

Dynamical Systems & Applications — 03/10/23

An introduction to Topological Data Analysis

Part III/IV: Persistent homology

<https://raphaeltinarrage.github.io>

Part I/IV: Topological invariants

Tuesday 26th

Part II/IV: Homology

Thursday 28th

Part III/IV: Persistent Homology

Tuesday 3rd

Part IV/IV: Python tutorial

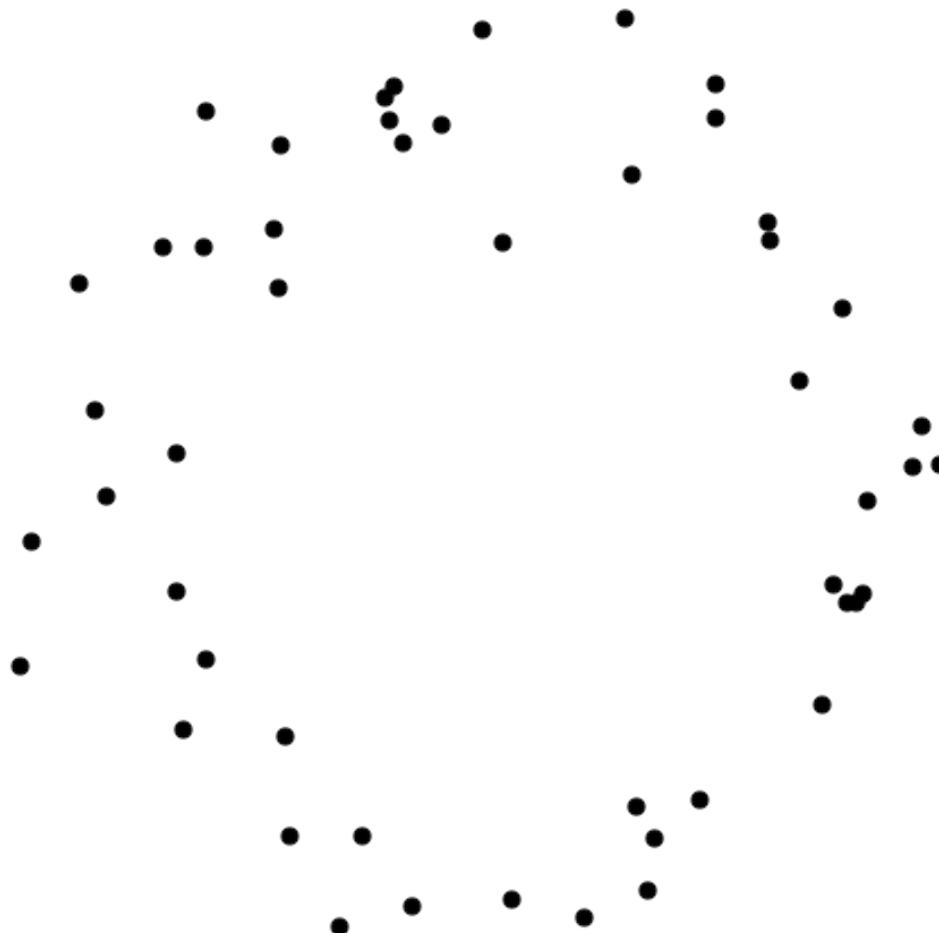
Thursday 5th

O problema da inferência homológica 3/46 (1/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

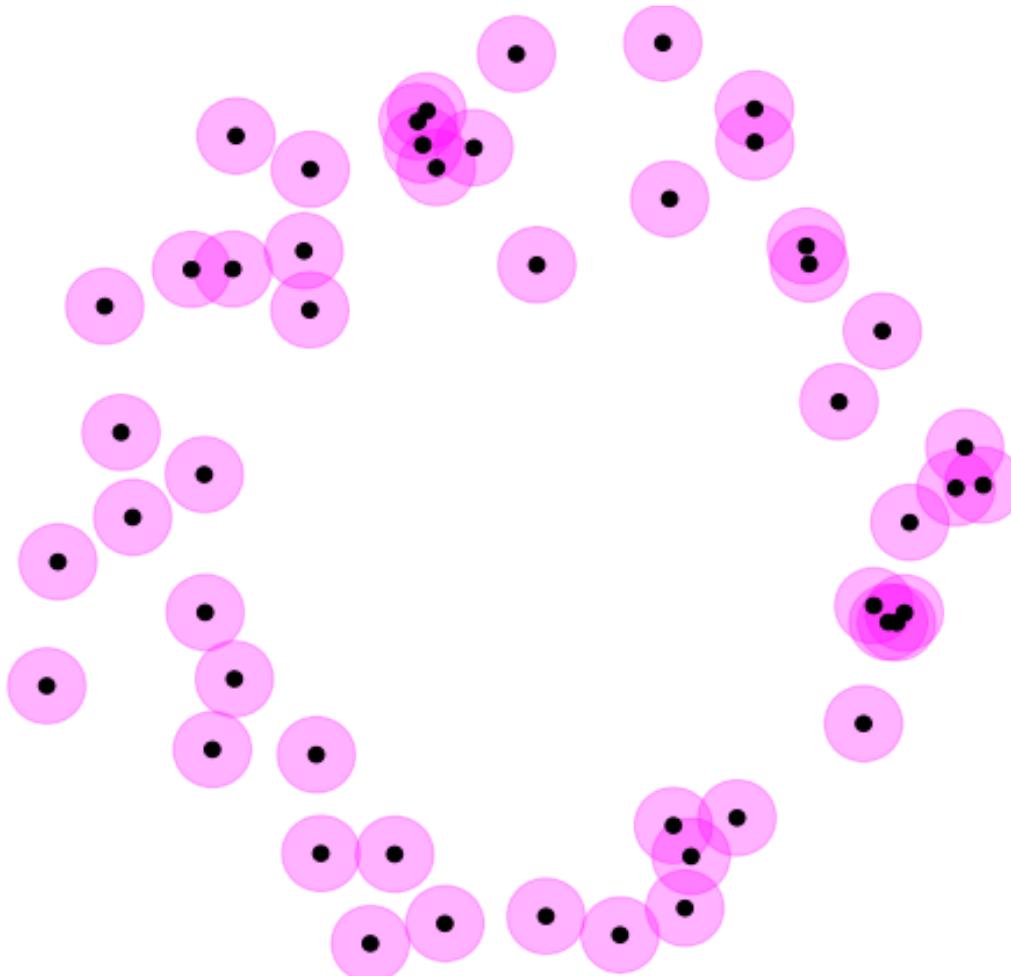


O problema da inferência homológica 3/46 (2/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

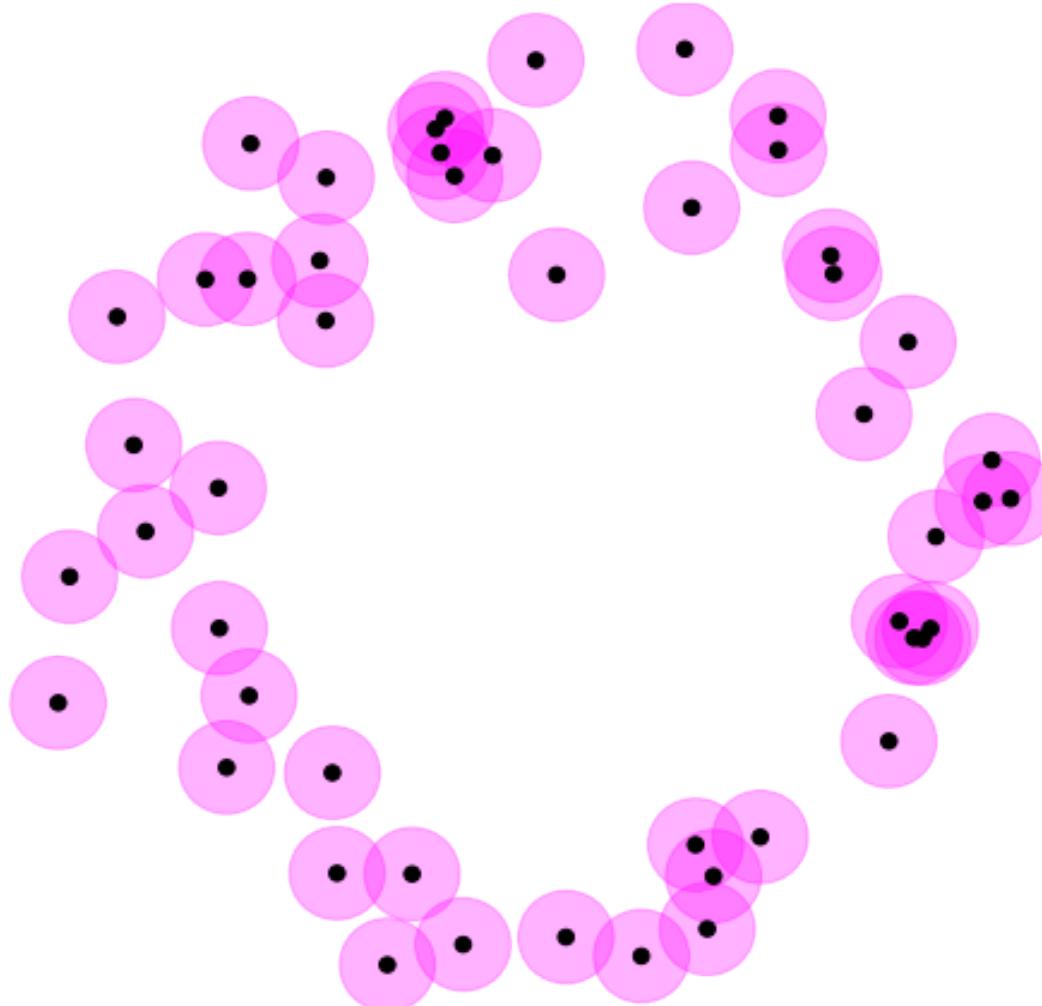


O problema da inferência homológica 3/46 (3/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

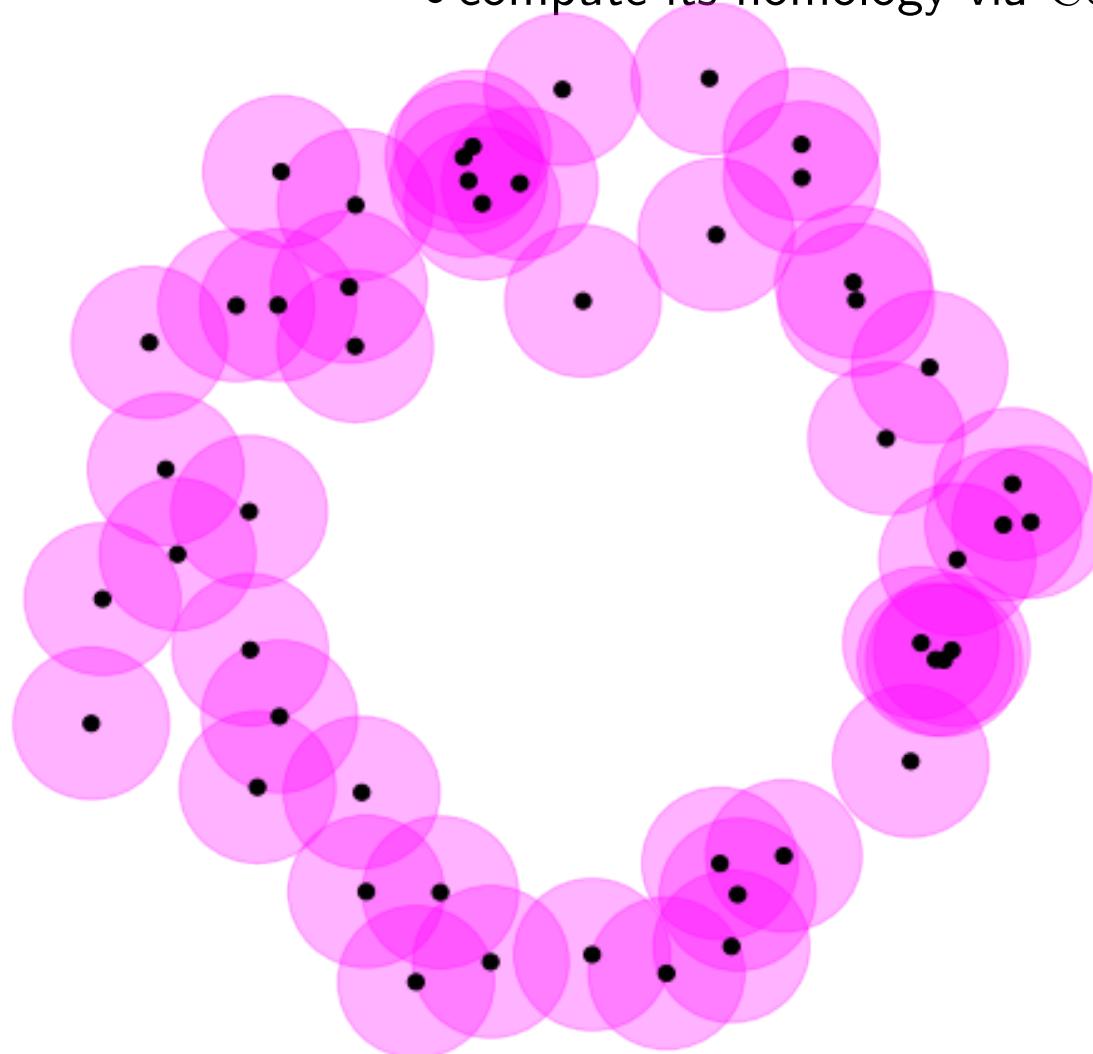


O problema da inferência homológica 3/46 (4/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

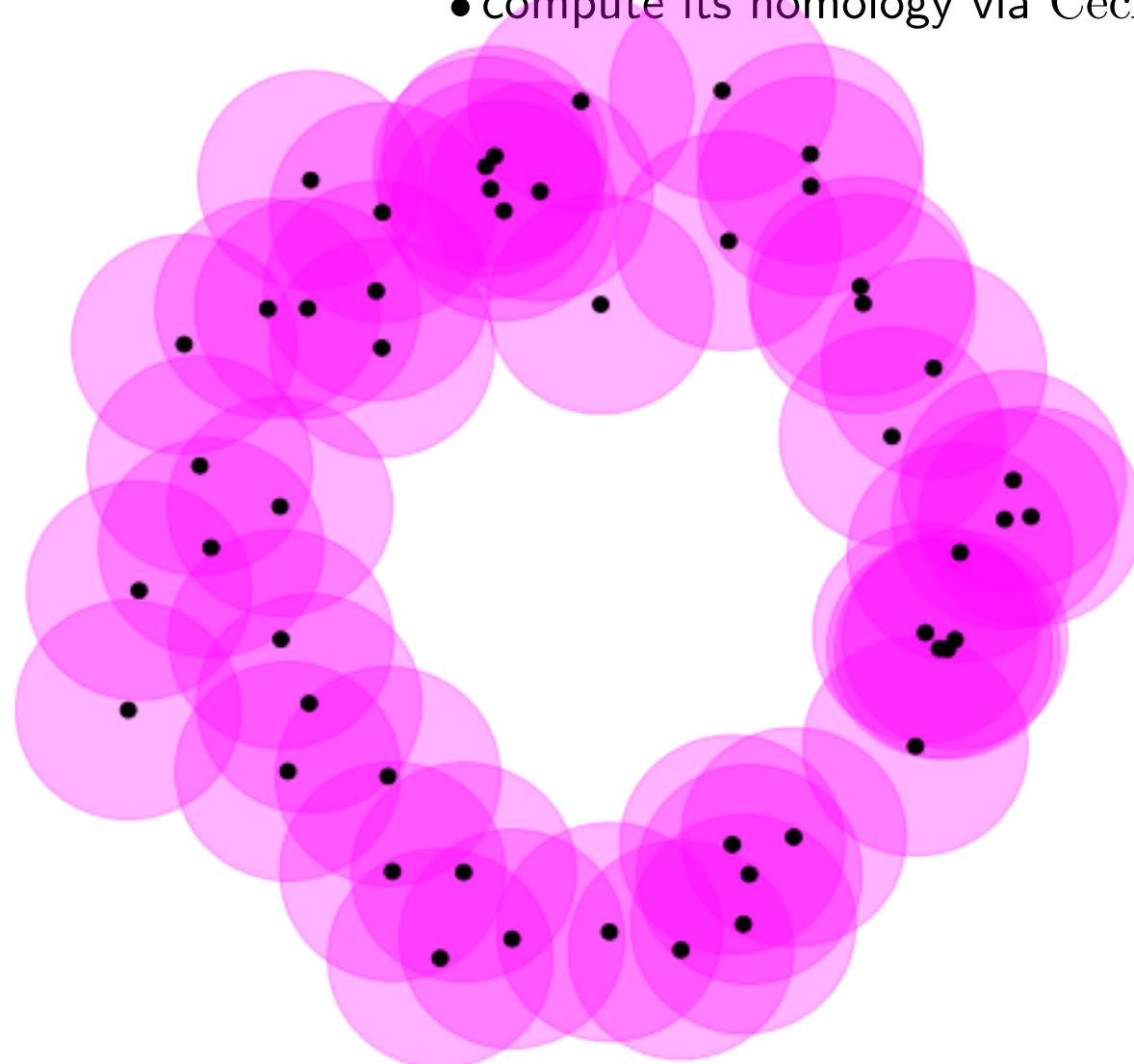


O problema da inferência homológica 3/46 (5/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

