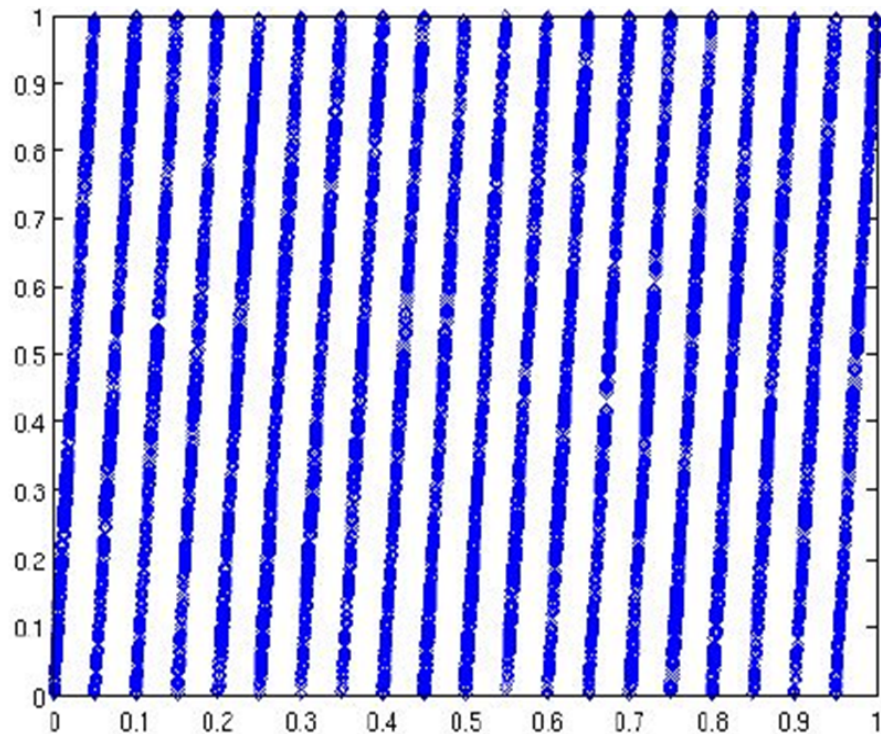
A diagram consisting of a red oval containing the expression  $O(8^{35^m} |W|)$ . A red arrow points from the right side of the oval to the text "Depend on the intrinsic dimension of  $X$ ".



# A synthetic example

$[0, 1] \times [0, 1]$

$\mathbb{R}^{1000}$

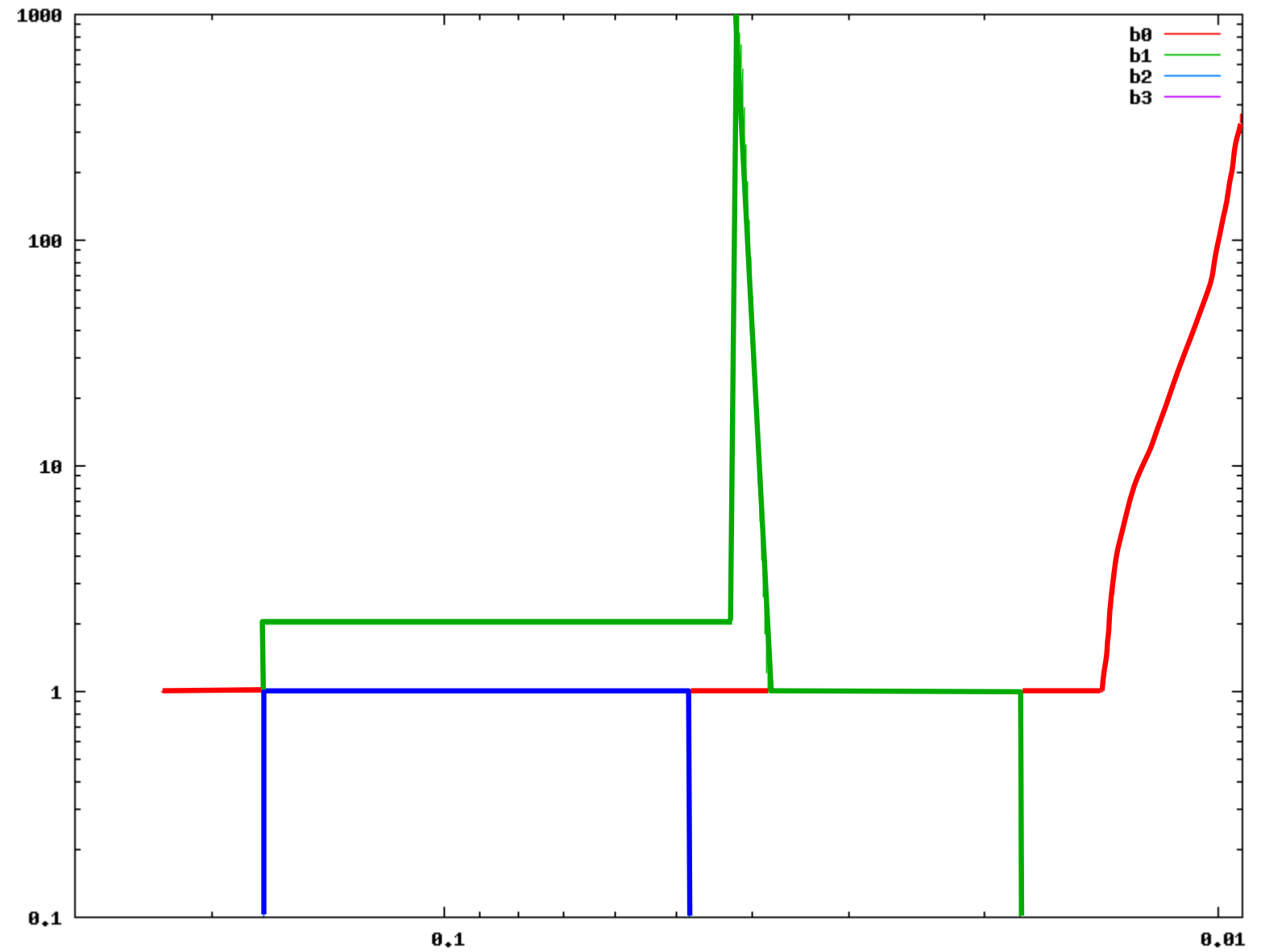
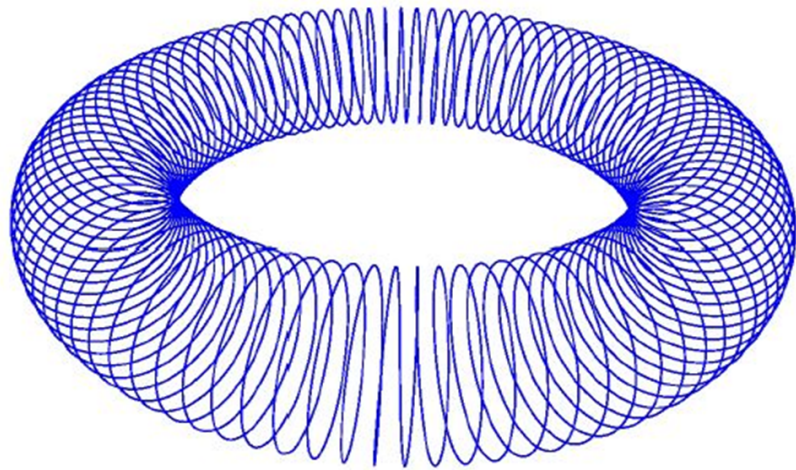


Non-linear embedding of  $S^1 \times S^1$  in  $\mathbb{R}^{1000}$

50,000 points sampled uniformly at random from a curve drawn on the 2-torus  $S^1 \times S^1$ .

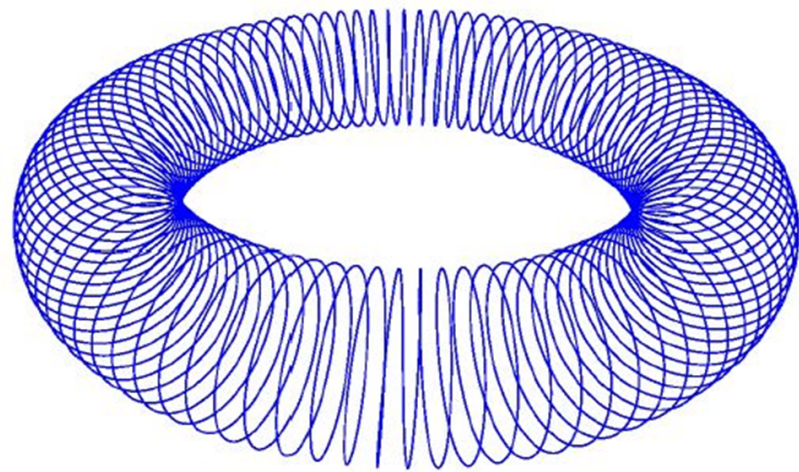


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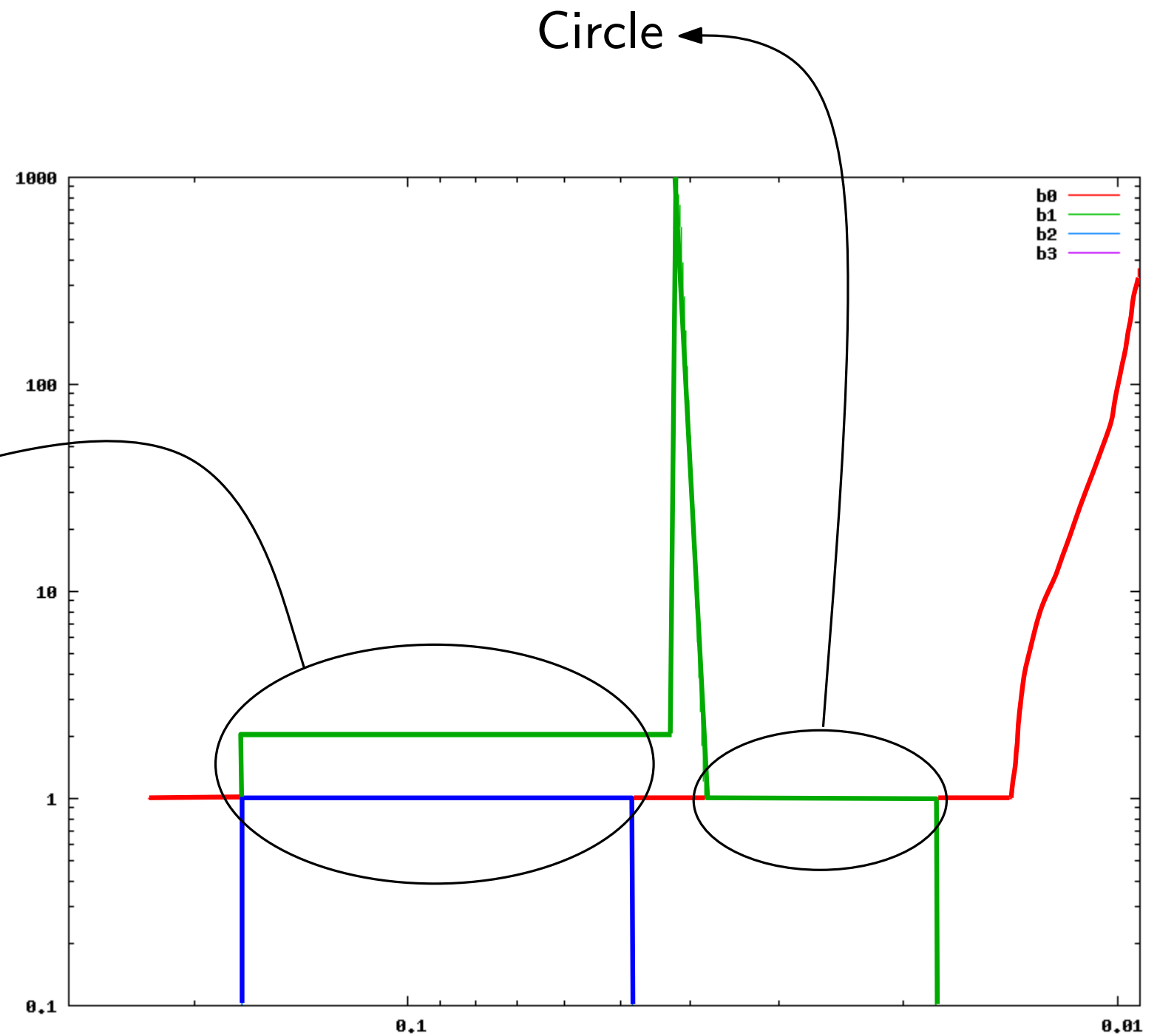


Output: sequence of Betti numbers on a log-log scale

# A synthetic example



Torus



Output: sequence of Betti numbers on a log-log scale

# An algorithm to compute Betti numbers

**Input:** A filtration of a simplicial complex  $\emptyset = K^0 \subset K^1 \subset \dots \subset K^m = K$ ,  
s. t.  $K^{i+1} = K^i \cup \sigma^{i+1}$  where  $\sigma^{i+1}$  is a simplex of  $K$ .

