

# Persistent homology

## A new tool for data analysis

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# The shape of data

What do we mean by data ?



FIGURE – A point cloud

# What do we mean by data



FIGURE – A Data set with a loop

# What do we mean by data



FIGURE – A data set with three tendrils

# What do we mean by data



FIGURE – A data set with loop corrupted by noises



# Table des matières

- 1 The building units of our spaces
- 2 Persistent homology
- 3 Filtration
- 4 Persistent diagram and bar codes
- 5 Stability of the persistent diagram



# Motivation

## Definition : p-simplex

A  $p$ -dimensional simplex (or  $p$ -simplex)  $\sigma^p = [e_0, e_1, \dots, e_p]$  is the smallest convex set in a Euclidean space  $R^m$  containing the  $p + 1$  points  $e_0, \dots, e_p$

## Example :

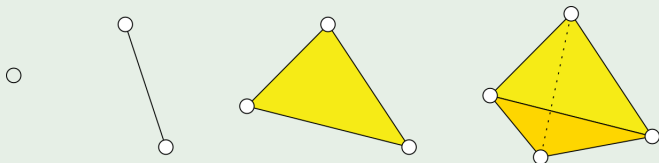


FIGURE – 0,1,2,3 simplex

## Definition : Simplicial complex

A simplicial complex is a set composed of points, line segments, triangles, and their  $p$ -dimensional counterparts.

Example :

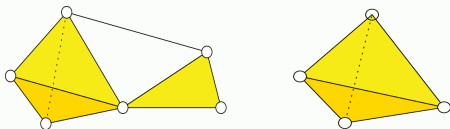


FIGURE – Simplicial complex

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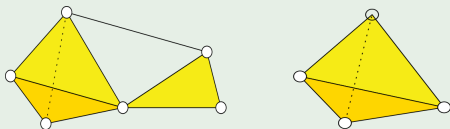


FIGURE – Simplicial complex

## Definition : Simplicial complex

A simplicial complex is a finite set of simplices satisfying the following conditions :

- 1 For all simplices  $A \in K$  with  $\alpha$  a face of  $A$ , we have  $\alpha \in K$ .
- 2  $A, B \in K \Rightarrow A, B$  are properly situated.

The dimension of a complex is the maximum dimension of the simplices contained in it.

# Motivation

Example :

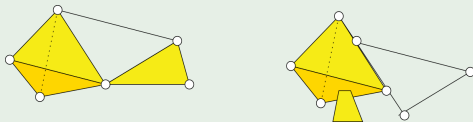


FIGURE – Simplicial complex on the left, not the case on the right

## Definition : Nerve

The nerve of a family  $D$  of subsets  $(c_i)_{i \in I}$ , denoted  $\mathcal{N}(D)$ , is the abstract simplicial complex  $\mathcal{K}$  whose elements are all sub-families  $(c_i)_{i \in J}$  such that

$$\bigcap_{i \in J} c_i \neq \emptyset$$

### Definition : Vietoris-Rips complex

Let  $(\mathcal{P}, d)$  be a metric space where  $\mathcal{P}$  is a point set. Given a real  $r > 0$ , the Vietoris-Rips complex is the abstract simplicial complex  $R^r(\mathcal{P})$  where a simplex  $\sigma \in R^r(\mathcal{P})$  if and only if  $d(p, q) \leq r$  for every pair of vertices  $p, q \in \sigma$ .

### Definition : Cech complex

The Cech complex  $C^r(\mathcal{P})$  is defined to be the nerve of the  $\frac{r}{2}$ -radius ball  $B(p, \frac{r}{2})/p \in \mathcal{P}$ ,

## Theorem ( Nerve lemma )

Let  $F$  be a finite collection of closed, convex sets in a Euclidean space. Then the nerve of  $F$  and the union of the sets in  $F$  are homotopic.