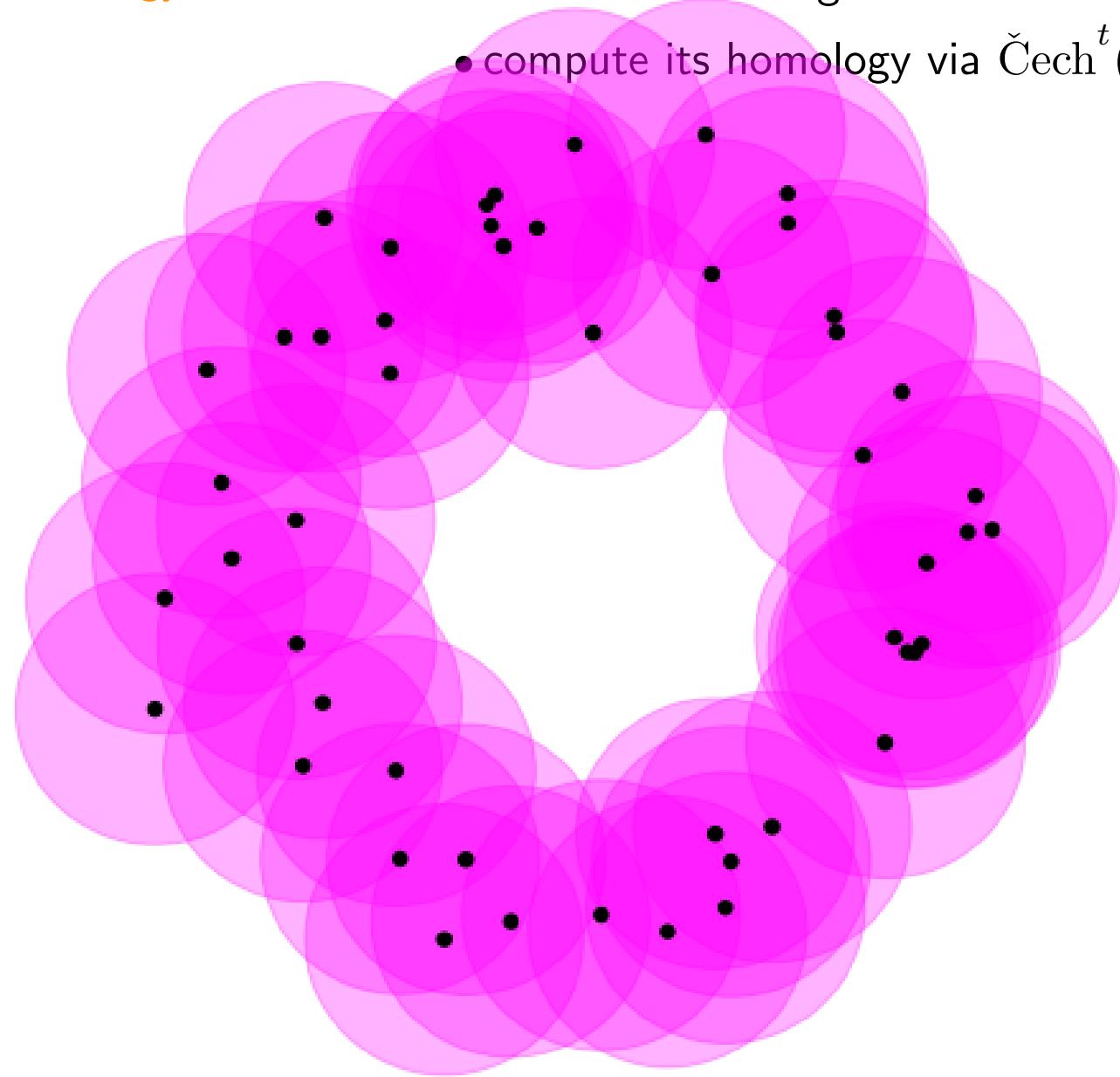


O problema da inferência homológica 3/46 (6/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

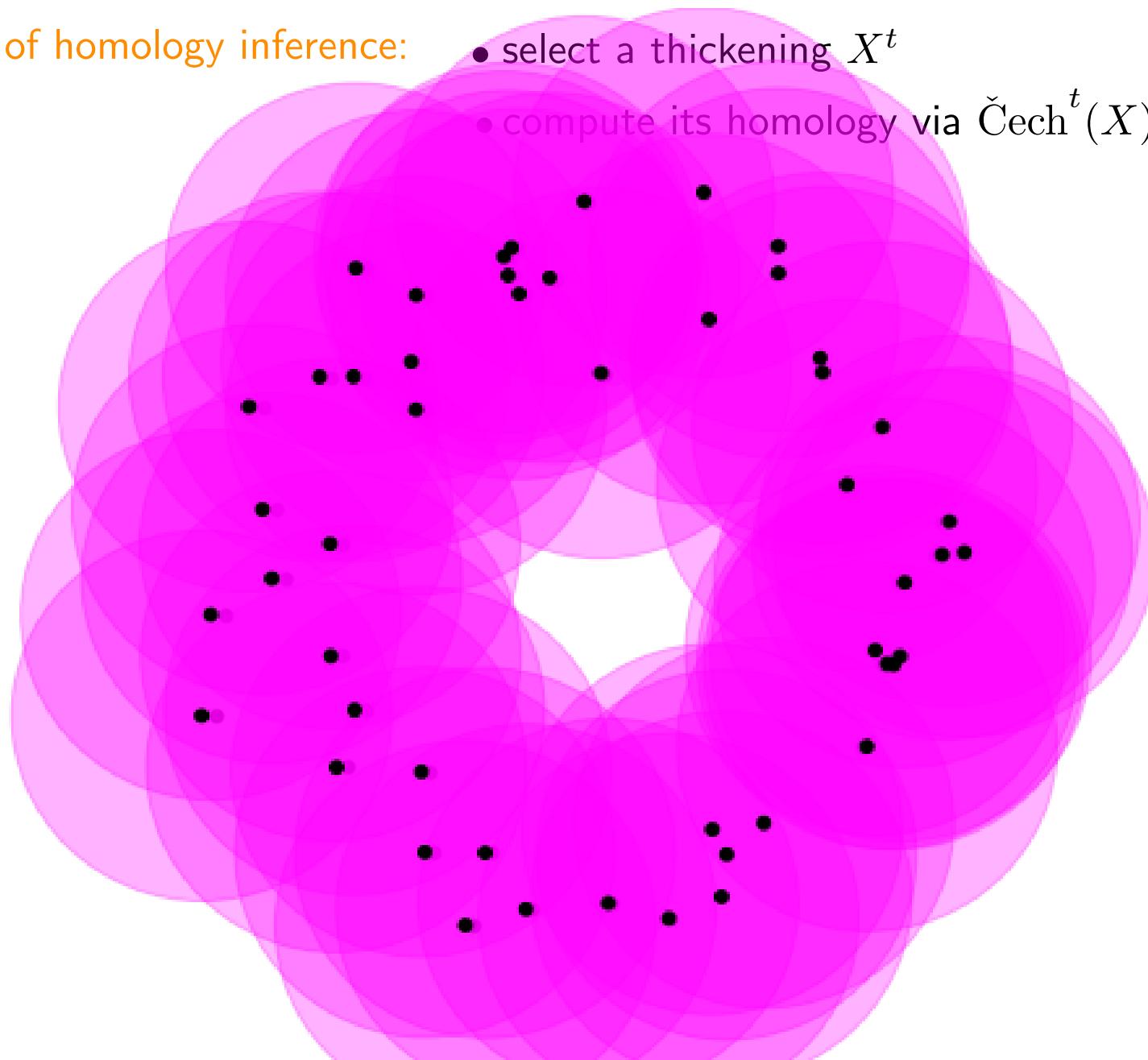


O problema da inferência homológica 3/46 (7/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

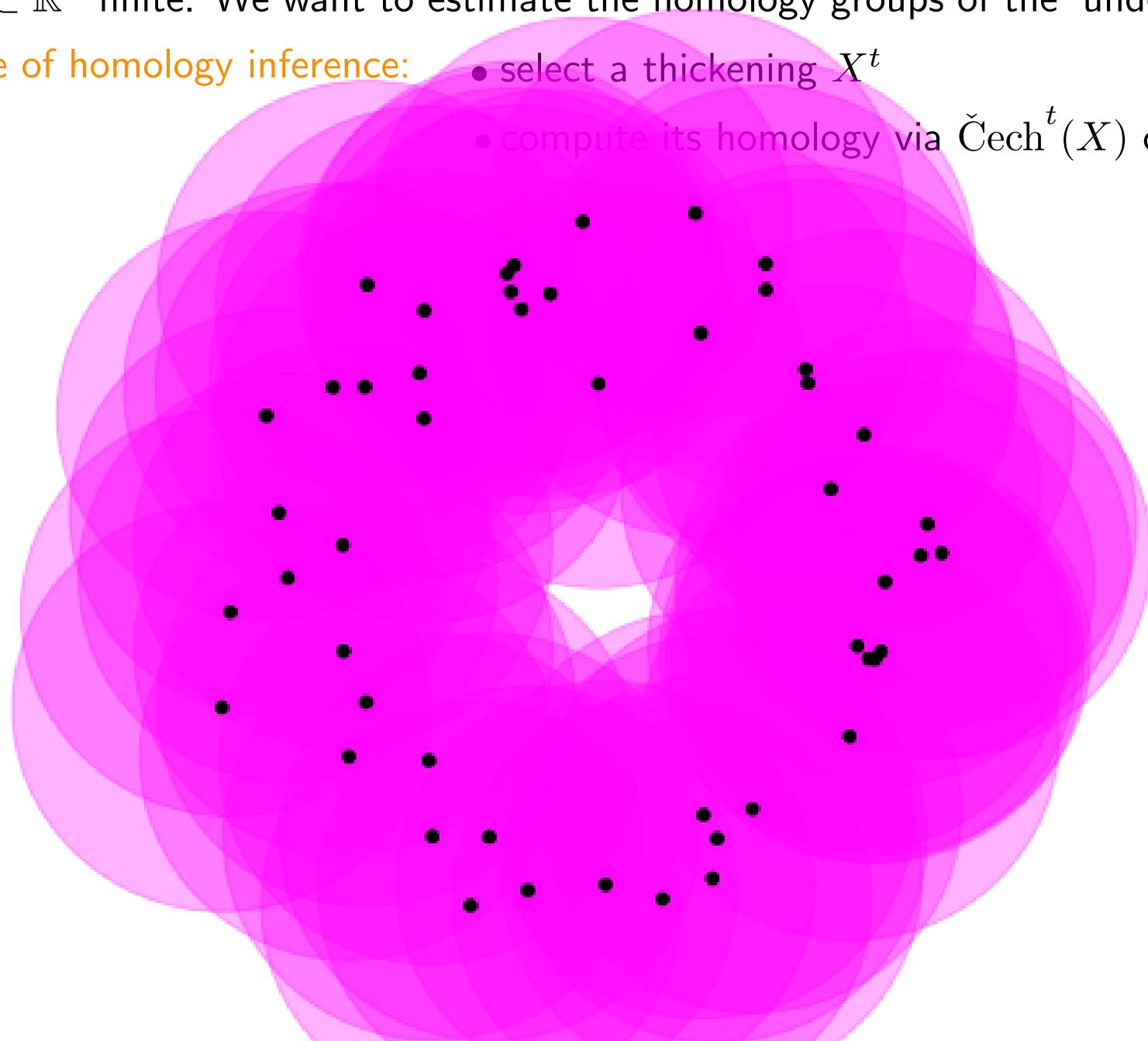


O problema da inferência homológica 3/46 (8/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

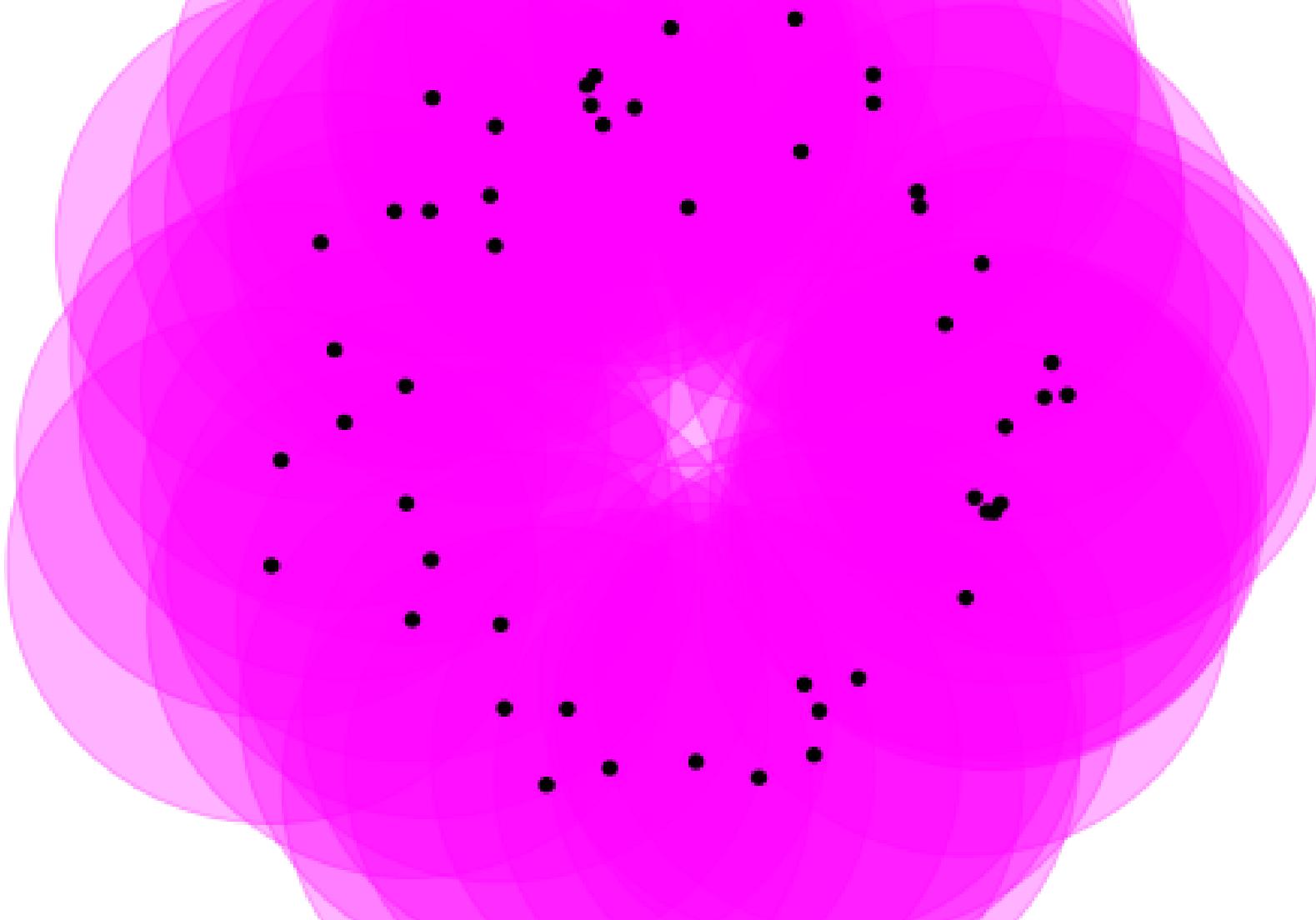


O problema da inferência homológica 3/46 (9/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$