**Output:** The Betti numbers  $\beta_0, \beta_1, \dots, \beta_d$  of  $K$ .

$\beta_0 = \beta_1 = \dots = \beta_d = 0;$

for  $i = 1$  to  $m$

$k = \dim \sigma^i - 1;$

    if  $\sigma^i$  is contained in a  $(k+1)$ -cycle in  $K^i$

        then  $\beta_{k+1} = \beta_{k+1} + 1;$

        else  $\beta_k = \beta_k - 1;$

    end if;

end for;

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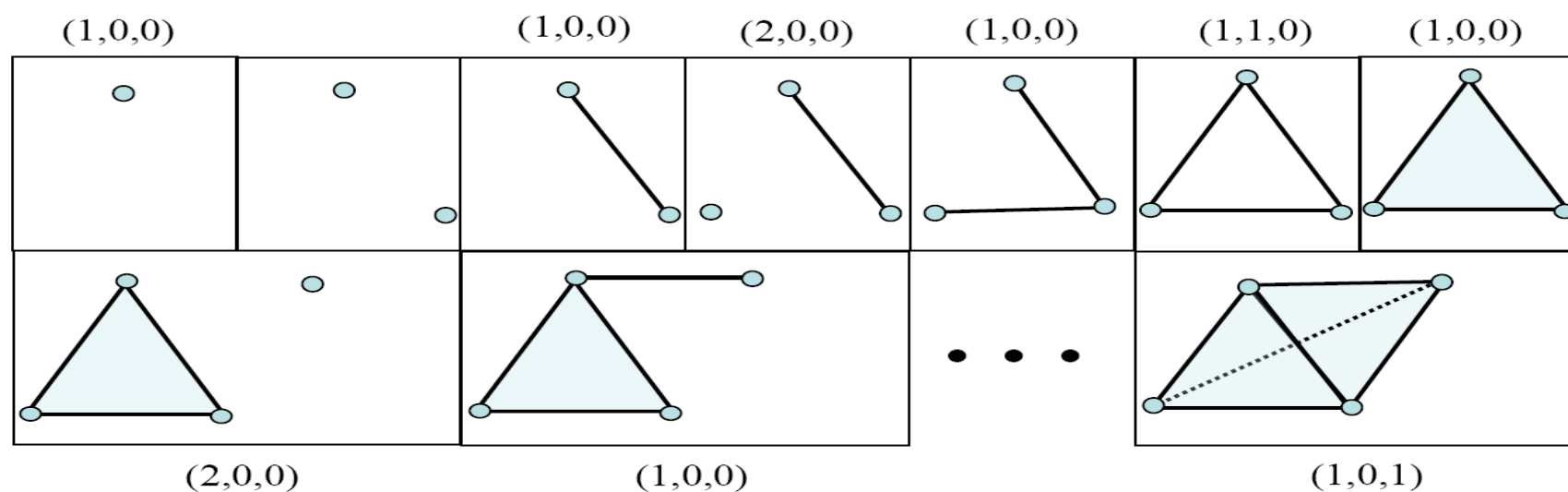
then  $\beta_{k+1} = \beta_{k+1} + 1$ ;