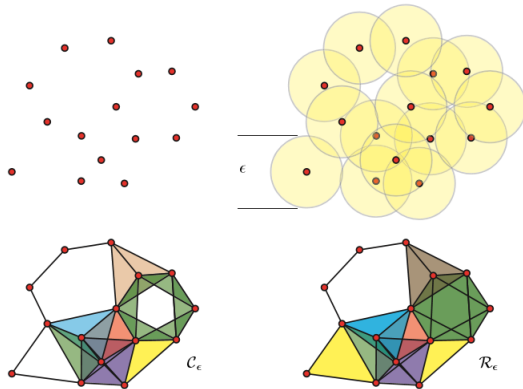


*Any data have a shape, and any shape have a meaning.*

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# Motivation



**FIGURE** – A fixed set of points [upper left] can be completed to a Cech complex [lower left] or to a Rips complex [lower right] based on a proximity parameter  $\epsilon$  [upper right].

*Which  $\varepsilon$  ?*. Converting a point cloud data set into a global complex requires a choice of parameter  $\varepsilon$ .

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## Definition

A filtration  $\mathcal{F}$  of complex  $K$  is any sequence of sub-complexes  $K_i$  of  $K$  verifying

$$\emptyset \subseteq K_1 \subset \dots \subset K_n = K.$$

# Motivation

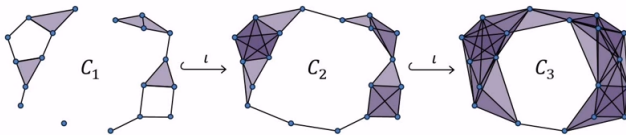


FIGURE – example of filtration

We say that a homology class is "born" in  $K_i$  when it does not exist in the image of the application induced by inclusion  $K_{i-1} \subset K_i$ .

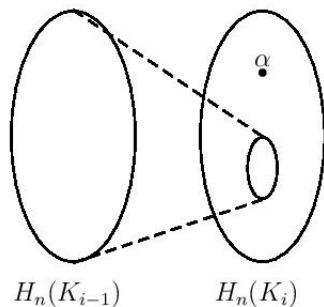


FIGURE – Birth of a homology class



It will be said that it "dies" in  $K_j$  when it does not exist in the image of the application induced by inclusion  $K_{i-1} \subset K_{j-1}$  but that it exists in the image of the induced application  $K_{i-1} \subset K_j$ .

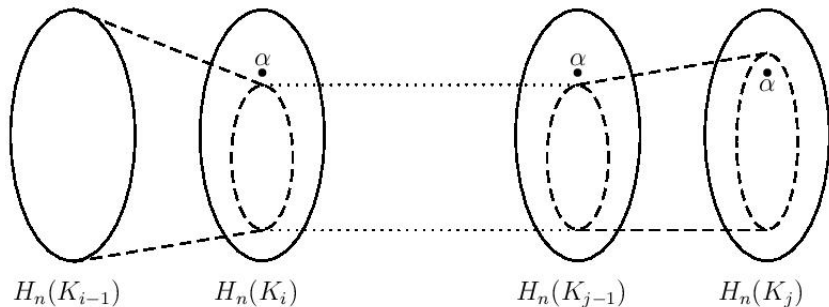
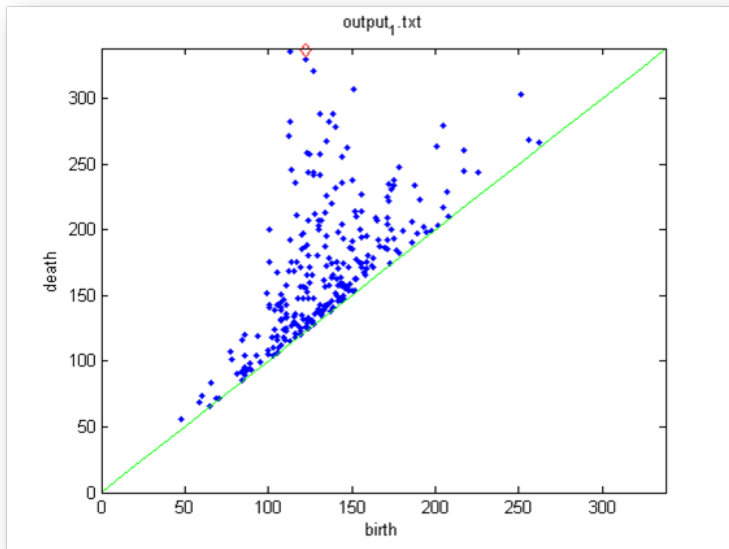