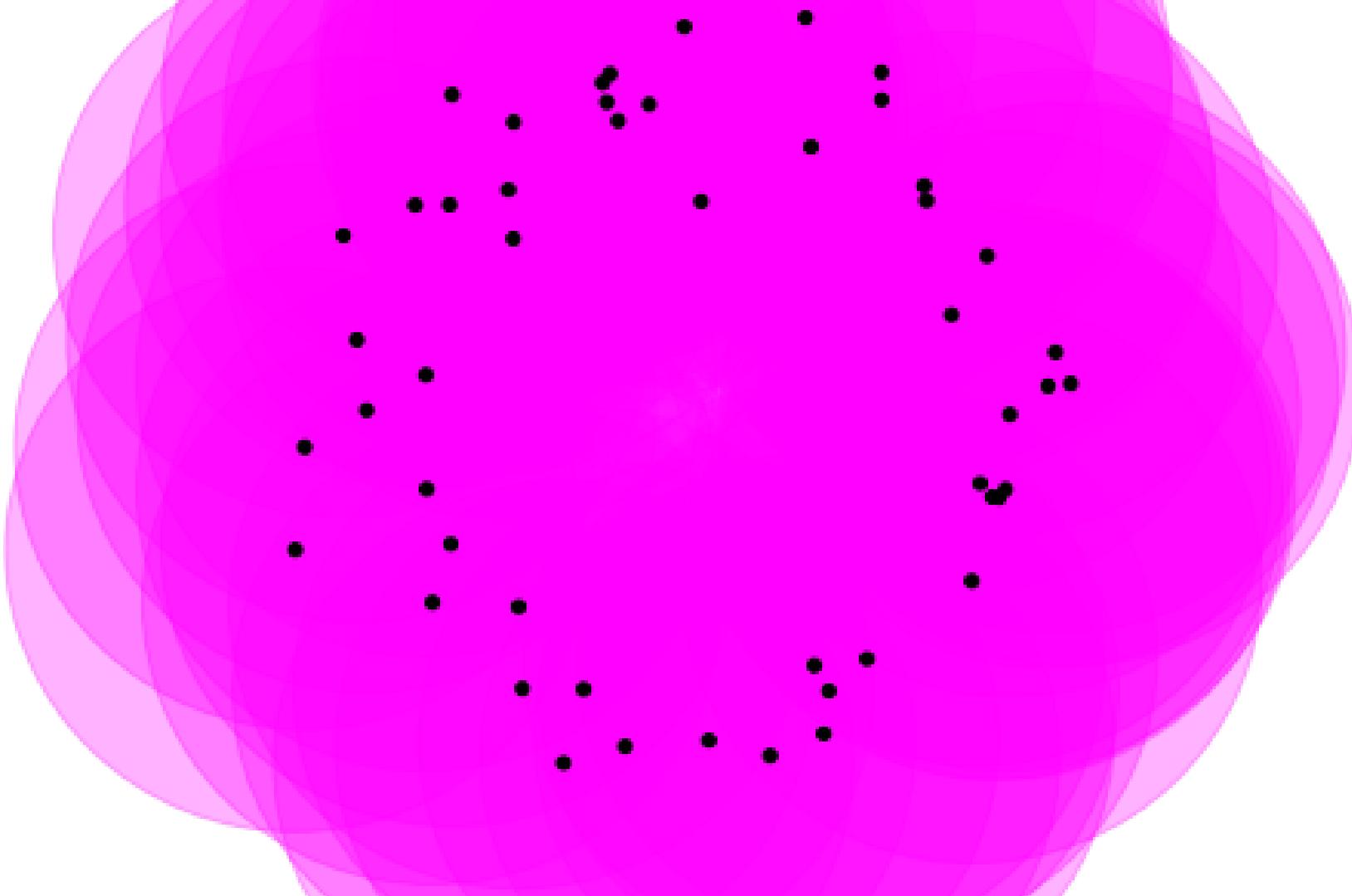
O problema da inferência homológica

3/46 (10/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t
- compute its homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$



O problema da inferência homológica

3/46 (11/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t

• compute homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$



O problema da inferência homológica

3/46 (12/13)

Let $X \subset \mathbb{R}^n$ finite. We want to estimate the homology groups of the ‘underlying shape’.

Pipeline of homology inference:

- select a thickening X^t

• compute homology via $\check{\text{C}}\text{ech}^t(X)$ or $\text{Rips}^t(X)$

How to choose t ?

O problema da inferência homológica

3/46 (13/13)



Data analyst

Persistence theory

I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

We have seen that (singular) homology transforms *topological spaces* into *vector spaces*

$$\begin{aligned} H_i: \text{Top} &\longrightarrow \text{Vect} \\ X &\longmapsto H_i(X) \end{aligned}$$

and transforms *continuous maps* into *linear maps*

$$(f: X \rightarrow Y) \longmapsto (f_*: H_i(X) \rightarrow H_i(Y))$$

We will adopt a simplicial point of view (simplicial homology).

$$\begin{aligned} H_i: \text{SimpComp} &\longrightarrow \text{Vect} \\ K &\longmapsto H_i(K) \\ (f: K \rightarrow L) &\longmapsto (f_*: H_i(K) \rightarrow H_i(L)) \end{aligned}$$

Functorialidade singular

6/46 (2/2)

We have seen that (singular) homology transforms *topological spaces* into *vector spaces*

$$\begin{aligned} H_i : \text{Top} &\longrightarrow \text{Vect} \\ X &\longmapsto H_i(X) \end{aligned}$$

and transforms *continuous maps* into *linear maps*

$$(f : X \rightarrow Y) \longmapsto (f_* : H_i(X) \rightarrow H_i(Y))$$

We will adopt a simplicial point of view (simplicial homology).

$$\begin{aligned} H_i : \text{SimpComp} &\longrightarrow \text{Vect} \\ K &\longmapsto H_i(K) \\ (f : K \rightarrow L) &\longmapsto (f_* : H_i(K) \rightarrow H_i(L)) \end{aligned}$$

what is a map between simplicial complexes?

Mapas simpliciais

7/46 (1/3)

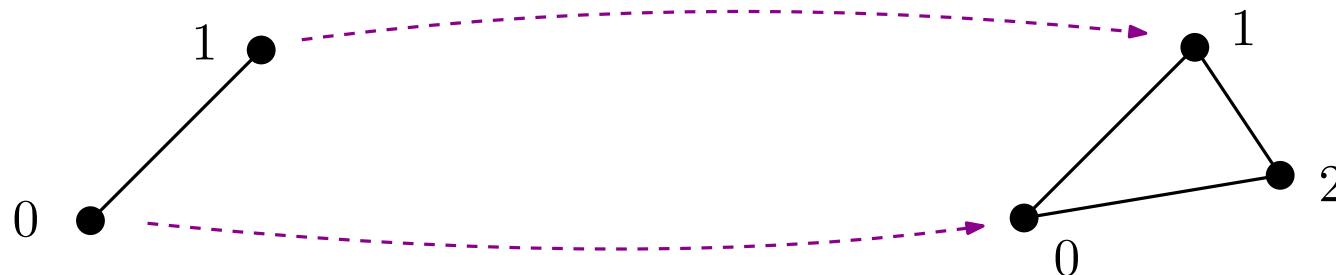
Definition: Let K and L be two simplicial complexes, and V_K, V_L their set of vertices. A **simplicial map** between K and L is a map $f: V_K \rightarrow V_L$ such that

$$\forall \sigma \in K, f(\sigma) \in L.$$

We may denote a simplicial map $f: K \rightarrow L$ instead of $f: V_K \rightarrow V_L$.

Example: Let $K = \{[0], [1], [0, 1]\}$, $L = \{[0], [1], [2], [0, 1], [0, 2], [1, 2]\}$ and

$$\begin{aligned}f: \{0, 1\} &\rightarrow \{0, 1, 2\} \\0 &\mapsto 0 \\1 &\mapsto 1\end{aligned}$$



It is simplicial since $f([0, 1]) = [0, 1]$ is a simplex of L .

Mapas simpliciais

7/46 (2/3)

Definition: Let K and L be two simplicial complexes, and V_K, V_L their set of vertices. A **simplicial map** between K and L is a map $f: V_K \rightarrow V_L$ such that

$$\forall \sigma \in K, f(\sigma) \in L.$$

We may denote a simplicial map $f: K \rightarrow L$ instead of $f: V_K \rightarrow V_L$.

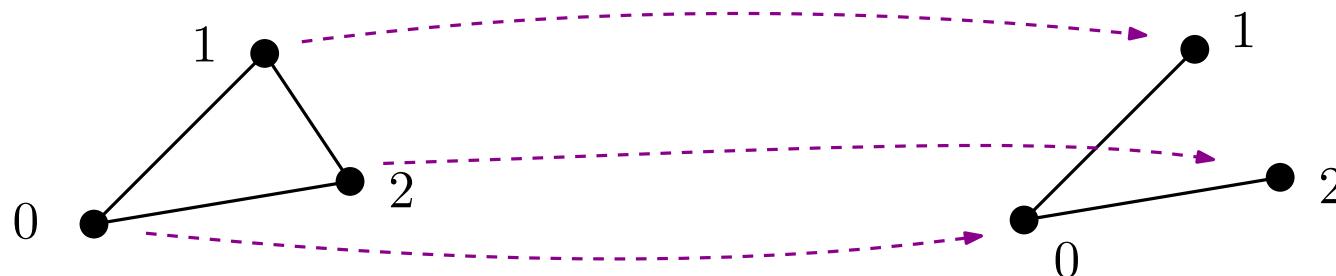
Example: Let $K = \{[0], [1], [2], [0, 1], [0, 2], [1, 2]\}$, $L = \{[0], [1], [2], [0, 1], [0, 2]\}$ and

$$f: \{0, 1\} \rightarrow \{0, 1, 2\}$$

$$0 \mapsto 0$$

$$1 \mapsto 1$$

$$2 \mapsto 2$$



It is not simplicial since $f([1, 2]) = [1, 2]$ is not a simplex of L .

Mapas simpliciais

7/46 (3/3)

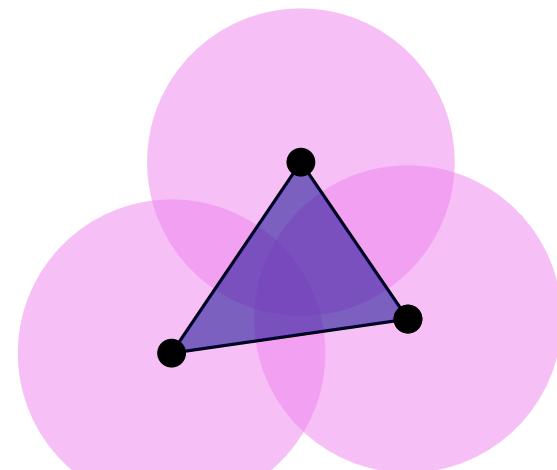
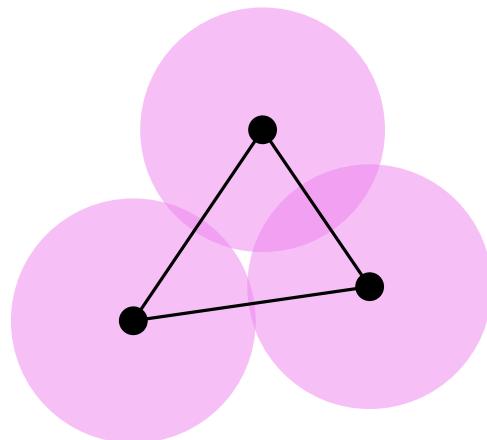
Definition: Let K and L be two simplicial complexes, and V_K, V_L their set of vertices. A **simplicial map** between K and L is a map $f: V_K \rightarrow V_L$ such that

$$\forall \sigma \in K, f(\sigma) \in L.$$

We may denote a simplicial map $f: K \rightarrow L$ instead of $f: V_K \rightarrow V_L$.

Example: Let $X \subset \mathbb{R}^n$ and $s, t \geq 0$ such that $s \leq t$. Consider the Čech complexes $\check{\text{Cech}}^s(X)$ and $\check{\text{Cech}}^t(X)$.

The inclusion map $i: \check{\text{Cech}}^s(X) \rightarrow \check{\text{Cech}}^t(X)$ is a simplicial map.



Indeed, the sequence of simplicial complexes $(\check{\text{Cech}}^t(X))_{t \geq 0}$ is non-decreasing.

Reminder: $\check{\text{Cech}}^s(X)$ is the nerve of the cover given by the balls $\{B(x, s), x \in X\}$.

Let $f: K \rightarrow L$ be a simplicial map. Let $n \geq 0$, and consider the groups of chains of K and L :

$$C_n(K) = \left\{ \sum_{\sigma \in K_{(n)}} \epsilon_\sigma \cdot \sigma \mid \forall \sigma \in K_{(n)}, \epsilon_\sigma \in \mathbb{Z}/2\mathbb{Z} \right\}$$

$$C_n(L) = \left\{ \sum_{\sigma \in K_{(n)}} \epsilon_\sigma \cdot \sigma \mid \forall \sigma \in L_{(n)}, \epsilon_\sigma \in \mathbb{Z}/2\mathbb{Z} \right\}$$

We define a linear map as follows:

$$f_n: C_n(K) \longrightarrow C_n(L)$$

$$\begin{aligned} \sigma &\longmapsto f(\sigma) \text{ if } \dim(f(\sigma)) = n, \\ &0 \text{ else.} \end{aligned}$$

Mapa linear induzido

8/46 (2/11)

Let $f: K \rightarrow L$ be a simplicial map. Let $n \geq 0$, and consider the groups of chains of K and L :

$$C_n(K) = \left\{ \sum_{\sigma \in K_{(n)}} \epsilon_\sigma \cdot \sigma \mid \forall \sigma \in K_{(n)}, \epsilon_\sigma \in \mathbb{Z}/2\mathbb{Z} \right\}$$

$$C_n(L) = \left\{ \sum_{\sigma \in L_{(n)}} \epsilon_\sigma \cdot \sigma \mid \forall \sigma \in L_{(n)}, \epsilon_\sigma \in \mathbb{Z}/2\mathbb{Z} \right\}$$

We define a linear map as follows:

$$f_n: C_n(K) \longrightarrow C_n(L)$$

$$\sigma \longmapsto f(\sigma) \text{ if } \dim(f(\sigma)) = n,$$

$$0 \text{ else.}$$

$$\begin{array}{ccccccc}
 \dashrightarrow & C_3(K) & \xrightarrow{\partial_3} & C_2(K) & \xrightarrow{\partial_2} & C_1(K) & \xrightarrow{\partial_1} C_0(K) \xrightarrow{\partial_0} \{0\} \\
 & \downarrow f_3 & & \downarrow f_2 & & \downarrow f_1 & & \downarrow f_0 & & \downarrow f_{-1} \\
 \dashrightarrow & C_3(L) & \xrightarrow{\partial_3} & C_2(L) & \xrightarrow{\partial_2} & C_1(L) & \xrightarrow{\partial_1} C_0(L) \xrightarrow{\partial_0} \{0\}
 \end{array}$$

Mapa linear induzido

8/46 (3/11)

$$\begin{array}{ccccccc}
 \dashrightarrow & C_3(K) & \xrightarrow{\partial_3} & C_2(K) & \xrightarrow{\partial_2} & C_1(K) & \xrightarrow{\partial_1} & C_0(K) & \xrightarrow{\partial_0} & \{0\} \\
 & \downarrow f_3 & & \downarrow f_2 & & \downarrow f_1 & & \downarrow f_0 & & \downarrow f_{-1} \\
 \dashrightarrow & C_3(L) & \xrightarrow{\partial_3} & C_2(L) & \xrightarrow{\partial_2} & C_1(L) & \xrightarrow{\partial_1} & C_0(L) & \xrightarrow{\partial_0} & \{0\}
 \end{array}$$

Lemma: For every $n \geq 0$, we have $\partial_n \circ f_n = f_{n-1} \circ \partial_n$.

Proof: Let $\sigma \in K_{(n)}$. We have the equalities

$$\partial_n \circ f_n(\sigma) = \sum_{\substack{\mu \subset f(\sigma) \\ |\mu|=|\sigma|-1}} \mu$$

$$f_{n-1} \circ \partial_n(\sigma) = \sum_{\substack{\tau \subset \sigma \\ |\tau|=|\sigma|-1}} f_n(\tau)$$

We should distinguish three cases:

- $|f(\sigma)| = |\sigma|$ (i.e. f is injective on σ),
- $|f(\sigma)| < |\sigma| - 1$,
- $|f(\sigma)| = |\sigma| - 1$.