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  end if;
end for;
output  $(\beta_0, \beta_1, \dots, \beta_d);$ 
```

**Remark:** At the  $i^{th}$  step of the algorithm, the vector  $(\beta_0, \dots, \beta_d)$  stores the Betti numbers of  $K^i$ .

# Proof

- If  $\sigma^i$  is contained in a  $(k+1)$ -cycle in  $K^i$ , this cycle is not a boundary in  $K^i$ .
- If  $\sigma^i$  is contained in a  $(k+1)$ -cycle  $c$  in  $K^i$ , then  $c$  cannot be homologous to a cycle in  $K^{i-1}$

$$\Rightarrow \beta_{k+1}(K^i) \geq \beta_{k+1}(K^{i-1}) + 1$$

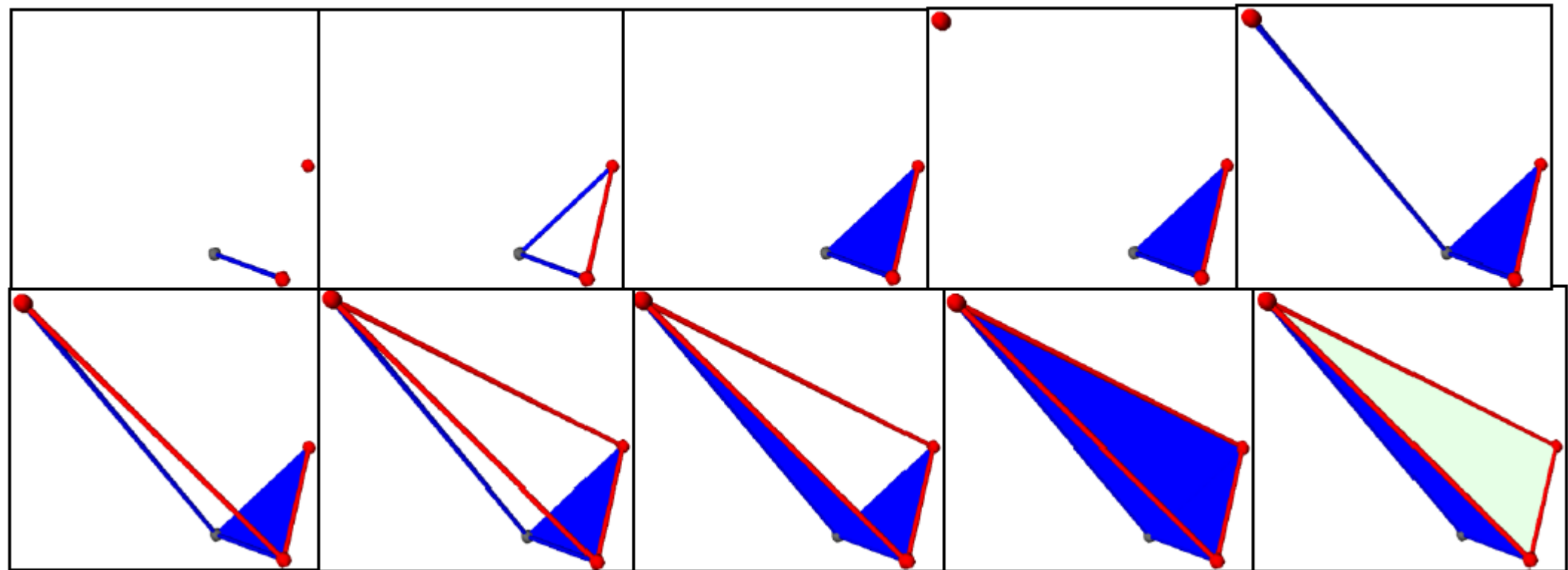
- If  $\sigma^i$  is not contained in a  $(k+1)$ -cycle  $c$  in  $K^i$ , then  $\partial\sigma^i$  is not a boundary in  $K^{i-1}$

$$\Rightarrow \beta_k(K^i) \leq \beta_k(K^{i-1}) - 1$$

- the previous inequalities are equalities.

# Positive and negative simplices

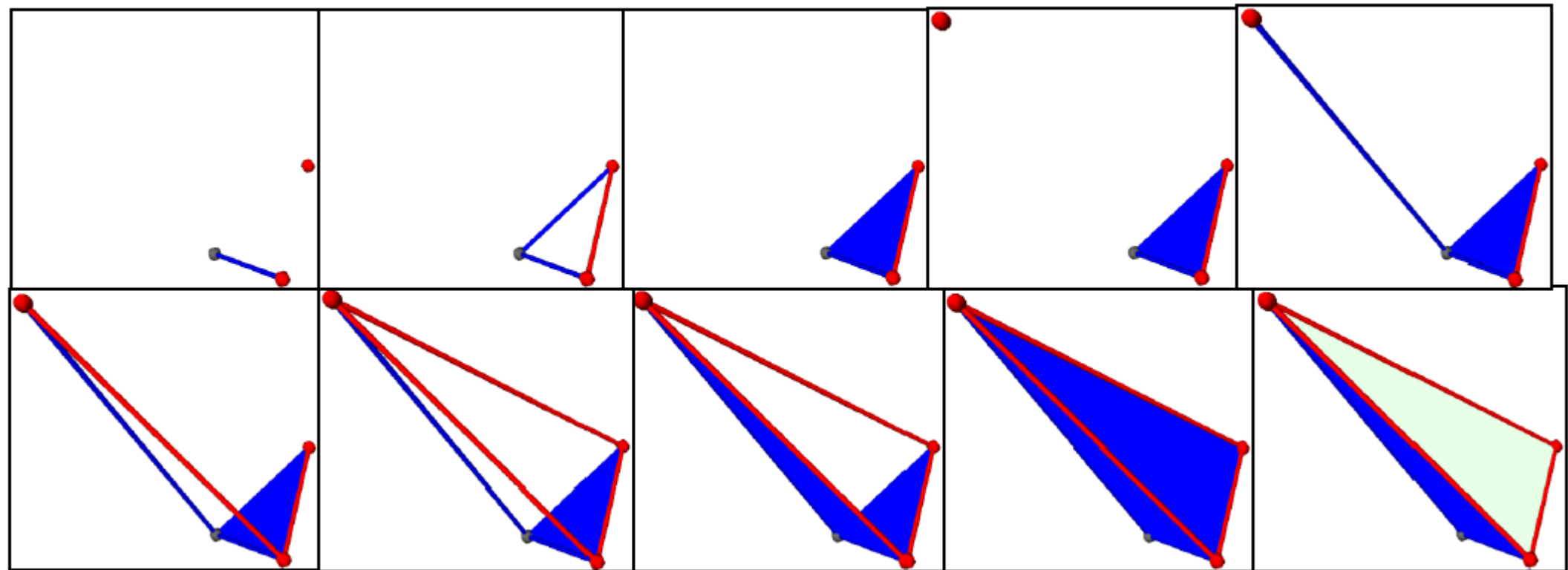
Let  $\emptyset = K^0 \subset K^1 \subset \dots \subset K^m = K$  be a filtration of a simplicial complex  $K$  s. t.  $K^{i+1} = K^i \cup \sigma^{i+1}$  where  $\sigma^{i+1}$  is a simplex of  $K$ .



**Definition:** A  $(k+1)$ -simplex  $\sigma^i$  is **positive** if it is contained in a  $(k+1)$ -cycle in  $K^i$ . It is **negative** otherwise.

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Destroy a  $k$ -cycle in  $K^i$

Create a new  $(k+1)$ -cycle in  $K^i$

$$\beta_k(K) = \#(\text{positive simplices}) - \#(\text{negative simplices})$$



# Getting more information

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- How to keep track of the evolution of the topology all along the filtration?
- What are the created/destroyed cycles?
- What is the lifetime of a cycle?
- How to compute  $\text{rank}(H_k(K^i) \rightarrow H_k(K^j))$ ?

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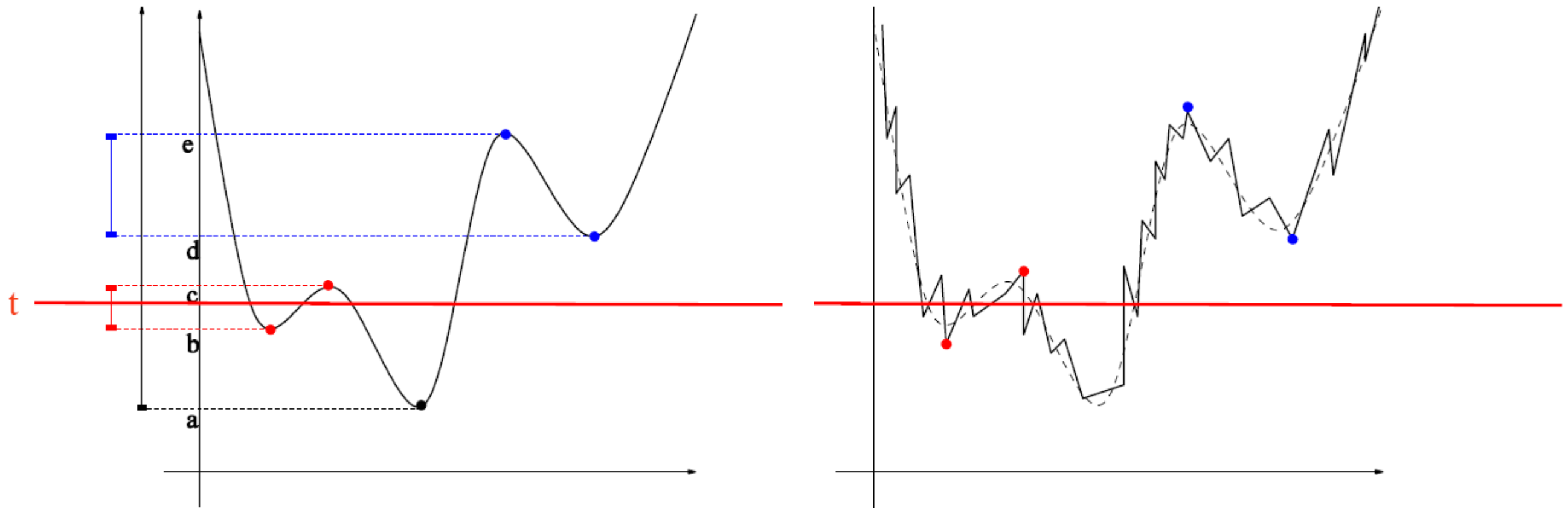
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This is where topological persistence comes into play!

# Topological persistence

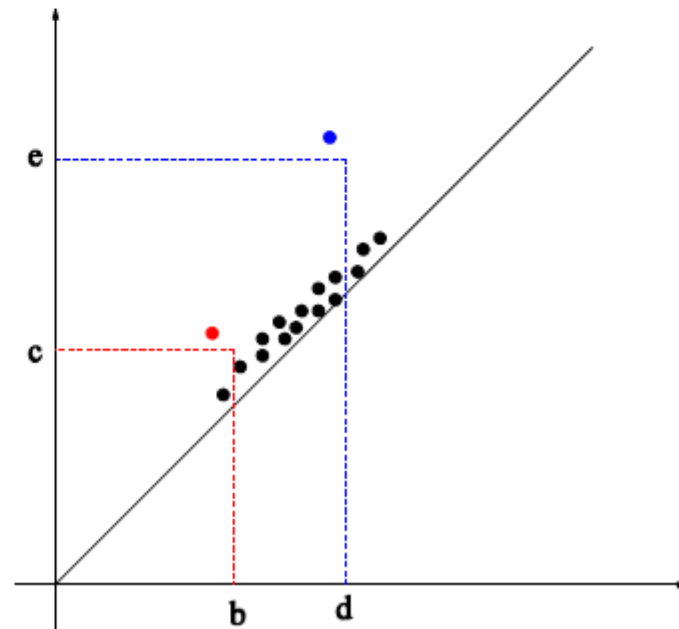
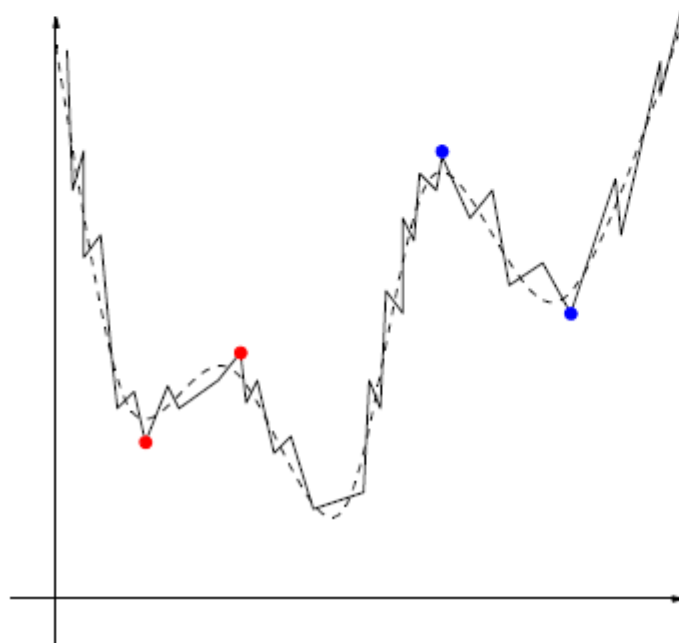
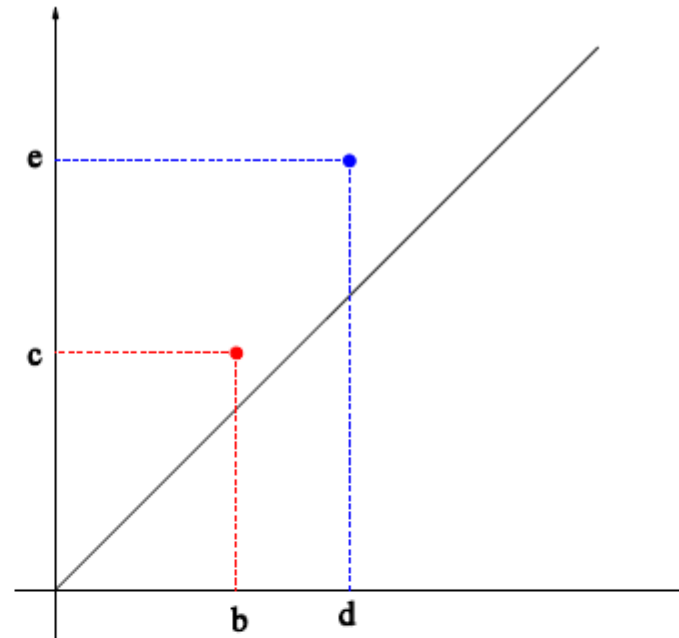
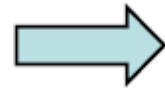
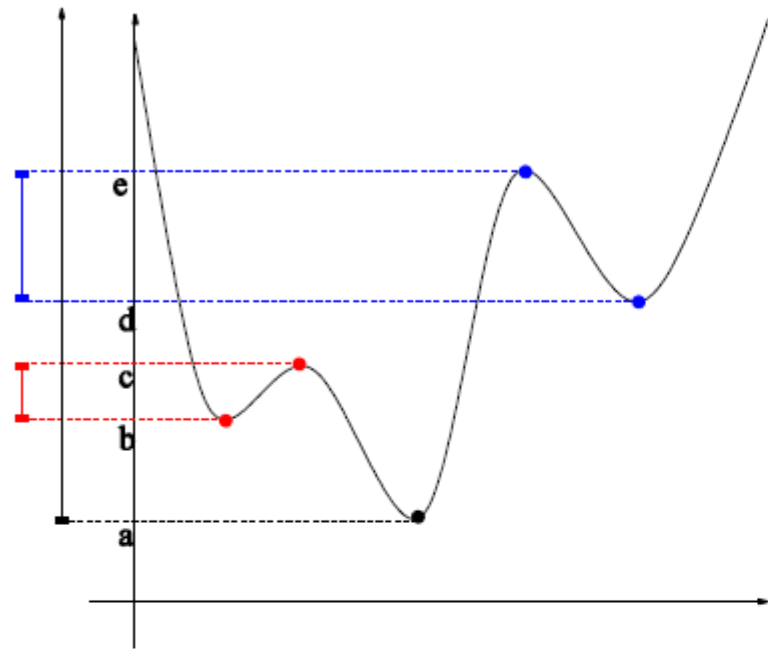
- a tool to study topological properties of data (represented by real valued functions on topological spaces).
- A method that allow to separate information from topological noise.
- References:
  - H. Edelsbrunner, D. Letscher and A. Zomorodian. *Topological persistence and simplification*. Discrete Comput. Geom., 28:511-533, 2002.