FIGURE – Death of a homology class

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Persistent diagrams summarize this information as *two – dimensional* point sets with multiplicities.



given a map  $f : X \rightarrow \mathbb{R}$  and a real  $t$ , we define  $R_t$  as the preimage of  $] -\infty; t]$ , ( $R_t = f^{-1} ] -\infty; t]$ ) the idea of persistent consists in calculating the homology groups  $H_n(R_t)$  as a function of  $t$ .

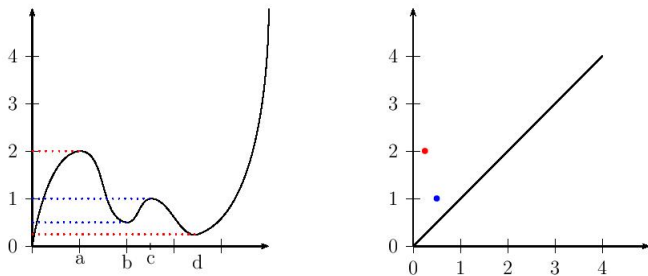


FIGURE – The graph and the associated diagram of  $f$

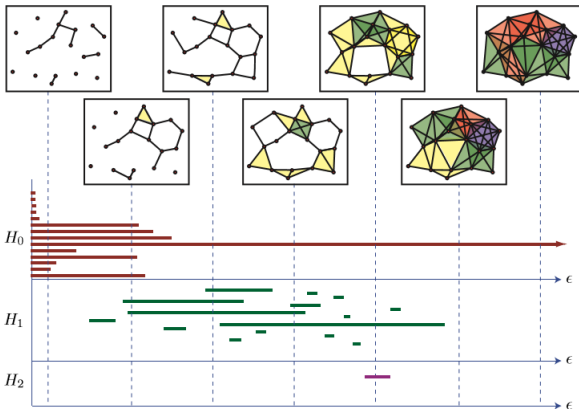


FIGURE – Example of bar codes

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## Definition : Bottleneck distance

Let  $X$  be a topological space with two continuous functions  $f, g : X \Rightarrow R$  and let  $Dgm(f), Dgm(g)$  their respective diagrams. We define their bottleneck distance by taking the infimum over all supremums

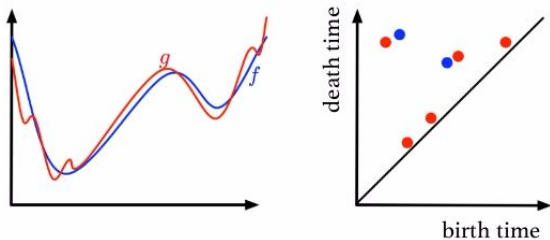
$$d_B(Dgm(f), Dgm(g)) := \inf_{\eta} \sup_x \|x - \eta(x)\|_{\infty},$$

where  $x \in Dgm(f)$  and  $\eta : Dgm(f) \rightarrow Dgm(g)$  ranges over all bijections, while  $\|x - y\|_{\infty} = \max|x|, |y|$  is the usual  $L_{\infty}$  norm.

## Theorem

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





**FIGURE** – Left : two functions with small distance. Right : the corresponding two persistence diagrams with small bottleneck distance.

*Thank YOU .*