Mapa linear induzido

8/46 (4/11)

$$\begin{array}{ccccccc}
 \dashrightarrow & C_3(K) & \xrightarrow{\partial_3} & C_2(K) & \xrightarrow{\partial_2} & C_1(K) & \xrightarrow{\partial_1} & C_0(K) & \xrightarrow{\partial_0} & \{0\} \\
 & \downarrow f_3 & & \downarrow f_2 & & \downarrow f_1 & & \downarrow f_0 & & \downarrow f_{-1} \\
 \dashrightarrow & C_3(L) & \xrightarrow{\partial_3} & C_2(L) & \xrightarrow{\partial_2} & C_1(L) & \xrightarrow{\partial_1} & C_0(L) & \xrightarrow{\partial_0} & \{0\}
 \end{array}$$

Lemma: For every $n \geq 0$, we have $\partial_n \circ f_n = f_{n-1} \circ \partial_n$.

Proposition: For every $c \in Z_n(K)$, we have $f_n(c) \in Z_n(L)$.
For every $c \in B_n(K)$, we also have $f_n(c) \in B_n(L)$.

Proof: First, let $c \in Z_n(K)$. We have

$$\partial_n \circ f_n(c) = f_{n-1} \circ \partial_n(c) = f_{n-1}(0) = 0,$$

hence $f_n(c) \in Z_n(L)$.

Secondly, let $c \in B_n(K)$, and write $c = \partial_{n+1}(c')$ with $c' \in C_{n+1}(K)$. We get

$$f_n(c) = f_n \circ \partial_{n+1}(c') = \partial_{n+1} \circ f_{n+1}(c'),$$

hence $f_n(c) \in B_n(L)$.

Mapa linear induzido

8/46 (5/11)

$$\begin{array}{ccccccc}
 \dashrightarrow & C_3(K) & \xrightarrow{\partial_3} & C_2(K) & \xrightarrow{\partial_2} & C_1(K) & \xrightarrow{\partial_1} & C_0(K) & \xrightarrow{\partial_0} & \{0\} \\
 & \downarrow f_3 & & \downarrow f_2 & & \downarrow f_1 & & \downarrow f_0 & & \downarrow f_{-1} \\
 \dashrightarrow & C_3(L) & \xrightarrow{\partial_3} & C_2(L) & \xrightarrow{\partial_2} & C_1(L) & \xrightarrow{\partial_1} & C_0(L) & \xrightarrow{\partial_0} & \{0\}
 \end{array}$$

Lemma: For every $n \geq 0$, we have $\partial_n \circ f_n = f_{n-1} \circ \partial_n$.

Proposition: For every $c \in Z_n(K)$, we have $f_n(c) \in Z_n(L)$.
 For every $c \in B_n(K)$, we also have $f_n(c) \in B_n(L)$.

The fact that $f(Z_n(K)) \subset Z_n(L)$ and $f(B_n(K)) \subset B_n(L)$ allows to define a linear map between quotient vector spaces:

$$(f_n)_*: Z_n(K)/B_n(K) \longrightarrow Z_n(L)/B_n(L).$$

By definition of the homology groups, we have defined a map

$$(f_n)_*: H_n(K) \longrightarrow H_n(L).$$

It is called the **induced map in homology**.

Mapa linear induzido

8/46 (6/11)

$$\begin{array}{ccccccc}
 \dashrightarrow & C_3(K) & \xrightarrow{\partial_3} & C_2(K) & \xrightarrow{\partial_2} & C_1(K) & \xrightarrow{\partial_1} & C_0(K) & \xrightarrow{\partial_0} & \{0\} \\
 & \downarrow f_3 & & \downarrow f_2 & & \downarrow f_1 & & \downarrow f_0 & & \downarrow f_{-1} \\
 \dashrightarrow & C_3(L) & \xrightarrow{\partial_3} & C_2(L) & \xrightarrow{\partial_2} & C_1(L) & \xrightarrow{\partial_1} & C_0(L) & \xrightarrow{\partial_0} & \{0\}
 \end{array}$$

Lemma: For every $n \geq 0$, we have $\partial_n \circ f_n = f_{n-1} \circ \partial_n$.

Proposition: For every $c \in Z_n(K)$, we have $f_n(c) \in Z_n(L)$.
 For every $c \in B_n(K)$, we also have $f_n(c) \in B_n(L)$.

$$\begin{array}{ccccc}
 \cdots & H_3(K) & H_2(K) & H_1(K) & H_0(K) \\
 & \downarrow (f_3)* & \downarrow (f_2)* & \downarrow (f_1)* & \downarrow (f_0)* \\
 \cdots & H_3(L) & H_2(L) & H_1(L) & H_0(L)
 \end{array}$$

$(f_n)_*$ can be defined as

$$c = \sum_{\sigma \in K_{(n)}} \epsilon_\sigma \cdot \sigma \longmapsto \sum_{\sigma \in K_{(n)}} \epsilon_\sigma \cdot f_n(\sigma)$$

Mapa linear induzido

8/46 (7/11)

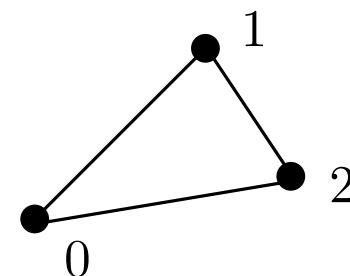
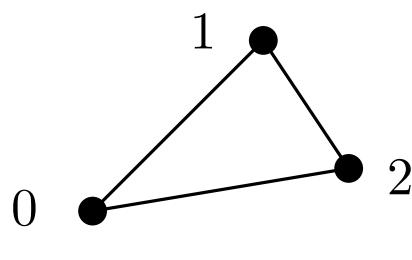
Example: Consider the simplicial complexes $K = L = \{[0], [1], [2], [0, 1], [0, 2], [1, 2]\}$.

The inclusion $i: K \rightarrow L$ induces the identity in H_0 :

$$(i_0)_*: H_0(K) \simeq \mathbb{Z}/2\mathbb{Z} \longrightarrow H_0(L) \simeq \mathbb{Z}/2\mathbb{Z}$$
$$1 \longmapsto 1$$

The inclusion $i: K \rightarrow L$ induces the identity in H_1 :

$$(i_1)_*: H_1(K) \simeq \mathbb{Z}/2\mathbb{Z} \longrightarrow H_1(L) \simeq \mathbb{Z}/2\mathbb{Z}$$
$$1 \longmapsto 1$$



$(f_n)_*$ can be defined as

$$c = \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot \sigma \longmapsto \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot f_n(\sigma)$$

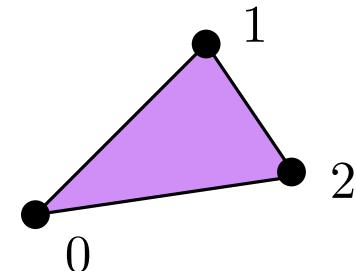
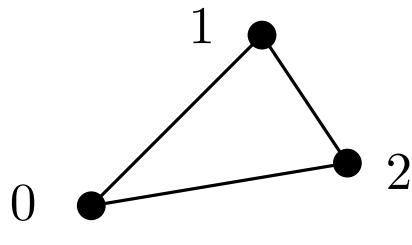
Mapa linear induzido

8/46 (8/11)

Example: Consider the simplicial complexes $K = \{[0], [1], [2], [0, 1], [0, 2], [1, 2]\}$ and $L = \{[0], [1], [2], [0, 1], [0, 2], [1, 2], [0, 1, 2]\}$.

The inclusion $i: K \rightarrow L$ induces the zero map in H^1 :

$$(i_1)_*: H_1(K) \simeq \mathbb{Z}/2\mathbb{Z} \longrightarrow H_1(L) \simeq \{0\}$$
$$1 \longmapsto 0$$



$(f_n)_*$ can be defined as

$$c = \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot \sigma \longmapsto \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot f_n(\sigma)$$

Mapa linear induzido

8/46 (9/11)

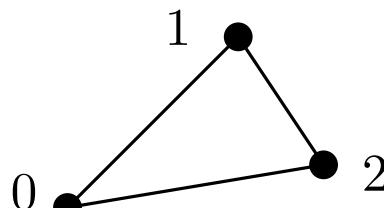
Example: Consider the simplicial complexes $K = \{[0], [1], [2], [0, 1], [0, 2], [1, 2]\}$ and $L = \{[0], [1], [2], [3], [0, 1], [0, 2], [1, 2], [1, 3], [2, 3]\}$.

The homology group $H_1(L)$ is isomorphic to the vector space $(\mathbb{Z}/2\mathbb{Z})^2$ by identifying $[0, 1] + [0, 2] + [1, 2] \mapsto (1, 0)$ and $[1, 2] + [2, 3] + [1, 3] \mapsto (0, 1)$.

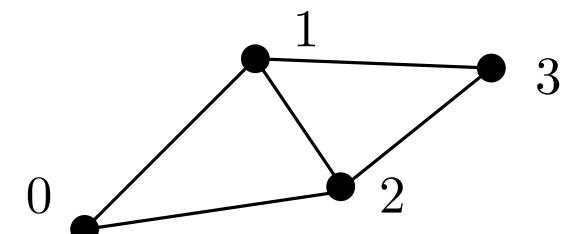
The inclusion $i: K \rightarrow L$ induces the following map between 1st homology groups:

$$(i_1)_*: H_1(K) \simeq \mathbb{Z}/2\mathbb{Z} \longrightarrow H_1(L) \simeq (\mathbb{Z}/2\mathbb{Z})^2$$
$$1 \longmapsto (1, 0)$$

It can be represented as the matrix $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$.



$(f_n)_*$ can be defined as



$$c = \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot \sigma \longmapsto \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot f_n(\sigma)$$

Mapa linear induzido

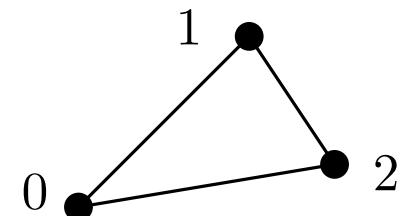
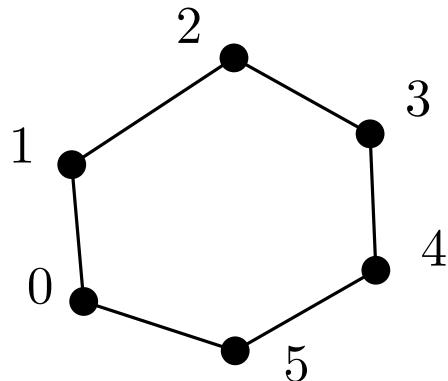
8/46 (10/11)

Example: Let $K = \{[0], [1], [2], [3], [4], [5], [0, 1], [1, 2], [2, 3], [3, 4], [4, 5], [5, 0]\}$ and $L = \{[0], [1], [2], [0, 1], [1, 2], [2, 0]\}$.

Consider the simplicial map $f: i \mapsto i \text{ (modulo 3)}$.

The induced map $(f_1)_*$ is

$$(f_1)_*: H_1(K) \simeq \mathbb{Z}/2\mathbb{Z} \longrightarrow H_1(L) \simeq \mathbb{Z}/2\mathbb{Z}$$
$$? \longmapsto ?$$



$(f_n)_*$ can be defined as

$$c = \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot \sigma \longmapsto \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot f_n(\sigma)$$

Mapa linear induzido

8/46 (11/11)

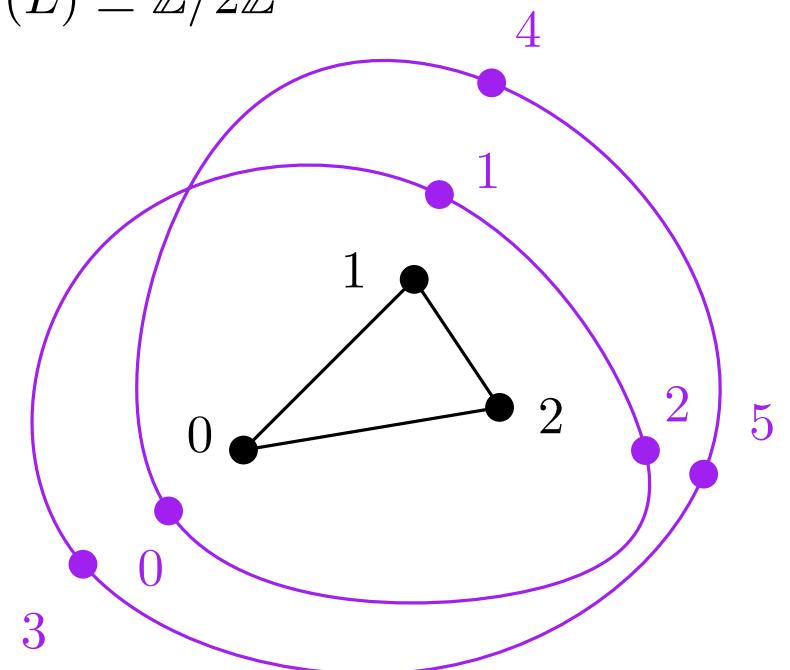
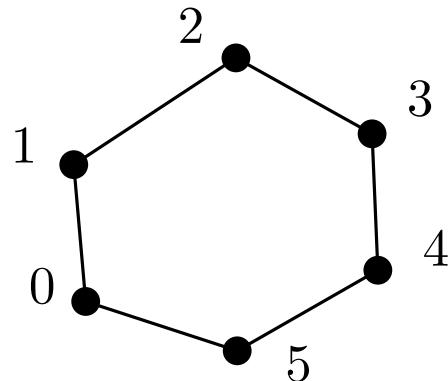
Example: Let $K = \{[0], [1], [2], [3], [4], [5], [0, 1], [1, 2], [2, 3], [3, 4], [4, 5], [5, 0]\}$ and $L = \{[0], [1], [2], [0, 1], [1, 2], [2, 0]\}$.

Consider the simplicial map $f: i \mapsto i$ (modulo 3).

The induced map $(f_1)_*$ is zero

$$(f_1)_*: H_1(K) \simeq \mathbb{Z}/2\mathbb{Z} \longrightarrow H_1(L) \simeq \mathbb{Z}/2\mathbb{Z}$$

$$1 \longmapsto 0$$



$(f_n)_*$ can be defined as

In $\mathbb{Z}/2\mathbb{Z}$, we have $2 = 0$

$$c = \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot \sigma \longmapsto \sum_{\sigma \in K(n)} \epsilon_\sigma \cdot f_n(\sigma)$$

Proposition: Let K, L, M be three simplicial complexes, and consider two simplicial maps $f: K \rightarrow L$ and $g: L \rightarrow M$.

For any $n \geq 0$, the induced map $((g \circ f)_n)_*: H_n(K) \rightarrow H_n(M)$ and $(g_n)_* \circ (f_n)_*: H_n(K) \rightarrow H_n(M)$ are equal.

$$\begin{array}{ccc}
 & \xrightarrow{g \circ f} & \\
 K \xrightarrow{f} L \xrightarrow{g} M, & & H_n(K) \xrightarrow{f_*} H_n(L) \xrightarrow{g_*} H_n(M).
 \end{array}$$

$\xrightarrow{(g \circ f)_*}$

Proposition: Let K, L, M be three simplicial complexes, and consider two simplicial maps $f: K \rightarrow L$ and $g: L \rightarrow M$.

For any $n \geq 0$, the induced map $((g \circ f)_n)_*: H_n(K) \rightarrow H_n(M)$ and $(g_n)_* \circ (f_n)_*: H_n(K) \rightarrow H_n(M)$ are equal.

$$K \xrightarrow{f} L \xrightarrow{g} M, \quad H_n(K) \xrightarrow{f_*} H_n(L) \xrightarrow{g_*} H_n(M).$$

$\xrightarrow{g \circ f}$ $\xrightarrow{(g \circ f)_*}$

Proof: Let $\sigma \in K_{(n)}$. The image $(g \circ f)_n(\sigma)$ is

- $(g \circ f)(\sigma)$ if $g \circ f$ is injective on σ ,
- 0 else.

If $g \circ f$ is injective on σ , then f is injective on σ **and** g is injective on $f(\sigma)$, hence $g_n \circ f_n(\sigma) = g \circ f(\sigma)$, and we deduce the result.

If $g \circ f$ is not injective on σ , then f is not injective on σ **or** g is not injective on $f(\sigma)$, hence $g_n \circ f_n(\sigma) = 0$, and we deduce the result.