  - D. Cohen-Steiner and H. Edelsbrunner and J. Harer, *Stability of Persistence Diagrams*, Proc. 21st ACM Sympos. Comput. Geom. 2005.
  - F. Chazal and D. Cohen-Steiner and L. J. Guibas and M. Glisse and S. Y. Oudot, *Proximity of Persistence Modules and their Diagrams*, Proc. 25th ACM Sympos. Comput. Geom. 2009.

# A simple example



- What is the relevant number of connected components of  $f^{-1}((-\infty, t])$ ?
- More generally, study the topology of the sublevel sets  $f^{-1}((-\infty, t])$  as  $t$  varies.

# A simple example: filter out topological noise

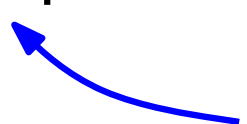


Persistence  
diagrams

# Functions defined over higher dimensional spaces

- $f : X \rightarrow \mathbb{R}$  continuous where  $X$  is a topological space
- Not only connected components but also cycles, voids, etc...  $\rightarrow$  persistence of homological features / evolution of  $H_k(f^{-1}((-\infty, t]))$

Relation between functions and filtrations:

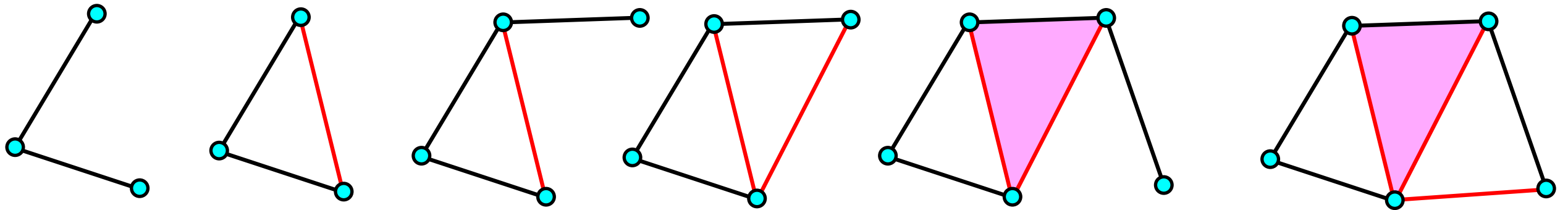
- $\forall t \leq t' \in \mathbb{R}, f^{-1}((-\infty, t]) \subseteq f^{-1}((-\infty, t']) \rightarrow$  filtration of  $X$  by the sublevel sets of  $f$ .
  - If  $f$  is defined at the vertices of a simplicial complex  $K$ , the sublevel sets filtration is a filtration of the simplicial complex  $K$ .
    - For  $\sigma = [v_0, \dots, v_k] \in K, f(\sigma) = \max_{i=0, \dots, k} f(v_i)$
    - The simplices of  $K$  are ordered according increasing  $f$  values (and dimension in case of equal values on different simplices).
- 

# Notations

In the following:

- Let  $\emptyset = K^0 \subset K^1 \subset \dots \subset K^m = K$  be a filtration of a simplicial complex  $K$  s. t.  $K^{i+1} = K^i \cup \sigma^{i+1}$  where  $\sigma^{i+1}$  is a simplex of  $K$ .
- $Z_k^i$  = the  $k$ -cycles of  $K^i$ ,  $B_k^i$  = the  $k$ -boundaries of  $K^i$  and  $H_k^i$  = the  $k^{th}$ -homology group of  $K^i$ .
- $Z_k^0 \subseteq Z_k^1 \subseteq \dots \subseteq Z_k^i \subseteq \dots \subseteq Z_k^m = Z_k(K)$
- $B_k^0 \subseteq B_k^1 \subseteq \dots \subseteq B_k^i \subseteq \dots \subseteq B_k^m = B_k(K)$

# Cycle associated to a positive simplex



**Lemma:** If  $\sigma^i$  is a positive  $k$ -cycle, then there exists a  $k$ -cycle  $c_\sigma$  s.t.:

- $c_\sigma$  is not a boundary in  $K^i$ ,
- $c_\sigma$  contains  $\sigma^i$  but no other positive  $k$ -simplex.

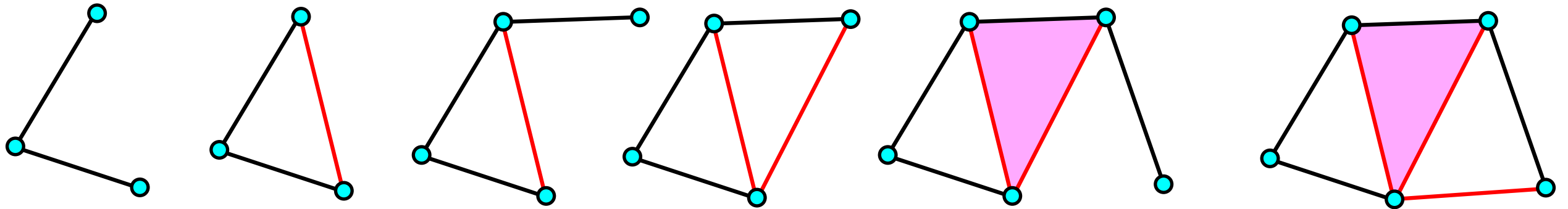
The cycle  $c^\sigma$  is unique.