I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

Acompanhar os ciclos ao longo do tempo

P1/46 (1/4)

Let $X \subset \mathbb{R}^n$. The collection its thickenings is an increasing **sequence of subsets**

$$\dots \subset X^{t_1} \subset X^{t_2} \subset X^{t_3} \subset \dots$$

By considering the corresponding Čech complexes, we obtain an non-decreasing **sequence of simplicial complexes**

$$\dots \subset \check{\text{Cech}}^{t_1}(X) \subset \check{\text{Cech}}^{t_2}(X) \subset \check{\text{Cech}}^{t_3}(X) \subset \dots$$

Let us denote i_s^t the inclusion map corresponding to $\check{\text{Cech}}^s(X) \subset \check{\text{Cech}}^t(X)$. We write

$$\dots \rightarrow \check{\text{Cech}}^{t_1}(X) \xrightarrow{i_{t_1}^{t_2}} \check{\text{Cech}}^{t_2}(X) \xrightarrow{i_{t_2}^{t_3}} \check{\text{Cech}}^{t_3}(X) \dots$$

Applying the i^{th} homology functor yields a **diagram of vector spaces**

$$\dots \rightarrow H_i(\check{\text{Cech}}^{t_1}(X)) \xrightarrow{(i_{t_1}^{t_2})_*} H_i(\check{\text{Cech}}^{t_2}(X)) \xrightarrow{(i_{t_2}^{t_3})_*} H_i(\check{\text{Cech}}^{t_3}(X)) \dots$$

where the maps $(i_s^t)_*$ are those induced in homology by the inclusions i_s^t .

Acompanhar os ciclos ao longo do tempo

1/46 (2/4)

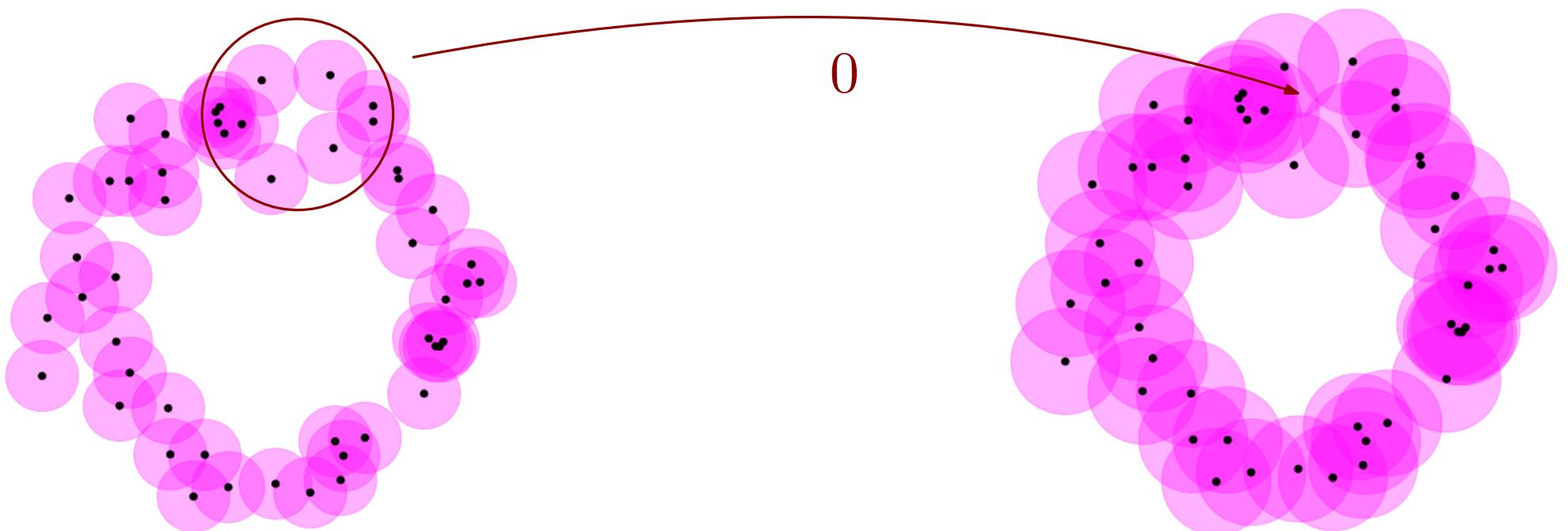
$$\text{-----} \rightarrow H_i(\check{\text{Cech}}^{t_1}(X)) \xrightarrow{(i_{t_1}^{t_2})_*} H_i(\check{\text{Cech}}^{t_2}(X)) \xrightarrow{(i_{t_2}^{t_3})_*} H_i(\check{\text{Cech}}^{t_3}(X)) \text{ -----}$$

Let $i \geq 0$, $t_0 \geq 0$ and consider a cycle $c \in H_i(\check{\text{Cech}}^{t_0}(X))$.

Its **death time** is: $\sup \{t \geq t_0 \mid (i_t^t)(c) \neq 0\}$,

its **birth time** is: $\inf \{t \geq t_0 \mid (i_t^{t_0})^{-1}(\{c\}) \neq \emptyset\}$,

its **persistence** is the difference.



Acompanhar os ciclos ao longo do tempo

P1/46 (3/4)

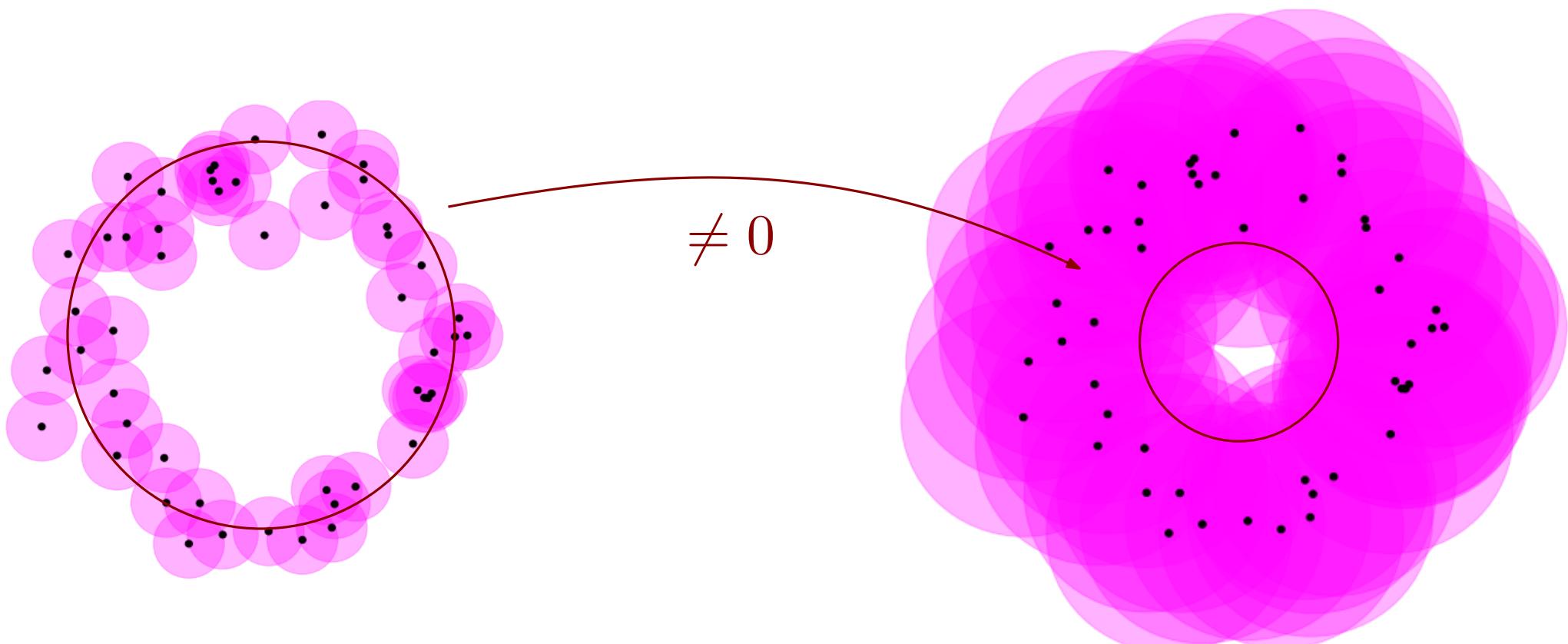
$$\cdots \rightarrow H_i(\check{\text{Cech}}^{t_1}(X)) \xrightarrow{(i_{t_1}^{t_2})_*} H_i(\check{\text{Cech}}^{t_2}(X)) \xrightarrow{(i_{t_2}^{t_3})_*} H_i(\check{\text{Cech}}^{t_3}(X)) \cdots$$

Let $i \geq 0$, $t_0 \geq 0$ and consider a cycle $c \in H_i(\check{\text{Cech}}^{t_0}(X))$.

Its **death time** is: $\sup \{t \geq t_0 \mid (i_{t_0}^t)(c) \neq 0\}$,

its **birth time** is: $\inf \{t \geq t_0 \mid (i_t^{t_0})^{-1}(\{c\}) \neq \emptyset\}$,

its **persistence** is the difference.



Acompanhar os ciclos ao longo do tempo

1/46 (4/4)

$$\cdots \rightarrow H_i(\check{\text{Cech}}^{t_1}(X)) \xrightarrow{(i_{t_1}^{t_2})_*} H_i(\check{\text{Cech}}^{t_2}(X)) \xrightarrow{(i_{t_2}^{t_3})_*} H_i(\check{\text{Cech}}^{t_3}(X)) \cdots$$

Let $i \geq 0$, $t_0 \geq 0$ and consider a cycle $c \in H_i(\check{\text{Cech}}^{t_0}(X))$.

Its **death time** is: $\sup \{t \geq t_0 \mid (i_{t_0}^t)(c) \neq 0\}$,

its **birth time** is: $\inf \{t \geq t_0 \mid (i_t^{t_0})^{-1}(\{c\}) \neq \emptyset\}$,

its **persistence** is the difference.

As a rule of thumb:

- cycles with **large persistence** correspond to important topological features of the dataset,
- cycles with **short persistence** corresponds to topological noise.

Definition: A **persistence module** \mathbb{V} over \mathbb{R}^+ with coefficients in $\mathbb{Z}/2\mathbb{Z}$ is a pair (\mathbb{V}, v) where $\mathbb{V} = (V^t)_{t \in \mathbb{R}^+}$ is a family of $\mathbb{Z}/2\mathbb{Z}$ -vector spaces, and $v = (v_s^t : V^s \rightarrow V^t)_{s \leq t \in \mathbb{R}^+}$ a family of linear maps such that:

- for every $t \in \mathbb{R}^+$, $v_t^t : V^t \rightarrow V^t$ is the identity map,
- for every $r, s, t \in \mathbb{R}^+$ such that $r \leq s \leq t$, we have $v_s^t \circ v_r^s = v_r^t$.

In practice, one builds persistence modules from **filtrations**, that is, non-decreasing families of simplicial complexes $\mathbb{S} = (S^t)_{t \in \mathbb{R}^+}$. For instance, the Čech complex, or the Rips complex.

By applying the i^{th} homology functor to a filtration, we obtain a persistence module $\mathbb{V}[\mathbb{S}] = (H_i(S^t))_{t \in \mathbb{R}^+}$, with maps $((i_s^t)_* : H_i(S^s) \rightarrow H_i(S^t))_{s \leq t}$ induced by the inclusions.

$$\begin{array}{ccccccc}
 \cdots & \longrightarrow & S^{t_1} & \xleftarrow{i_{t_1}^{t_2}} & S^{t_2} & \xleftarrow{i_{t_2}^{t_3}} & S^{t_3} \xleftarrow{i_{t_3}^{t_4}} S^{t_4} \cdots \\
 \cdots & \longrightarrow & H_i(S^{t_1}) & \xrightarrow{(i_{t_1}^{t_2})_*} & H_i(S^{t_2}) & \xrightarrow{(i_{t_2}^{t_3})_*} & H_i(S^{t_3}) \xrightarrow{(i_{t_3}^{t_4})_*} H_i(S^{t_4}) \cdots
 \end{array}$$

Definition: An **isomorphism** between two persistence modules (\mathbb{V}, v) and (\mathbb{W}, w) is a family of isomorphisms of vector spaces $\phi = (\phi_t: V^t \rightarrow W^t)_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^s & \xrightarrow{w_s^t} & W^t \end{array}$$

I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

Definition: Let (\mathbb{V}, v) and (\mathbb{W}, w) be two persistence modules.

Their **sum** is the persistence module $\mathbb{V} \oplus \mathbb{W}$ defined with the vector spaces $(V \oplus W)^t = V^t \oplus W^t$ and the linear maps

$$(v \oplus w)_s^t : (x, y) \in (V \oplus W)^s \longmapsto (v_s^t(x), w_s^t(y)) \in (V \oplus W)^t.$$

A persistence module \mathbb{U} is **indecomposable** if for every pair of persistence modules \mathbb{V} and \mathbb{W} such that \mathbb{U} is isomorphic to the sum $\mathbb{V} \oplus \mathbb{W}$, then one of the summands has to be a trivial persistence module, that is, equal to zero for every $t \in \mathbb{R}^+$. In other words,

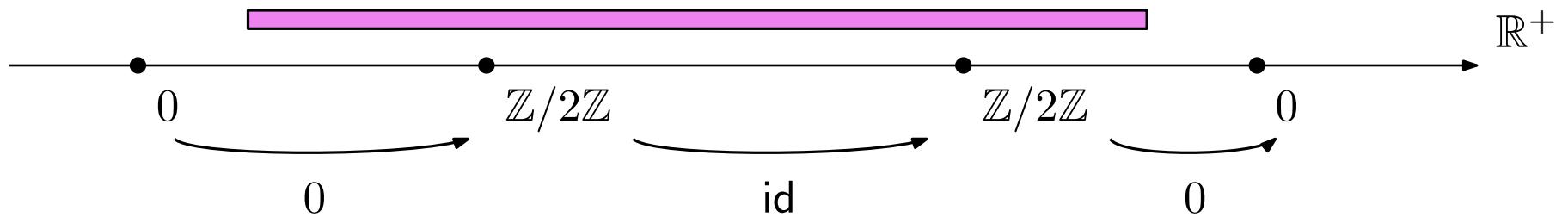
$$\mathbb{U} \simeq \mathbb{V} \oplus \mathbb{W} \implies \mathbb{V} = 0 \text{ or } \mathbb{W} = 0.$$

Otherwise, \mathbb{U} is said **decomposable**.

Definition: Let $I \subset \mathbb{R}^+$ be an interval: $[a, b]$, $(a, b]$, $[a, b)$ or (a, b) , with $a, b \in \mathbb{R}^+$ such that $a \leq b$, and potentially $a = -\infty$ or $b = +\infty$.

The **interval module** associated to I is the persistence module $\mathbb{B}[I]$ with vector spaces $\mathbb{B}^t[I]$ and linear maps $v_s^t: \mathbb{B}^s[I] \rightarrow \mathbb{B}^t[I]$ defined as

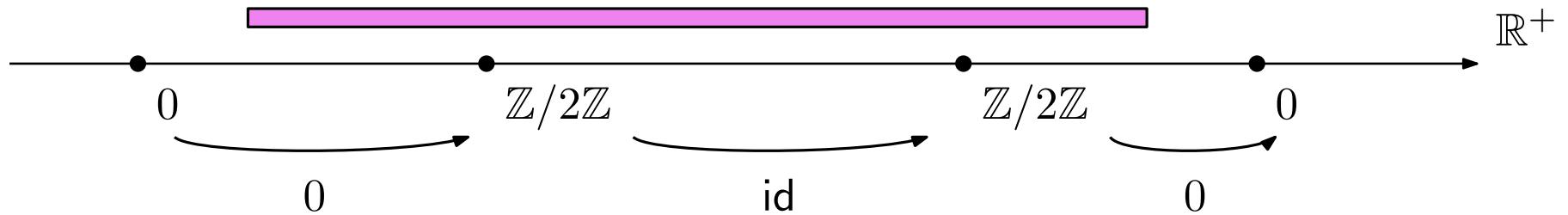
$$\mathbb{B}^t[I] = \begin{cases} \mathbb{Z}/2\mathbb{Z} & \text{if } t \in I, \\ 0 & \text{otherwise,} \end{cases} \quad \text{and} \quad v_s^t = \begin{cases} \text{id} & \text{if } s, t \in I, \\ 0 & \text{otherwise.} \end{cases}$$



Definition: Let $I \subset \mathbb{R}^+$ be an interval: $[a, b]$, $(a, b]$, $[a, b)$ or (a, b) , with $a, b \in \mathbb{R}^+$ such that $a \leq b$, and potentially $a = -\infty$ or $b = +\infty$.

The **interval module** associated to I is the persistence module $\mathbb{B}[I]$ with vector spaces $\mathbb{B}^t[I]$ and linear maps $v_s^t: \mathbb{B}^s[I] \rightarrow \mathbb{B}^t[I]$ defined as

$$\mathbb{B}^t[I] = \begin{cases} \mathbb{Z}/2\mathbb{Z} & \text{if } t \in I, \\ 0 & \text{otherwise,} \end{cases} \quad \text{and} \quad v_s^t = \begin{cases} \text{id} & \text{if } s, t \in I, \\ 0 & \text{otherwise.} \end{cases}$$



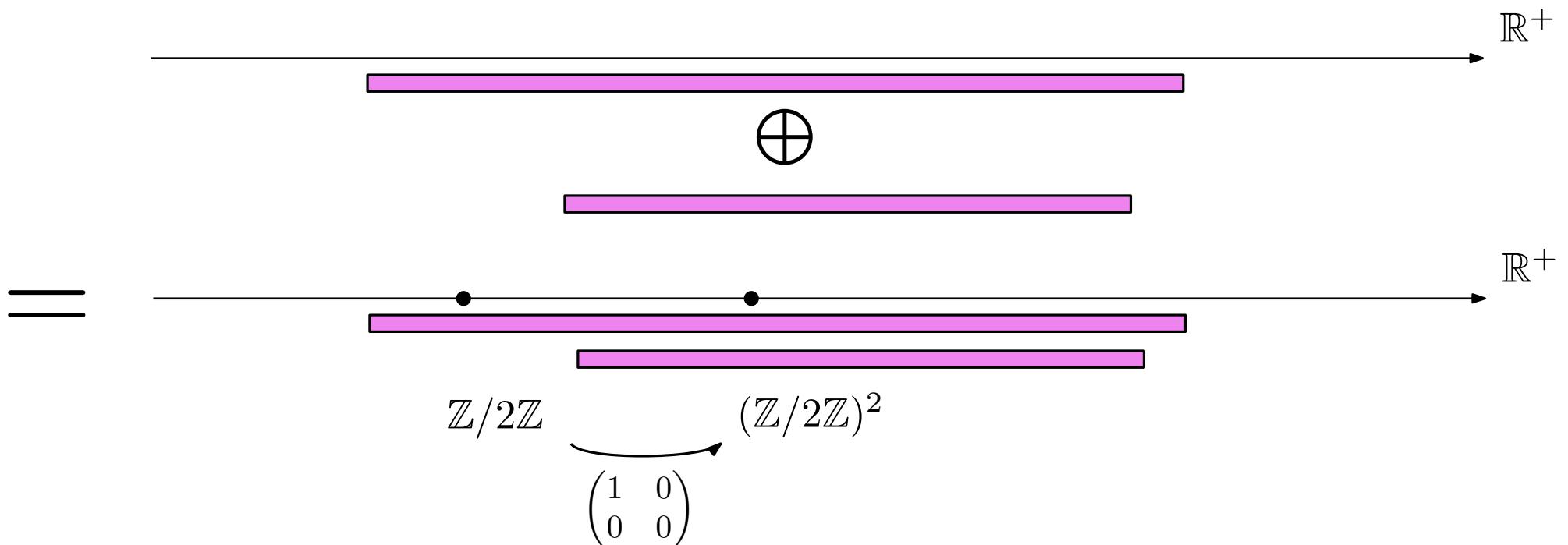
Lemma: Interval modules are indecomposable.

Definition: Let $I \subset \mathbb{R}^+$ be an interval: $[a, b]$, $(a, b]$, $[a, b)$ or (a, b) , with $a, b \in \mathbb{R}^+$ such that $a \leq b$, and potentially $a = -\infty$ or $b = +\infty$.

The **interval module** associated to I is the persistence module $\mathbb{B}[I]$ with vector spaces $\mathbb{B}^t[I]$ and linear maps $v_s^t: \mathbb{B}^s[I] \rightarrow \mathbb{B}^t[I]$ defined as

$$\mathbb{B}^t[I] = \begin{cases} \mathbb{Z}/2\mathbb{Z} & \text{if } t \in I, \\ 0 & \text{otherwise,} \end{cases} \quad \text{and} \quad v_s^t = \begin{cases} \text{id} & \text{if } s, t \in I, \\ 0 & \text{otherwise.} \end{cases}$$

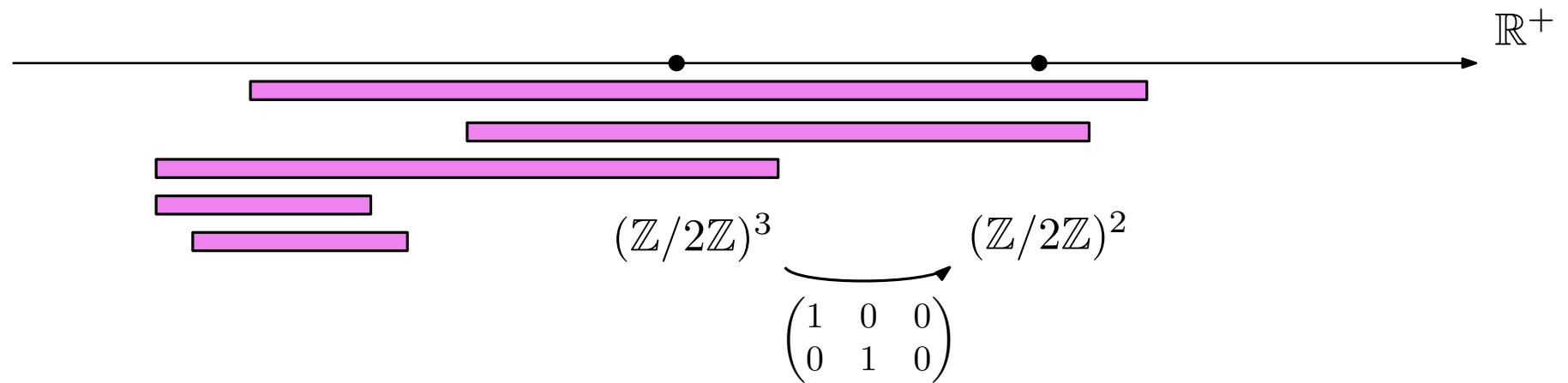
We can sum interval modules:



Definition: Let $I \subset \mathbb{R}^+$ be an interval: $[a, b]$, $(a, b]$, $[a, b)$ or (a, b) , with $a, b \in \mathbb{R}^+$ such that $a \leq b$, and potentially $a = -\infty$ or $b = +\infty$.

The **interval module** associated to I is the persistence module $\mathbb{B}[I]$ with vector spaces $\mathbb{B}^t[I]$ and linear maps $v_s^t: \mathbb{B}^s[I] \rightarrow \mathbb{B}^t[I]$ defined as

$$\mathbb{B}^t[I] = \begin{cases} \mathbb{Z}/2\mathbb{Z} & \text{if } t \in I, \\ 0 & \text{otherwise,} \end{cases} \quad \text{and} \quad v_s^t = \begin{cases} \text{id} & \text{if } s, t \in I, \\ 0 & \text{otherwise.} \end{cases}$$



A persistence module \mathbb{V} **decomposes into interval module** if there exists a multiset \mathcal{I} of intervals of T such that

$$\mathbb{V} \simeq \bigoplus_{I \in \mathcal{I}} \mathbb{B}[I].$$

Multiset means that \mathcal{I} may contain several copies of the same interval I .

Theorem (consequence of Krull–Remak–Schmidt–Azumaya): If a persistence module decomposes into interval modules, then the multiset \mathcal{I} of intervals is unique.

In this case, \mathcal{I} is called the **persistence barcode** of \mathbb{V} . It is written $\text{Barcode}(\mathbb{V})$.



A persistence module \mathbb{V} **decomposes into interval module** if there exists a multiset \mathcal{I} of intervals of T such that

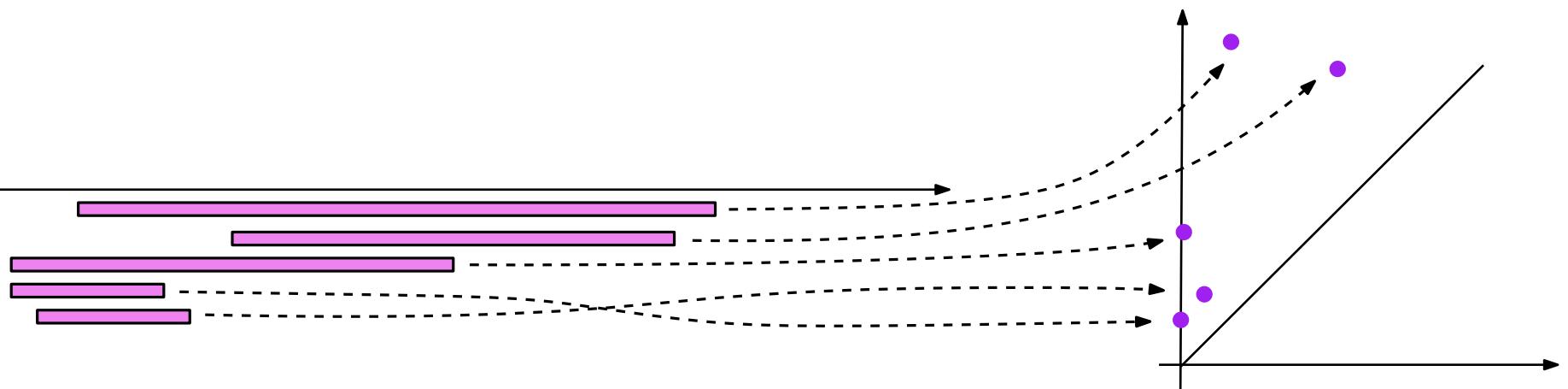
$$\mathbb{V} \simeq \bigoplus_{I \in \mathcal{I}} \mathbb{B}[I].$$

Multiset means that \mathcal{I} may contain several copies of the same interval I .

Theorem (consequence of Krull–Remak–Schmidt–Azumaya): If a persistence module decomposes into interval modules, then the multiset \mathcal{I} of intervals is unique.

In this case, \mathcal{I} is called the **persistence barcode** of \mathbb{V} . It is written $\text{Barcode}(\mathbb{V})$.

For every $[a, b]$, $(a, b]$, $[a, b)$ or (a, b) in $\text{Barcode}(\mathbb{V})$, consider the point (a, b) of \mathbb{R}^2 . The collection of all such points is the **persistence diagram** of \mathbb{V} .



A persistence module \mathbb{V} **decomposes into interval module** if there exists a multiset \mathcal{I} of intervals of T such that

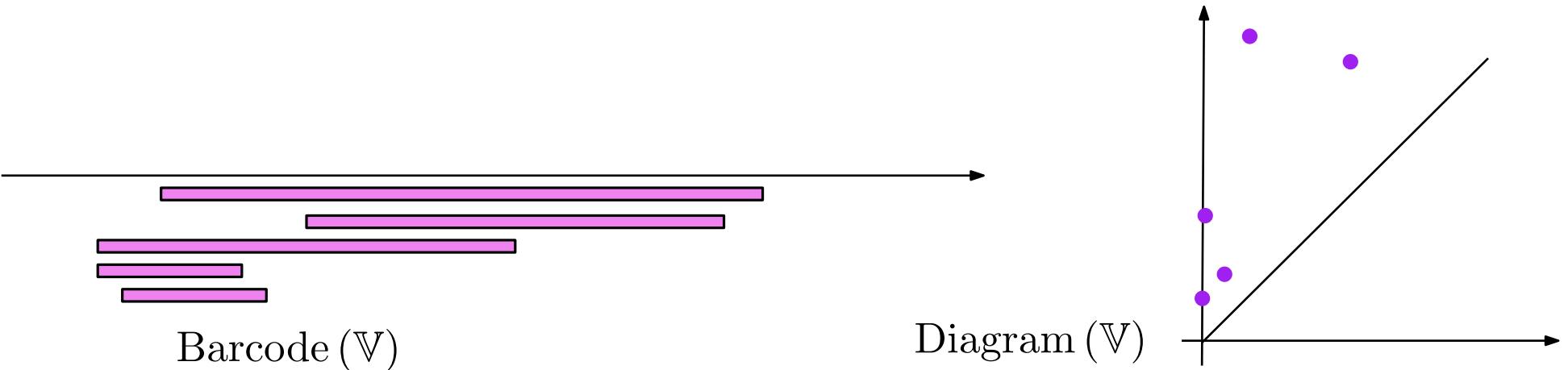
$$\mathbb{V} \simeq \bigoplus_{I \in \mathcal{I}} \mathbb{B}[I].$$

Multiset means that \mathcal{I} may contain several copies of the same interval I .

Theorem (consequence of Krull–Remak–Schmidt–Azumaya): If a persistence module decomposes into interval modules, then the multiset \mathcal{I} of intervals is unique.

In this case, \mathcal{I} is called the **persistence barcode** of \mathbb{V} . It is written $\text{Barcode}(\mathbb{V})$.

For every $[a, b]$, $(a, b]$, $[a, b)$ or (a, b) in $\text{Barcode}(\mathbb{V})$, consider the point (a, b) of \mathbb{R}^2 . The collection of all such points is the **persistence diagram** of \mathbb{V} .



A persistence module \mathbb{V} is said **pointwise finite dimensional** if $\dim V^t < +\infty$ for all t .

Theorem (Crawley-Boevey, 2015): Every pointwise finite-dimensional persistence module decomposes into interval modules.

A persistence module \mathbb{V} is said **pointwise finite dimensional** if $\dim V^t < +\infty$ for all t .

Theorem (Crawley-Boevey, 2015): Every pointwise finite-dimensional persistence module decomposes into interval modules.

Proof (Zomorodian, Carlsson, 2005): Simpler case: the persistence module is finite-dimensional *and* has finitely many terms.

We can write our persistence module as

$$V^1 \xrightarrow{v_1^2} V^2 \xrightarrow{v_2^3} V^3 \xrightarrow{v_3^4} V^4 \longrightarrow \dots \longrightarrow V^n$$

Consider the vector space $\mathcal{V} = \bigotimes_{1 \leq i \leq n} V^i = V^1 \times \dots \times V^n$.

Let $\mathbb{Z}/2\mathbb{Z}[x]$ denote the space of polynomials with coefficients in $\mathbb{Z}/2\mathbb{Z}$. We give \mathcal{V} an action of $\mathbb{Z}/2\mathbb{Z}[x]$ via

$$x \cdot (a^1, a^2, \dots, a^n) = (0, v_1^2(a^1), v_2^3(a^2), \dots, v_{n-1}^n(a^{n-1})).$$

Hence \mathcal{V} can be seen as a finitely generated module over the principal ideal domain $\mathbb{Z}/2\mathbb{Z}[x]$. By classification, \mathcal{V} is isomorphic to a sum

$$\mathcal{V} \simeq \bigoplus_{i \in I} \mathbb{Z}/2\mathbb{Z}[x]/x^i \cdot \mathbb{Z}/2\mathbb{Z}[x].$$

We identify the components $\mathbb{Z}/2\mathbb{Z}[x]/x^i \cdot \mathbb{Z}/2\mathbb{Z}[x]$ with bars of the barcode of length i .

Barcodes

17/46 (6/7)

On a barcode we can read homology **at each step**, and see how it **evolves**.

The Čech or the Rips filtration define an increasing sequence of simplices

$$\dots \subset \check{\text{Cech}}^{t_1}(X) \subset \check{\text{Cech}}^{t_2}(X) \subset \check{\text{Cech}}^{t_3}(X) \subset \dots$$

We can turn it consistently into an ordering of the simplices, by inserting the simplices by order of apparition in the filtration.