**Proof:**

By induction on the order of appearance of the simplices in the filtration.

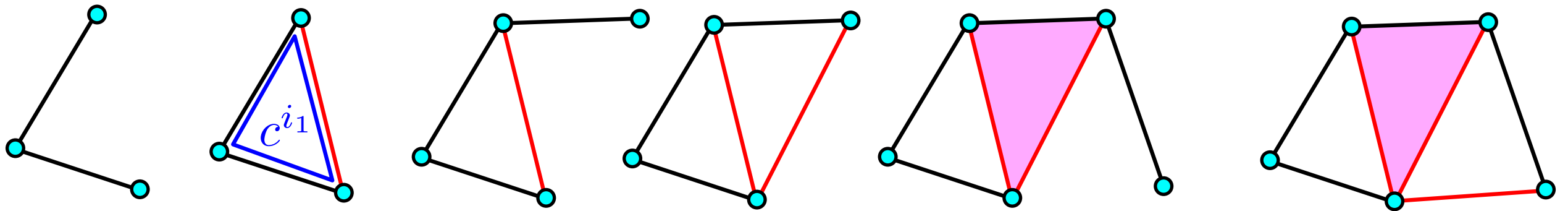


# Homology basis



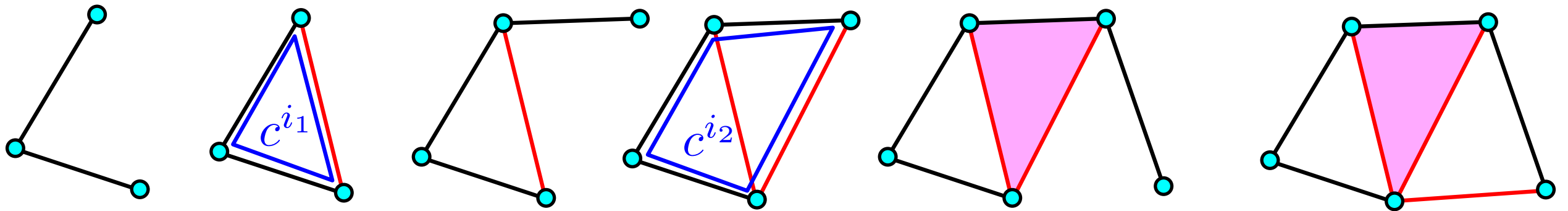
- At the beginning: the basis of  $H_k^0$  is empty.
- If a basis of  $H_k^{i-1}$  has been built and  $\sigma^i$  is a positive  $k$ -simplex then one adds the homology class of the cycle  $c^i$  associated to  $\sigma^i$  to the basis of  $H_k^{i-1} \Rightarrow$  basis of  $H_k^i$ .
- If a basis of  $H_k^{j-1}$  has been built and  $\sigma^j$  is a negative  $(k+1)$ -simplex:
  - let  $c^{i_1}, \dots, c^{i_p}$  be the cycles associated to the positive simplices  $\sigma^{i_1}, \dots, \sigma^{i_p}$  that form a basis of  $H_k^{j-1}$
  - $d = \partial\sigma^j = \sum_{k=1}^p \varepsilon_k c^{i_k} + b$
  - $l(j) = \max\{i_k : \varepsilon_k = 1\}$
  - Remove the homology class of  $c^{l(j)}$  from the basis of  $H_k^{j-1} \Rightarrow$  basis of  $H_k^j$ .

# Homology basis



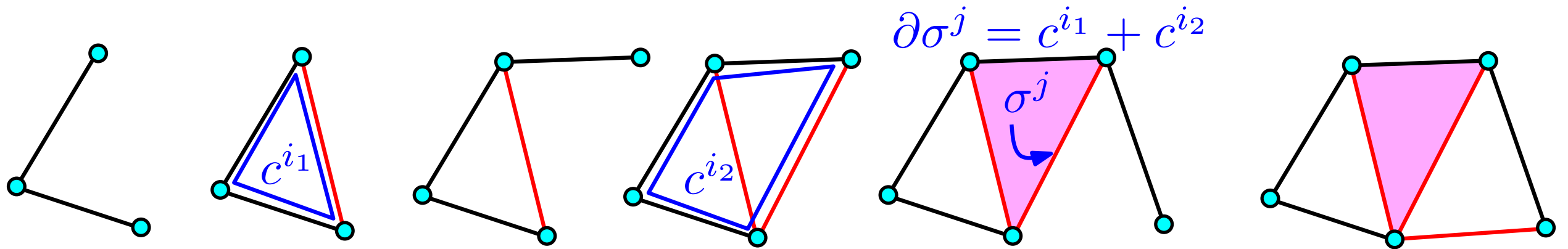
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# Pairing simplices

- If a basis of  $H_k^{j-1}$  has been built and  $\sigma^j$  is a negative  $(k+1)$ -simplex:
  - let  $c^{i_1}, \dots, c^{i_p}$  be the cycles associated to the positive simplices  $\sigma^{i_1}, \dots, \sigma^{i_p}$  that form a basis of  $H_k^{j-1}$
  - $d = \partial\sigma^j = \sum_{k=1}^p \varepsilon_k c^{i_k} + b$
  - $l(j) = \max\{i_k : \varepsilon_k = 1\}$
  - Remove the homology class of  $c^{l(j)}$  from the basis of  $H_k^{j-1} \Rightarrow$  basis of  $H_k^j$ .

The simplices  $\sigma^{l(j)}$  and  $\sigma^j$  are paired to form a **persistent pair**  $(\sigma^{l(j)}, \sigma^j)$ .  
→ The homology class created by  $\sigma^{l(j)}$  in  $K^{l(j)}$  is killed by  $\sigma^j$  in  $K^j$ . The **persistence** (or life-time) of this cycle is :  $j - l(j) - 1$ .

**Remark:** filtrations of  $K$  can be indexed by increasing sequences  $\alpha_i$  of real numbers (useful when working with a function defined on the vertices of a simplicial complex).

# The persistence algorithm: first version

**Input:**  $\emptyset = K^0 \subset K^1 \subset \dots \subset K^m = K$  a  $d$ -dimensional filtration of a simplicial complex  $K$  s. t.  $K^{i+1} = K^i \cup \sigma^{i+1}$  where  $\sigma^{i+1}$  is a simplex of  $K$ .

$$L_0 = L_1 = \dots = L_{d-1} = \emptyset$$

For  $j = 0$  to  $m$

$$k = \dim \sigma^j - 1;$$

if  $\sigma^j$  is a negative simplex

$$l(j) = \text{highest index of the positive simplices associated to } \partial\sigma^j;$$