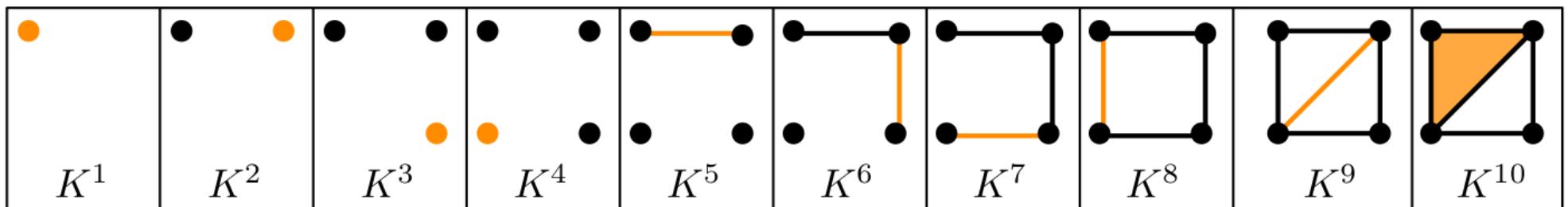
$$\sigma^1 < \sigma^2 < \dots < \sigma^n$$

Denote $t(\sigma)$ the time of apparition of the simplex σ in the filtration. The total order on the simplices satisfies

$$t(\sigma^i) < t(\sigma^j) \text{ for all } i < j.$$

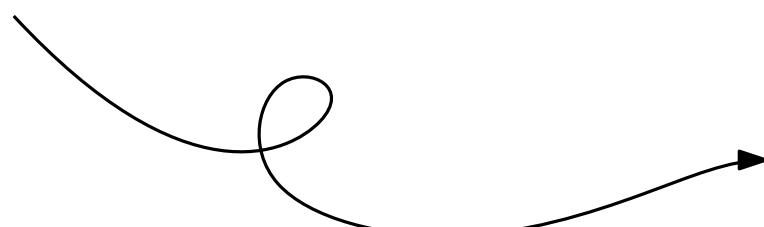
In practice several simplices may appear at the same time. If this occurs, choose an order of the simplices.

→ Consider the boundary matrix, and compute a Gauss reduction.



Algoritmo

18/46 (2/12)



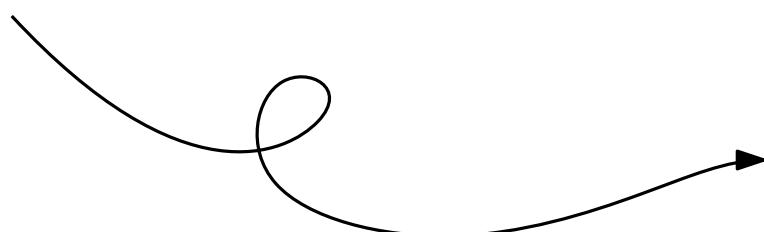
Algoritmo

18/46 (3/12)

For any $j \in [1, n]$,

$$\delta(j) = \max\{i \in \llbracket 1, n \rrbracket \mid \Delta_{i,j} \neq 0\},$$

and $\Delta_{i,j} = 0$ for all j , then $\delta(j)$ is undefined.



Algoritmo

18/46 (4/12)

Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$\left(\sigma^{\delta(j)}, \sigma^j\right).$$

Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**.

Algoritmo

18/46 (5/12)

Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$\left(\sigma^{\delta(j)}, \sigma^j\right).$$

Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**.

Algoritmo

18/46 (6/12)

Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$\left(\sigma^{\delta(j)}, \sigma^j\right).$$

Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**

Algoritmo

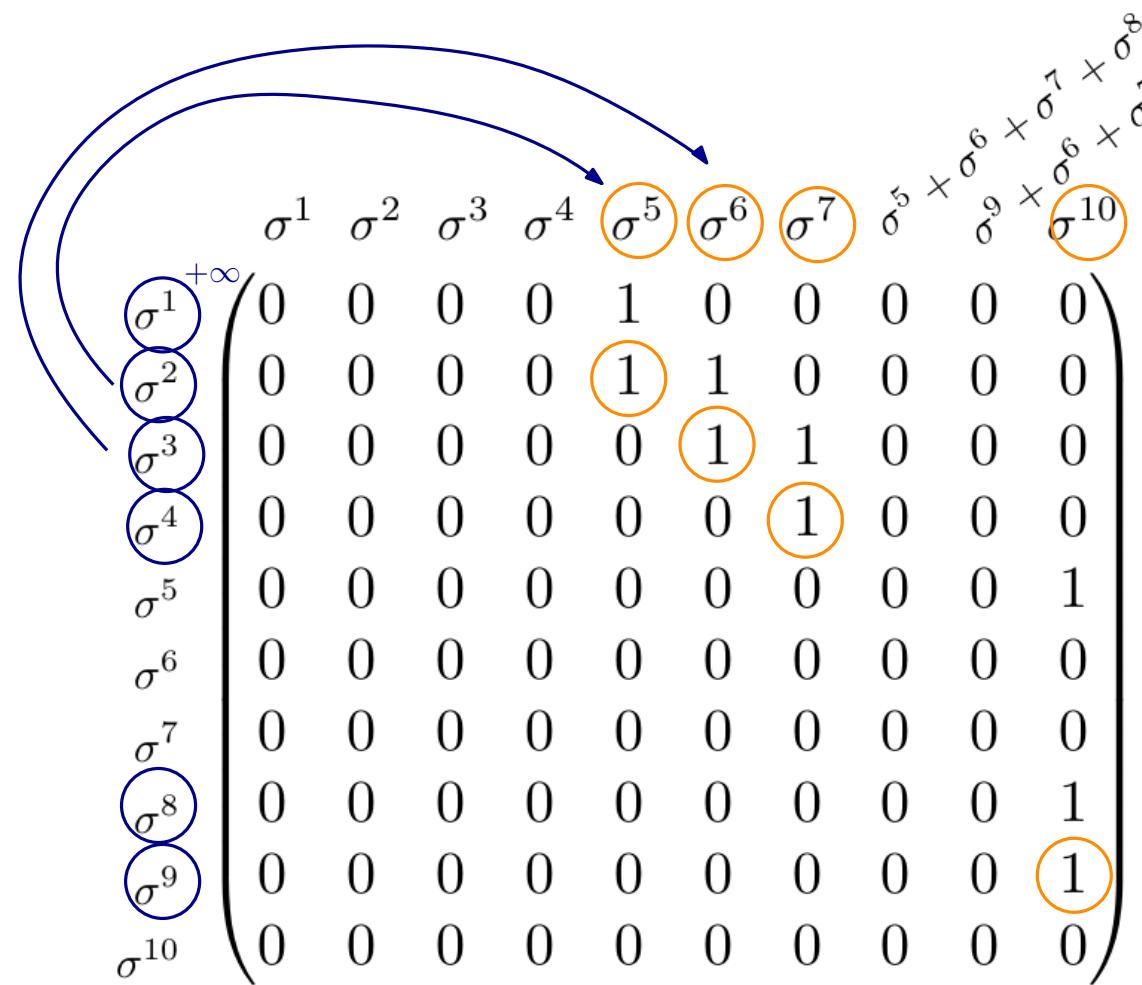
18/46 (7/12)

Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$\left(\sigma^{\delta(j)}, \sigma^j\right).$$

Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**



Algoritmo

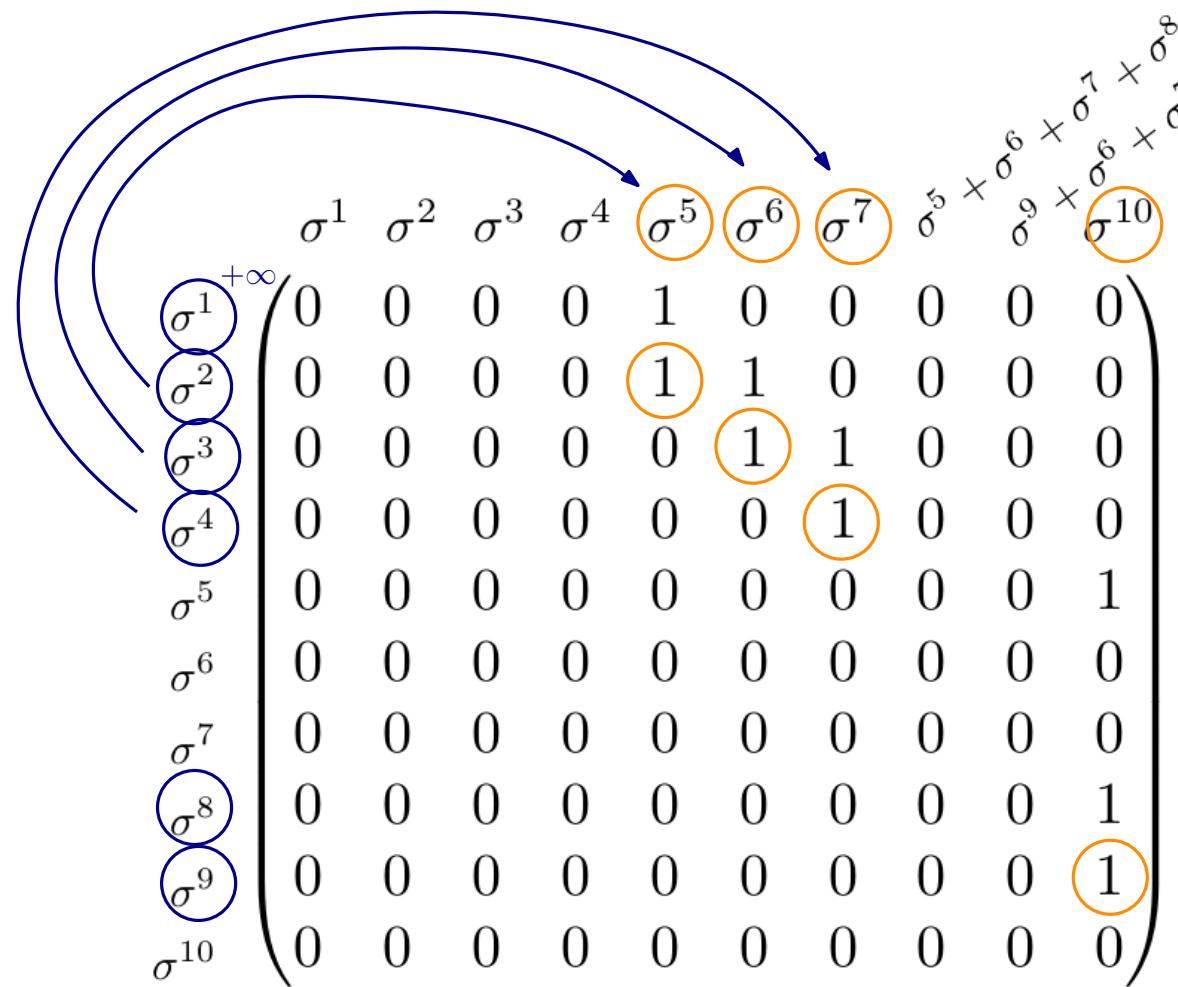
18/46 (8/12)

Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$\left(\sigma^{\delta(j)}, \sigma^j\right).$$

Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**



Algoritmo

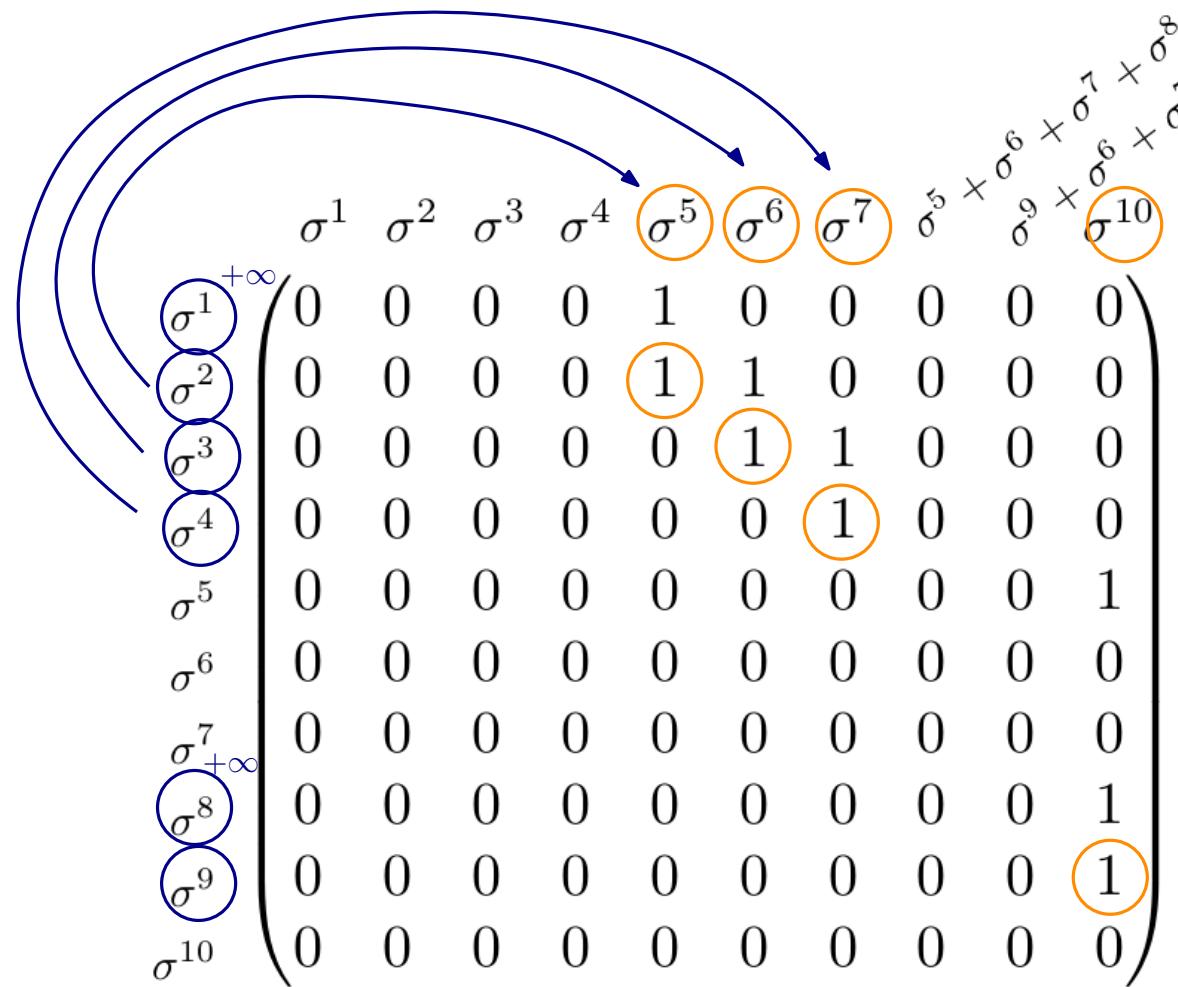
18/46 (9/12)

Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$\left(\sigma^{\delta(j)}, \sigma^j\right).$$

Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**.



Algoritmo

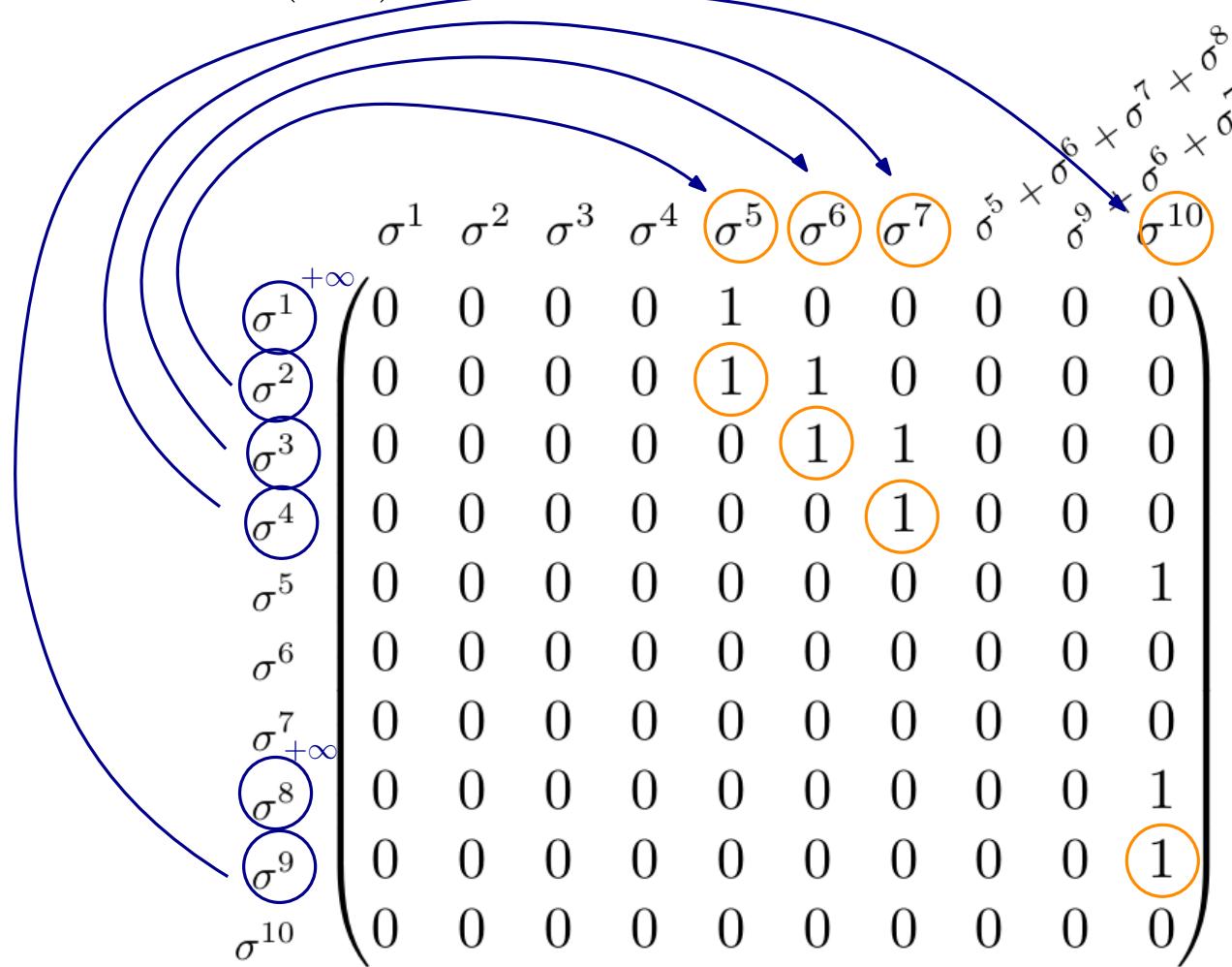
18/46 (10/12)

Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$\left(\sigma^{\delta(j)}, \sigma^j\right).$$

Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**.