$$L_k = L_k \cup \{(\sigma^{l(j)}, \sigma^j)\};$$

end if

end for

output  $L_0, L_1, \dots, L_{d-1}$  ;

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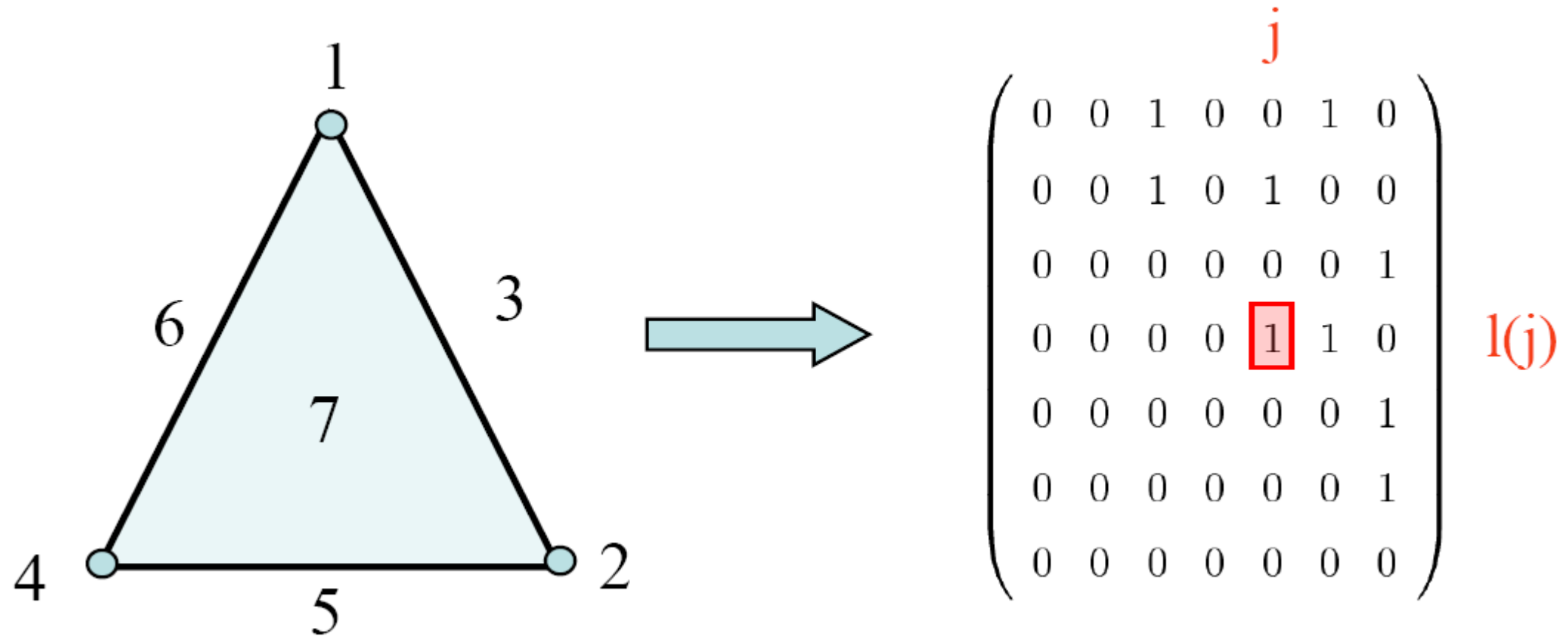
end if

end for

output  $L_0, L_1, \dots, L_{d-1}$  ;

How to test this condition?

# The matrix of the boundary operator



- $M = (m_{ij})_{i,j=1,\dots,m}$  with coefficient in  $\mathbb{Z}/2$  defined by

$$m_{ij} = 1 \text{ if } \sigma^i \text{ is a face of } \sigma^j \text{ and } m_{ij} = 0 \text{ otherwise}$$

- For any column  $C_j$ ,  $l(j)$  is defined by

$$(i = l(j)) \Leftrightarrow (m_{ij} = 1 \text{ and } m_{i'j} = 0 \forall i' > i)$$



# The persistence algorithm: second version

**Input:**  $\emptyset = K^0 \subset K^1 \subset \dots \subset K^m = K$  a  $d$ -dimensional filtration of a simplicial complex  $K$  s. t.  $K^{i+1} = K^i \cup \sigma^{i+1}$  where  $\sigma^{i+1}$  is a simplex of  $K$ .

For  $j = 0$  to  $m$

    While (there exists  $j' < j$  such that  $l(j') == l(j)$ )

$C_j = C_j + C_{j'} \bmod(2);$

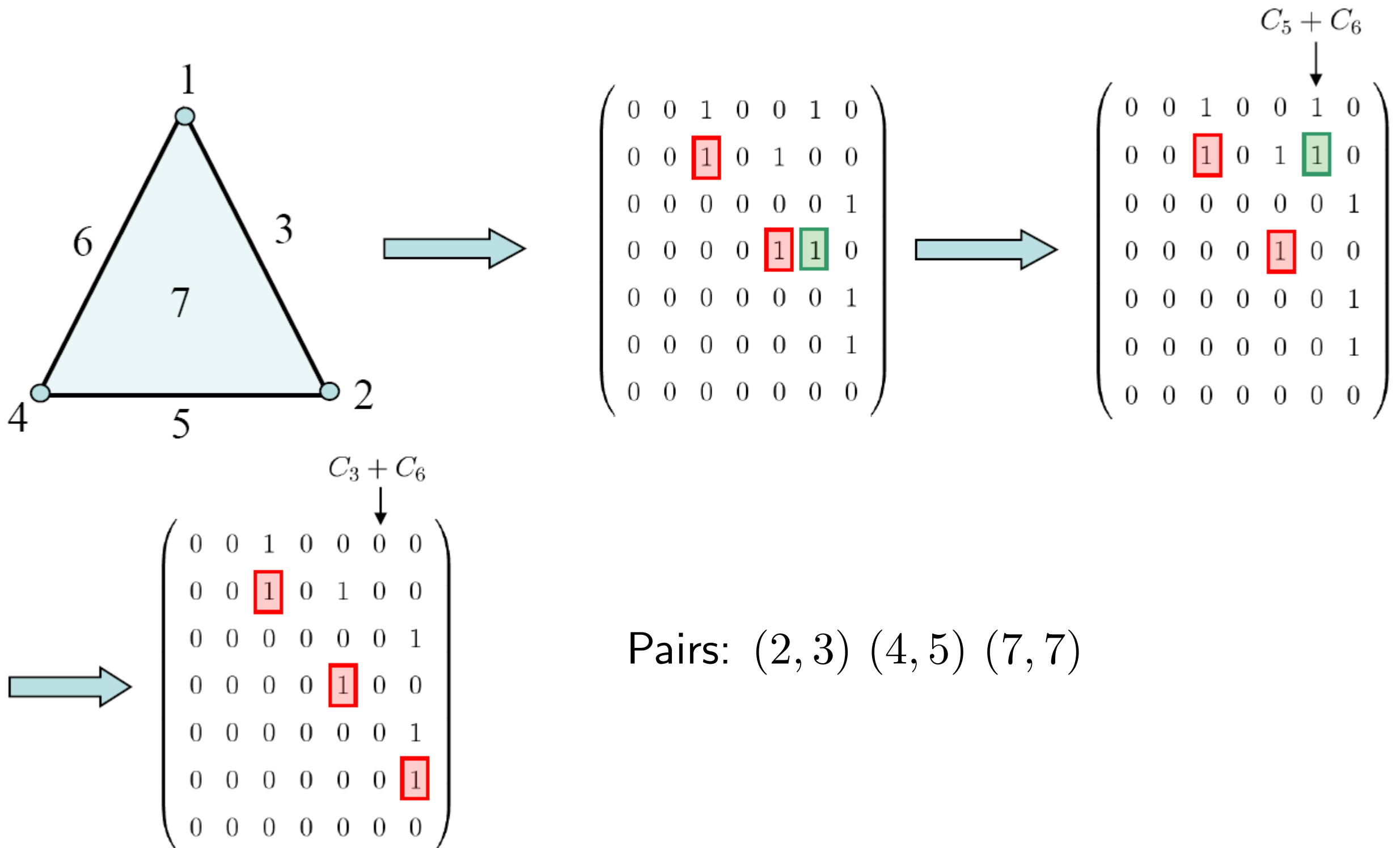
    End while

End for

Output the pairs  $(l(j), j);$

**Remark:** The worst case complexity of the algorithm is  $O(m^3)$  but much lower in most practical cases.

# A very simple example



# Correctness of the second algorithm

**Proposition:** the second algorithm outputs the persistence pairs.

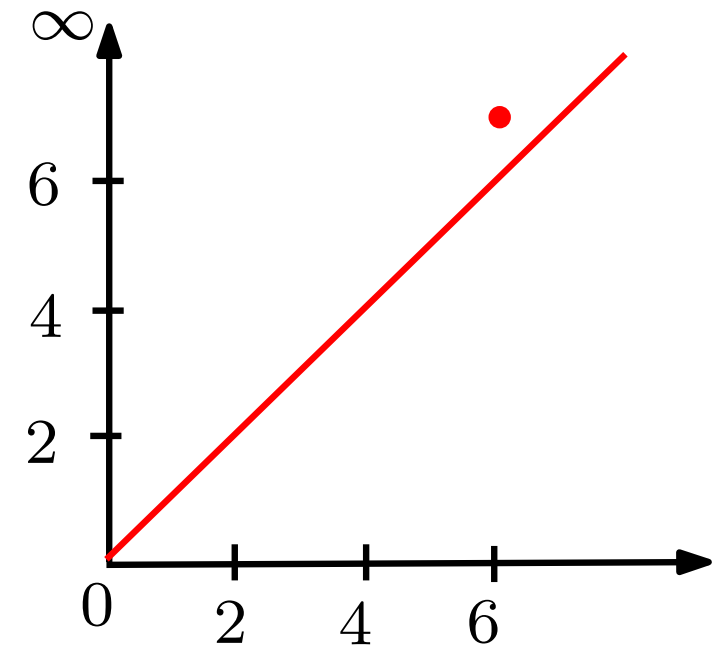
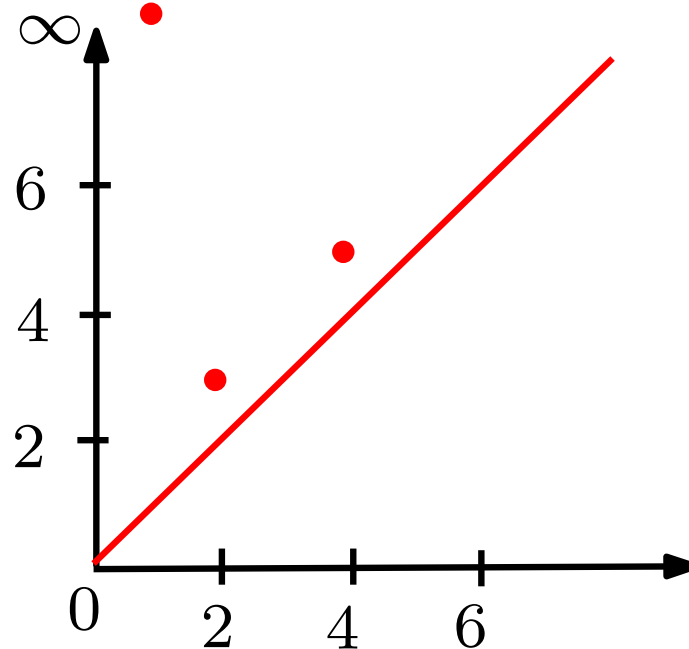
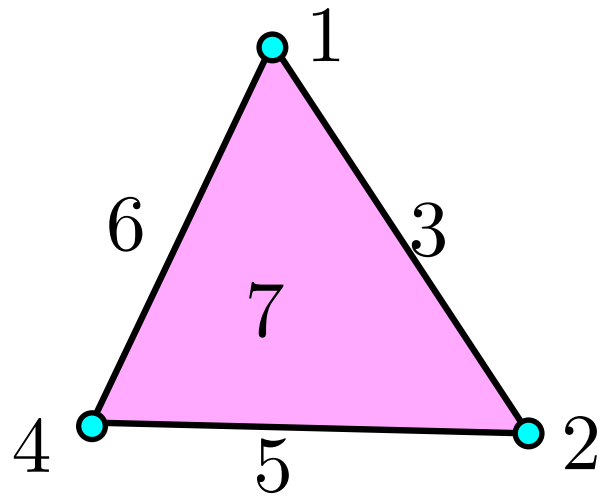
**Proof:** follows from the four remarks below.

1. At each step of the algorithm, the column  $C_j$  represents a chain of the form

$$\partial \left( \sigma^j + \sum_{i < j} \varepsilon_i \sigma^i \right) \text{ with } \varepsilon_i \in \{0, 1\}$$

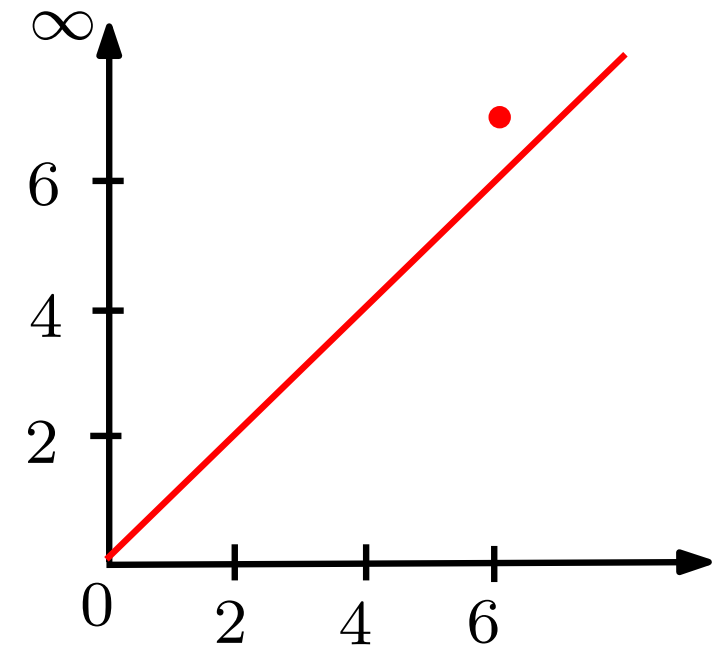
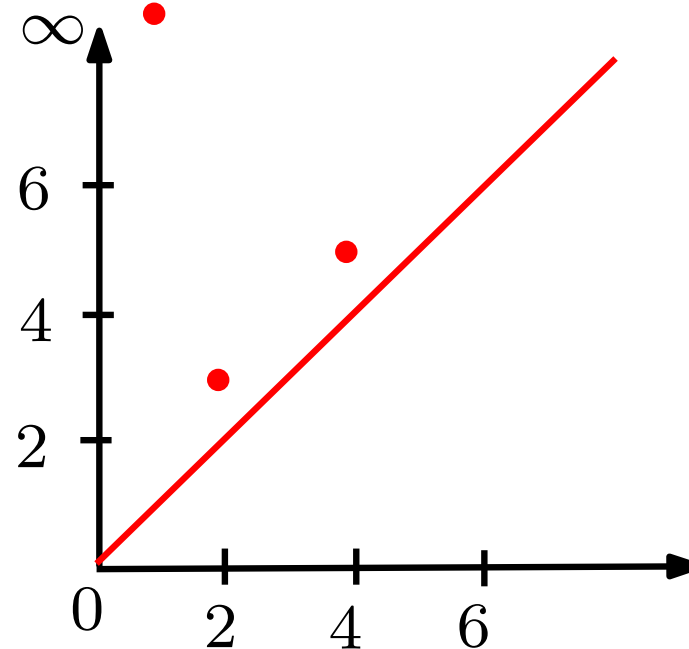
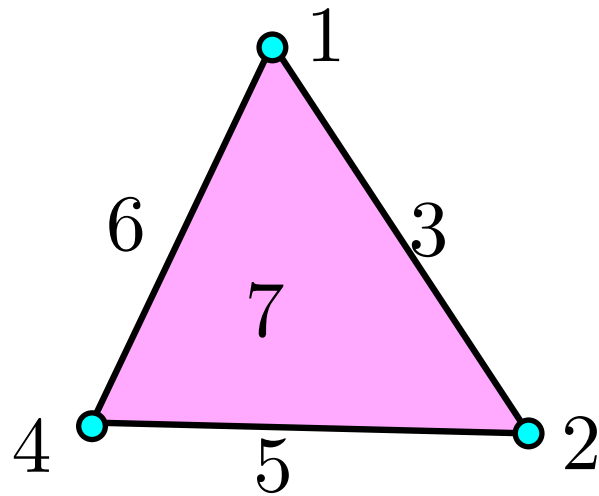
2. At this end of the algorithm, if  $j$  is s.t.  $l(j)$  is defined then  $\sigma^{l(j)}$  is a positive simplex.
3. If at the end of the algorithm if the column  $C_j$  is zero then  $\sigma^j$  is positive.
4. If at the end of the algorithm the column  $C_j$  is not zero then  $(\sigma^{l(j)}, \sigma^j)$  is a persistence pair.

# Persistence diagrams



- each pair  $(\sigma^{l(j)}, \sigma^j)$  is represented by  $(l(j), j)$  or  $(f(\sigma^{l(j)}), f(\sigma^j)) \in \mathbb{R}^2$  when considering filtrations induced by functions.
- The diagonal  $\{y = x\}$  is added to the persistence diagram.
- Unpaired positive simplex  $\sigma^i \rightarrow (i, +\infty)$ .

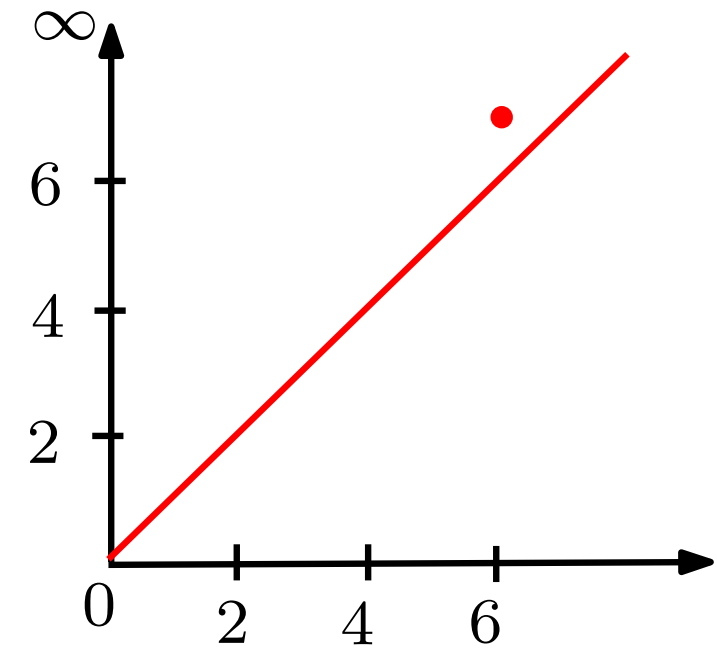
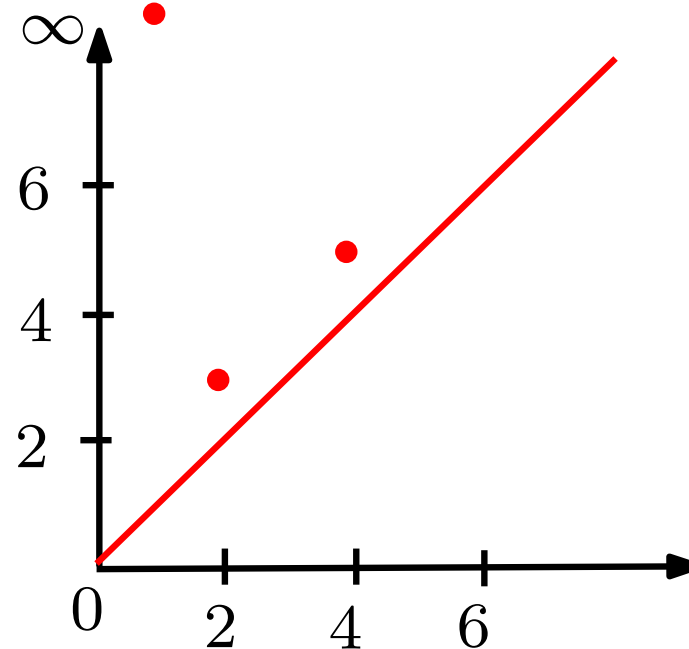
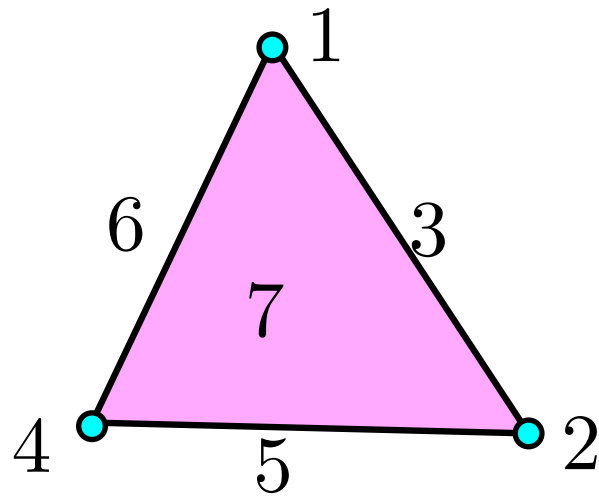
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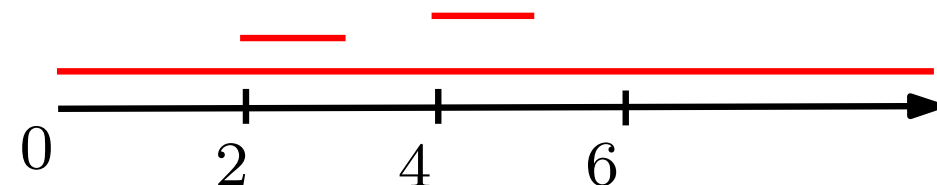
Warning: in this case, points may have multiplicity.

# Persistence diagrams

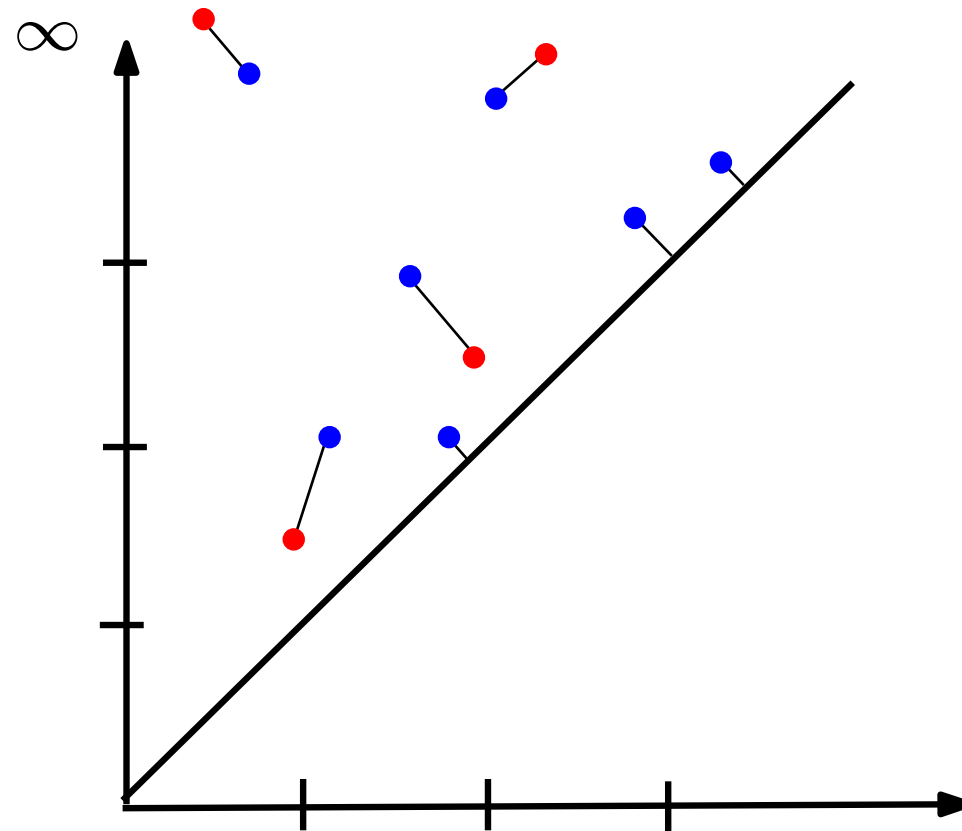


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- The diagonal  $\{y = x\}$  is added to the persistence diagram.
- Unpaired positive simplex  $\sigma^i \rightarrow (i, +\infty)$ .

**Barcodes:** an alternative (equivalent) representation where each pair  $(i, j)$  is represented by the interval  $[i, j]$



# Distance between persistence diagrams



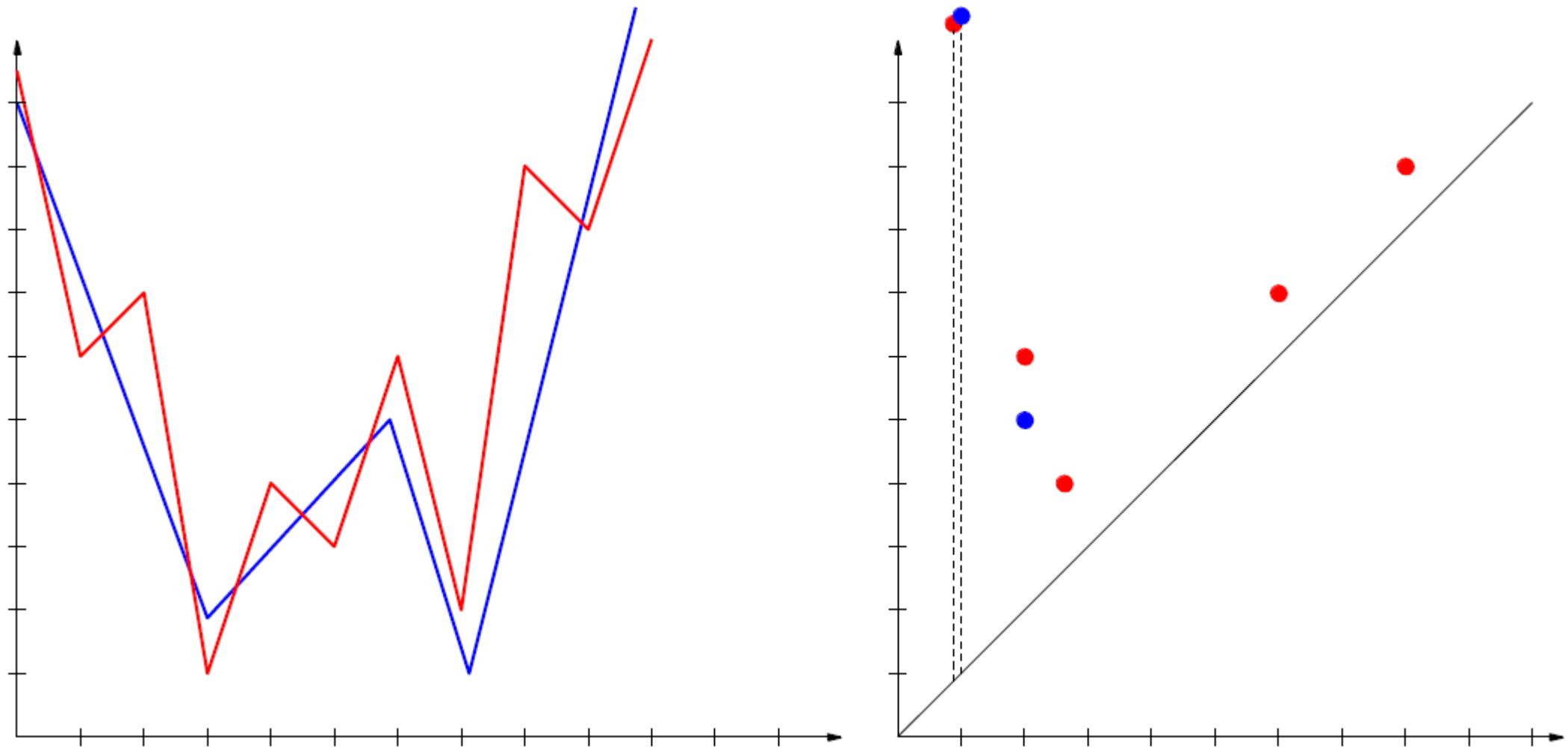
Let  $K$  be a simplicial complex and  $f, g$  two functions defined on the vertices of  $K$ . Let  $D_f$  and  $D_g$  be the persistence diagrams of  $f$  and  $g$ .

The **bottleneck distance** between  $D_f$  and  $D_g$  is

$$d_B(D_f, D_g) = \inf_{\gamma \in \Gamma} \sup_{p \in D_f} \|p - \gamma(p)\|_{\infty}$$

where  $\Gamma$  is the set of all the bijections between  $D_f$  and  $D_g$  and  $\|p - q\|_{\infty} = \max(|x_p - x_q|, |y_p - y_q|)$ .

# Stability of persistence diagrams



**Theorem:** Let  $K$  be a simplicial complex and let  $f, g : K \rightarrow \mathbb{R}$ .

$$d_B(D_f, D_g) \leq \|f - g\|_\infty$$

where  $\|f - g\|_\infty = \sup_{v \in \text{vertices}(K)} |f(v) - g(v)|$ .



# Stability of persistence diagrams

- Let  $K$  and  $K'$  be two simplicial complexes homeomorphic to a topological space  $X$ .
- Let  $\phi : K \rightarrow X$  and  $\phi' : K' \rightarrow X$  be homeomorphisms
- Let  $f : X \rightarrow \mathbb{R}$  be a continuous function and  $D_f(K)$  (resp.  $D_f(K')$ ) the persistence diagram of  $f \circ \phi$  (resp.  $f \circ \phi'$ ).

**Theorem:** Let  $\varepsilon > 0$  be such that for any simplex  $\sigma \in K$  (resp.  $\in K'$ ),  $\sup_{x,y \in \sigma} |f \circ \phi(x) - f \circ \phi(y)| < \varepsilon$  (resp.  $\sup_{x,y \in \sigma} |f \circ \phi'(x) - f \circ \phi'(y)| < \varepsilon$ ). Then one has

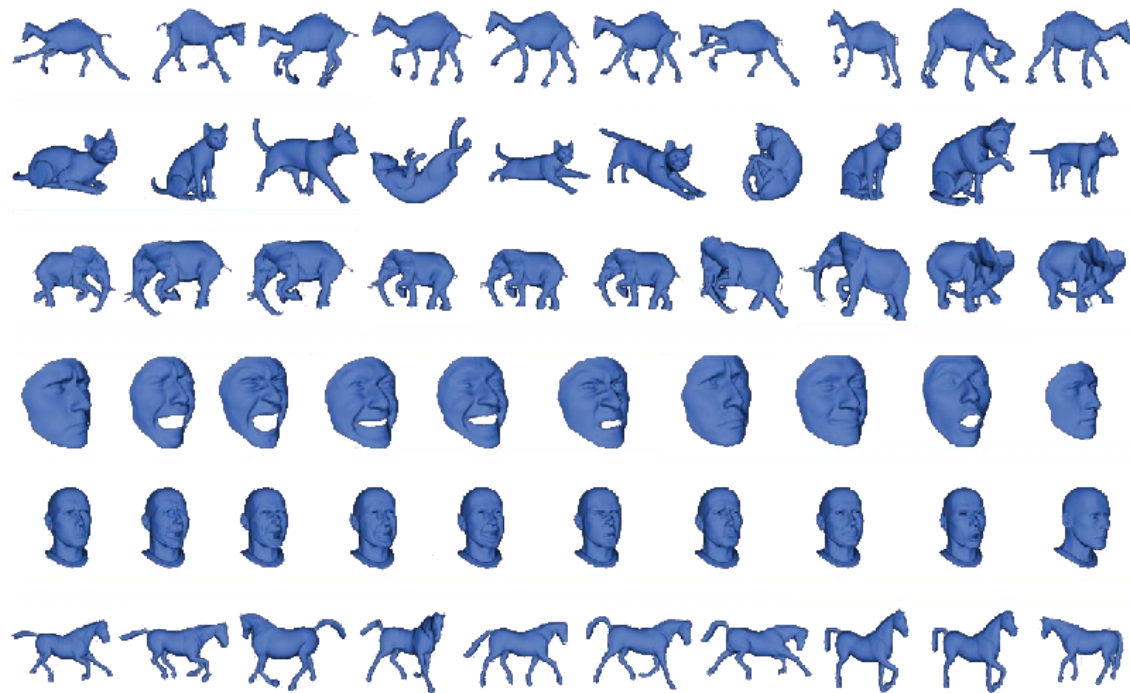
$$d_B(D_f(K), D_f(K')) \leq 2\varepsilon$$

**Remark:** this is a particular (and weaker) version of a much more general result. See:

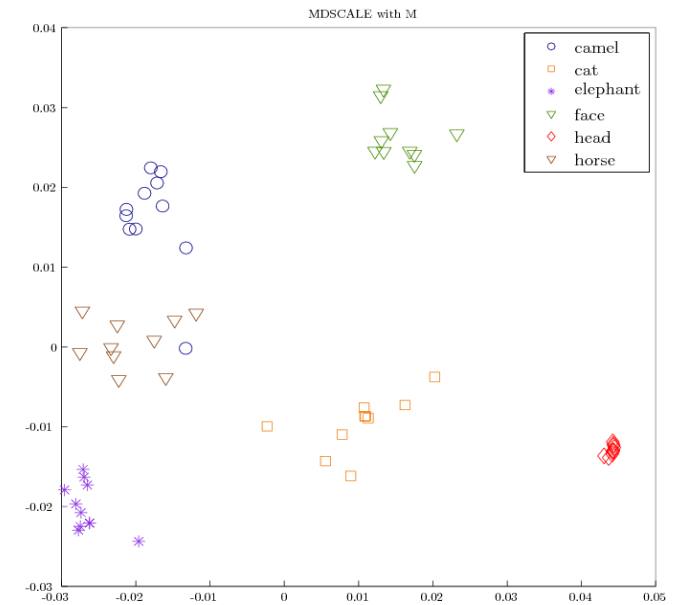
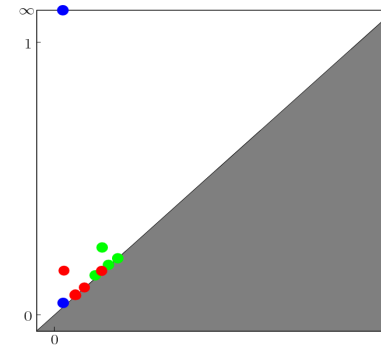
- D. Cohen-Steiner and H. Edelsbrunner and J. Harer, *Stability of Persistence Diagrams*, Proc. 21st ACM Sympos. Comput. Geom. 2005.
- F. Chazal and D. Cohen-Steiner and L. J. Guibas and M. Glisse and S. Y. Oudot, *Proximity of Persistence Modules and their Diagrams*, Proc. 25th ACM Sympos. Comput. Geom. 2009.

# Consequences of the stability

- Persistence diagrams are defined and stable for a large class of continuous functions defined over (pre-)compact metric spaces.



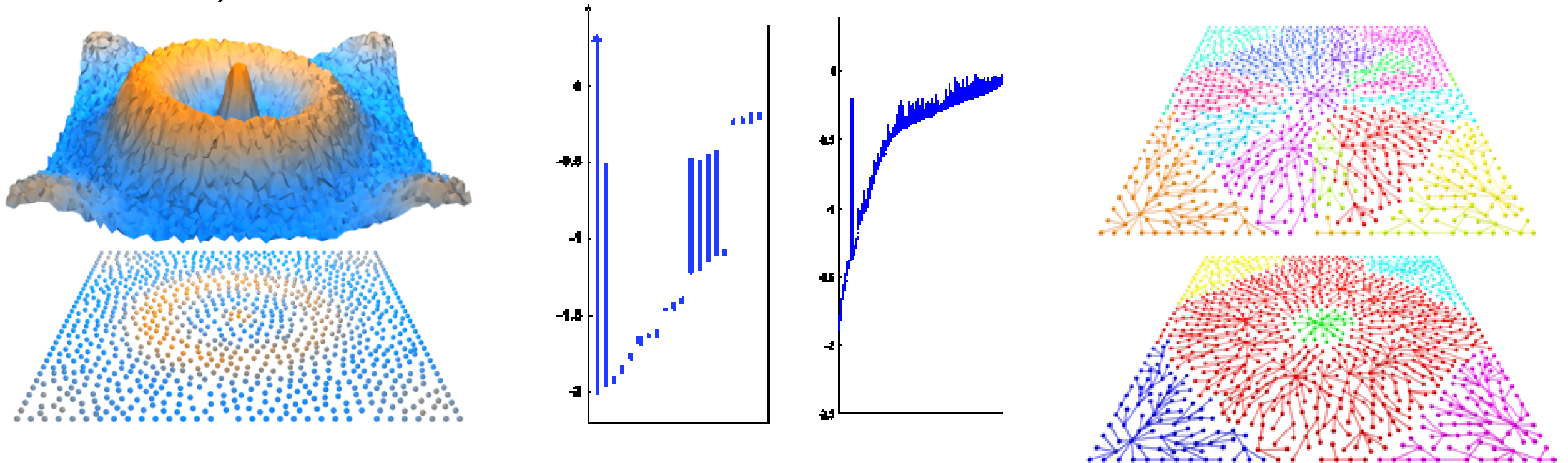
3D shapes database



- definition stable (Gromov-Hausdorff distance) topological signatures for compact metric spaces.
- Efficient algorithm to compute signatures.
- applications to shape classification.

# Consequences of the stability

- Persistence diagrams can be reliably estimated from data (functions known through a point cloud data set approximating a topological space).



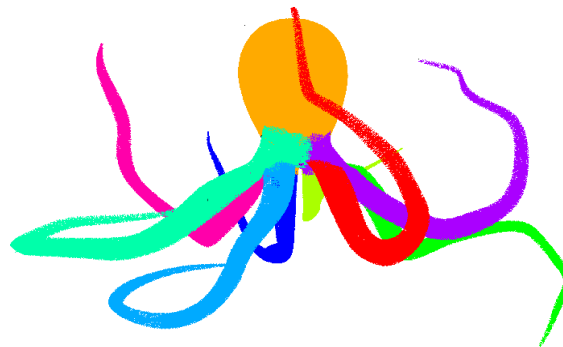
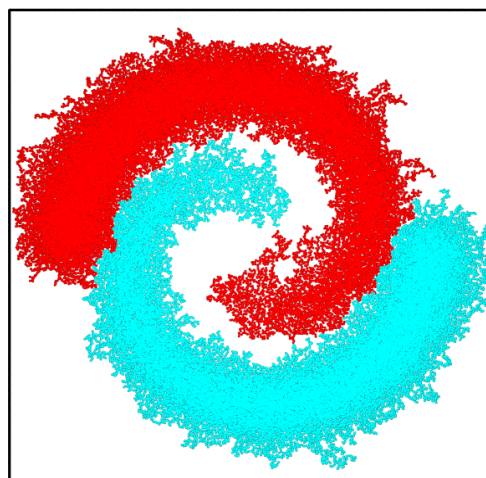
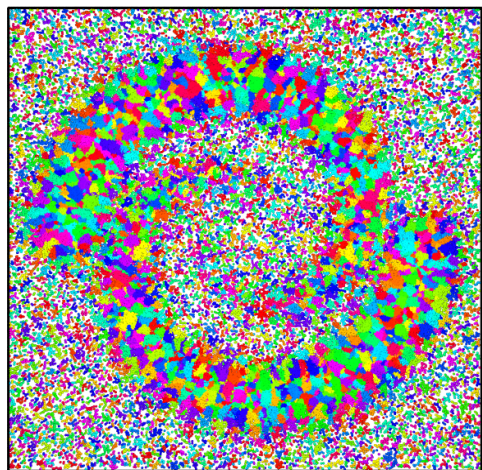
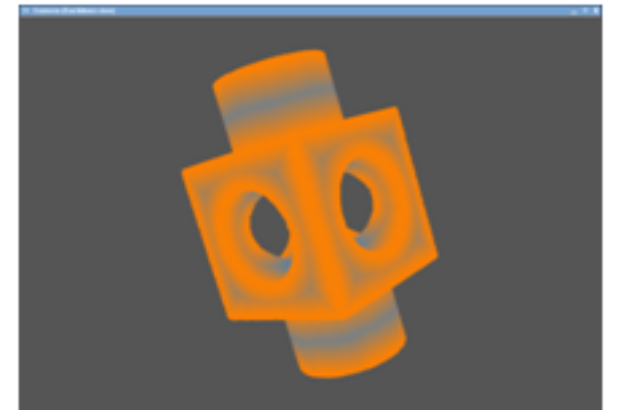
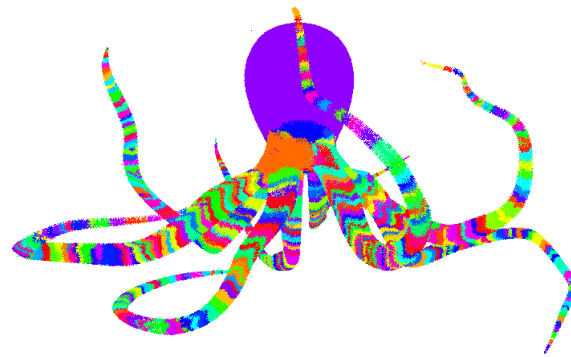
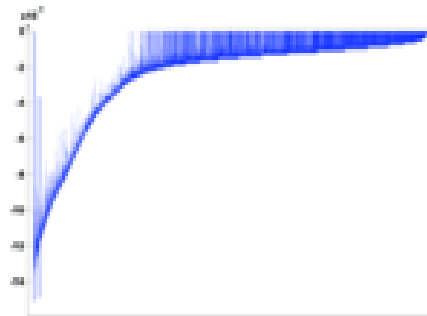
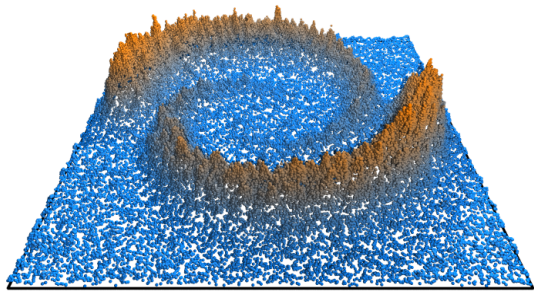
Previous approach can be generalized, leading to robust algorithms to compute the topological persistence of functions defined over point clouds sampled around unknown shapes

## Ref:

- F. Chazal, L. Guibas, S. Oudot, P. Skraba, *Analysis of Scalar Fields over Point Cloud Data*, proc. ACM Symposium on Discrete Algorithms 2009.
- F. Chazal, S. Oudot, *Toward Persistence-Based Reconstruction in Euclidean Spaces*, proc. ACM Symposium on Computational Geometry 2008.

# Consequences of the stability

- Persistence diagrams can be reliably estimated from data (functions known through a point cloud data set approximating a topological space).



Applications to clustering, segmentations, sensor networks,...

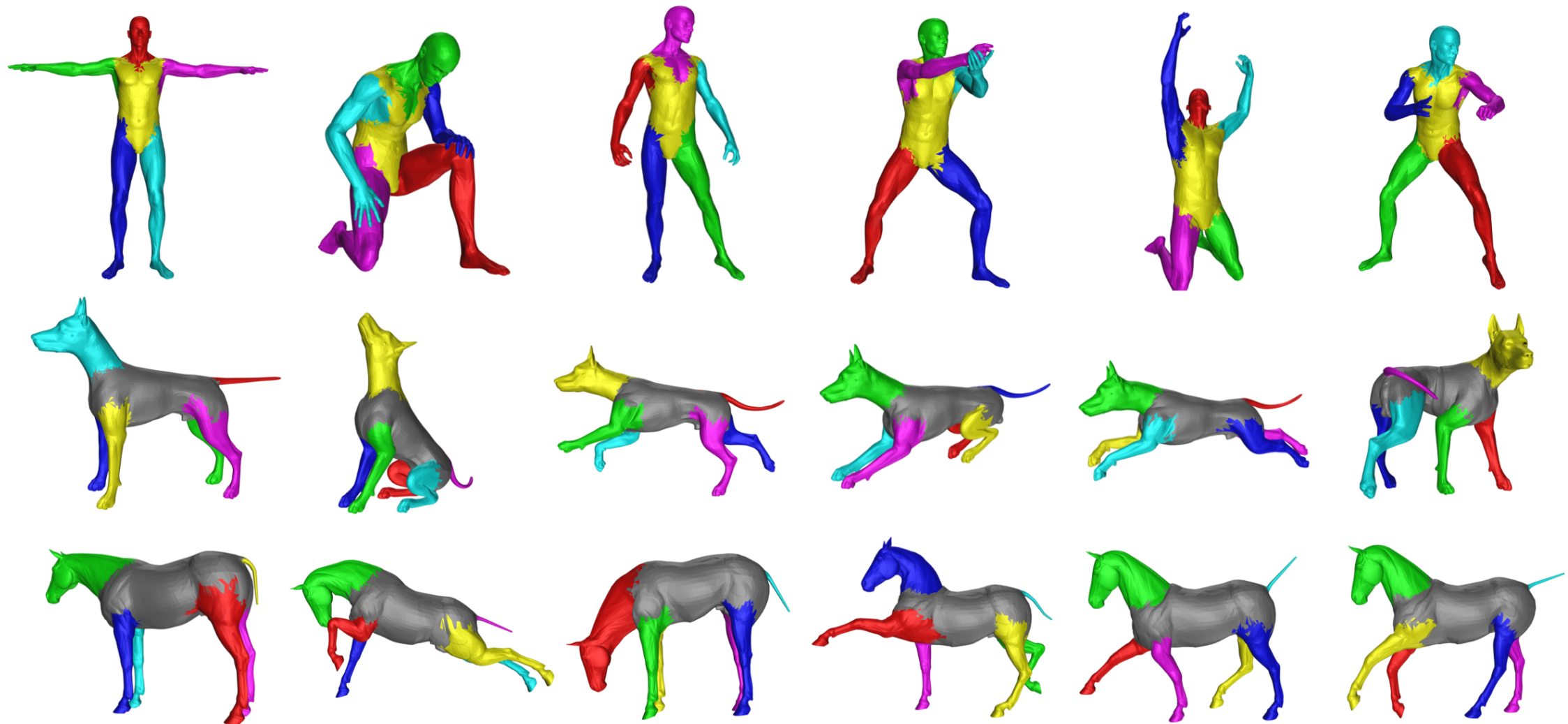
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- F. Chazal, L. Guibas, S. Oudot, P. Skraba, *Analysis of Scalar Fields over Point Cloud Data*, proc. ACM Symposium on Discrete Algorithms 2009.
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- Persistence diagrams can be reliably estimated from data (functions known through a point cloud data set approximating a topological space).



Applications to non rigid shapes segmentation

Ref:

- P. Skraba, M. Ovsjanikov, F. Chazal, L. Guibas, Persistence-Based Segmentation of Deformable Shapes, Proc. Workshop on Nonrigid Shape Analysis and Deformable Image Alignment (NORDIA), Proc. CVPR 2010