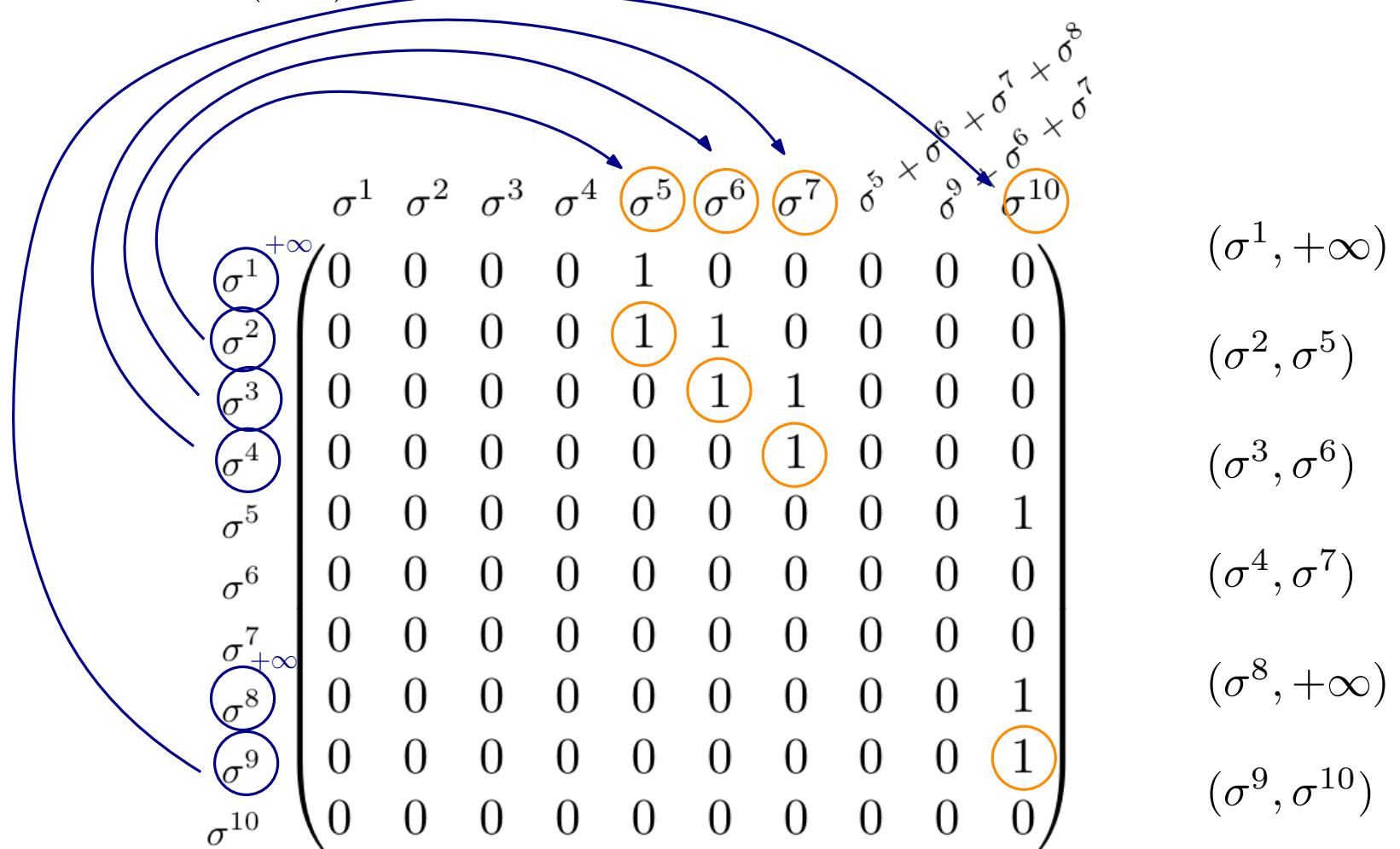


Now, for all j such that $\delta(j)$ is defined, consider the pair of simplices

$$(\sigma^{\delta(j)}, \sigma^j).$$

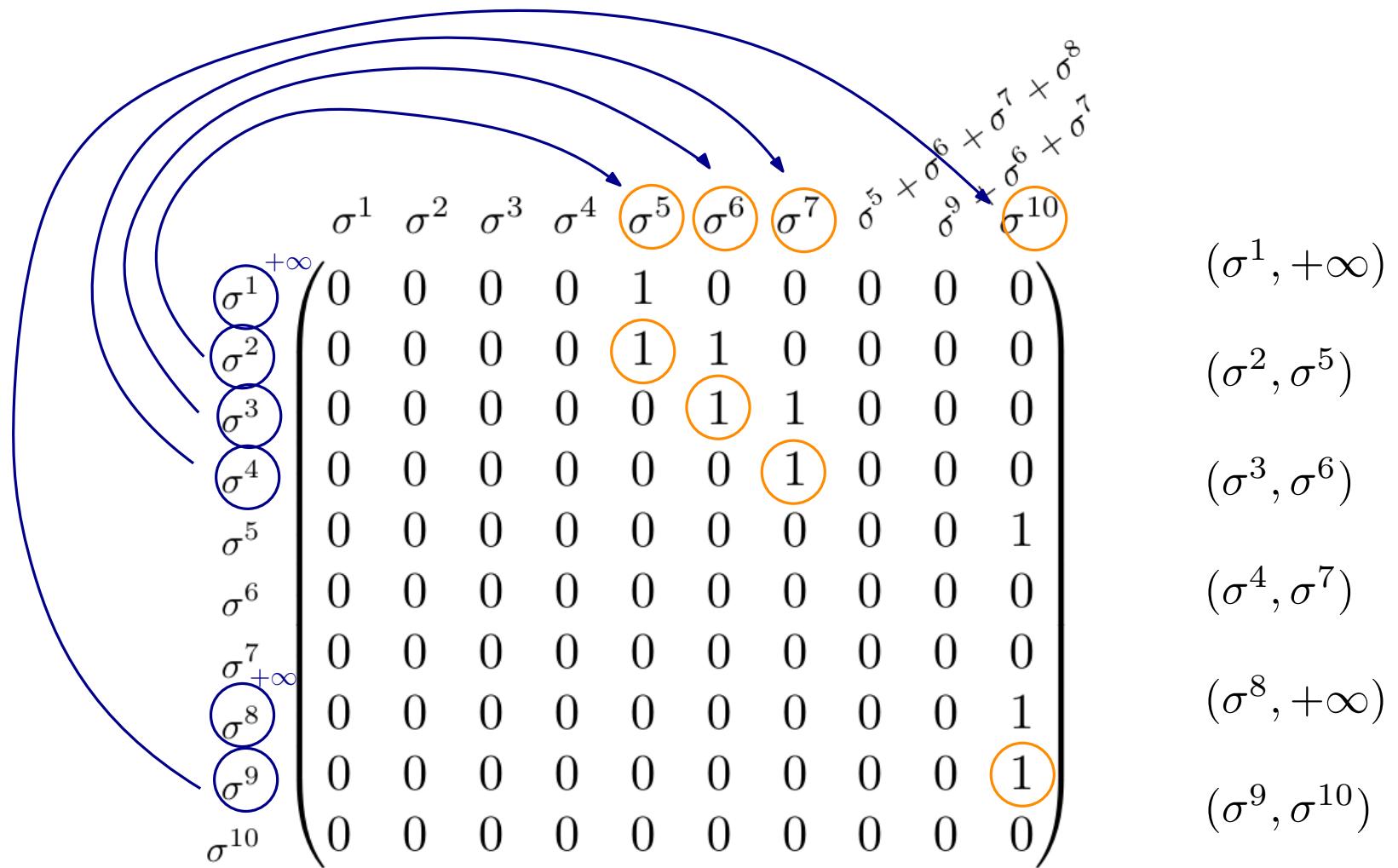
Also, for the values $i \notin \text{Im}(\delta)$, we set: $(\sigma^i, +\infty)$.

The pairs of simplices (σ, τ) are called **persistence pairs**.



Proposition: The barcodes of the filtration consists in the intervals

$$\mathcal{I} = \{ (t(\sigma), t(\tau)) \text{ for all persistence pair } (\sigma, \tau) \text{ such that } t(\sigma) \neq t(\tau) \}.$$



I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

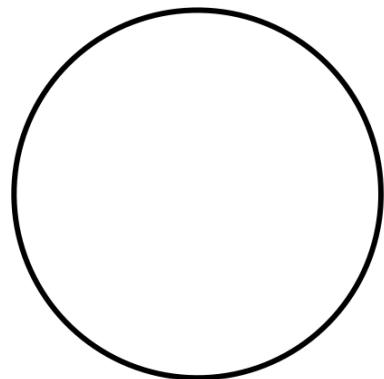
2 - Machine learning

3 - Variations on persistent homology

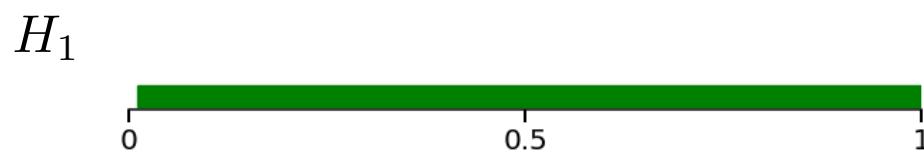
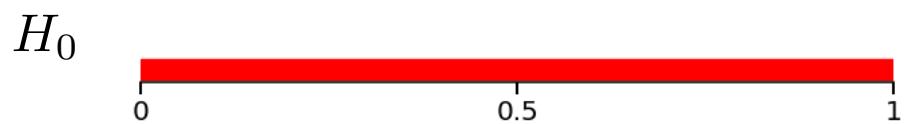
O problema da estabilidade

20/46

Let $X \subset \mathbb{R}^n$ finite, seen as a sample of \mathcal{M} .



Barcodes of the Čech
filtration



I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

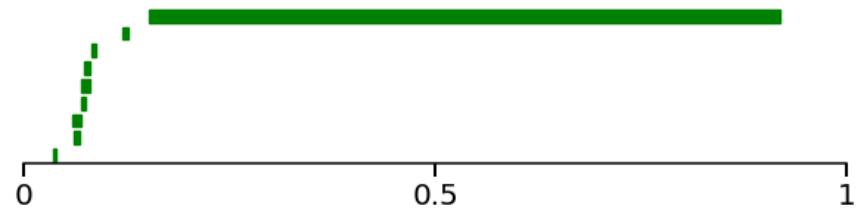
2 - Machine learning

3 - Variations on persistent homology

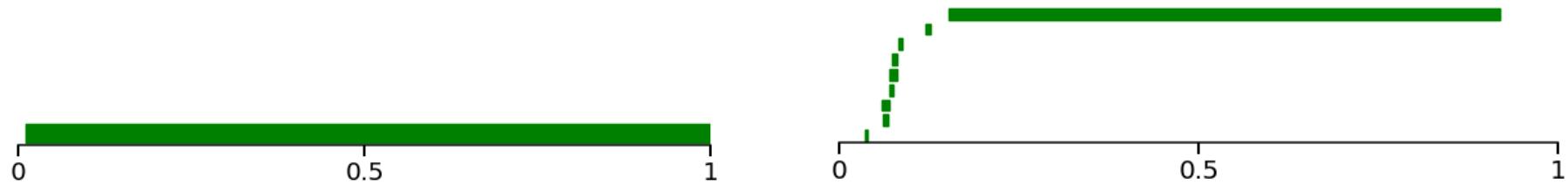
Distância bottleneck

22/46 (1/12)

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.



Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.

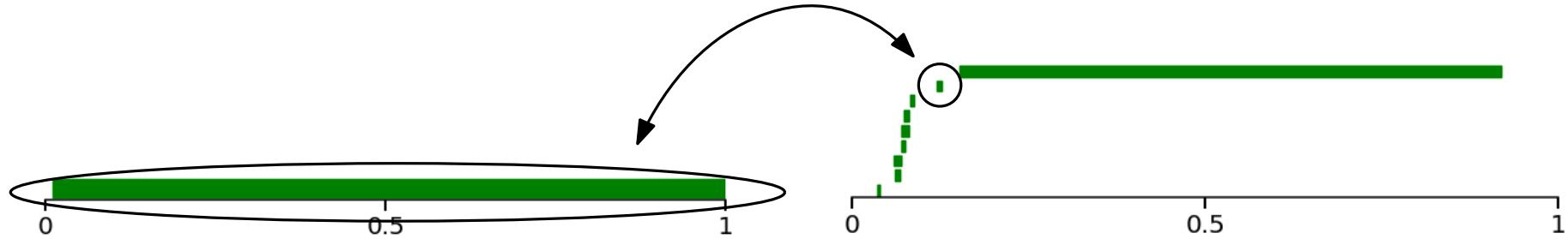


A **partial matching** between the barcodes is a subset $M \subset P \times Q$ such that

- for every $p \in P$, there exists at most one $q \in Q$ such that $(p, q) \in M$,
- for every $q \in Q$, there exists at most one $p \in P$ such that $(p, q) \in M$.

The bars $p \in P$ (resp. $q \in Q$) such that there exists $q \in Q$ (resp. $p \in P$) with $(p, q) \in M$ are said **matched** by M .

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.

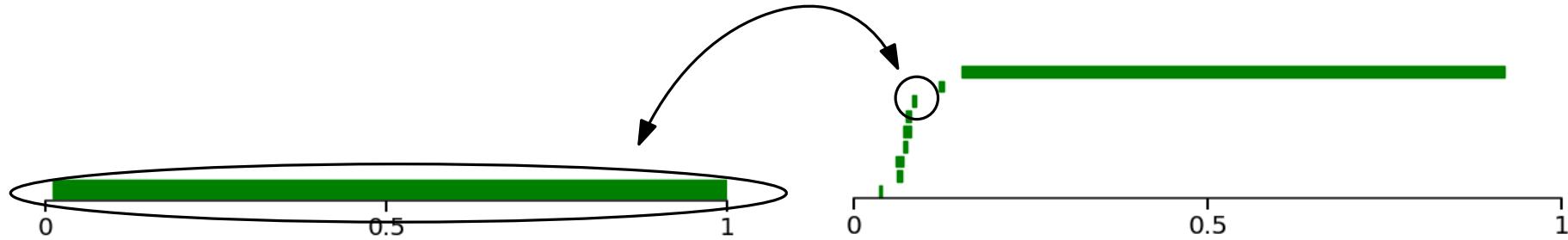


A **partial matching** between the barcodes is a subset $M \subset P \times Q$ such that

- for every $p \in P$, there exists at most one $q \in Q$ such that $(p, q) \in M$,
- for every $q \in Q$, there exists at most one $p \in P$ such that $(p, q) \in M$.

The bars $p \in P$ (resp. $q \in Q$) such that there exists $q \in Q$ (resp. $p \in P$) with $(p, q) \in M$ are said **matched** by M .

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.

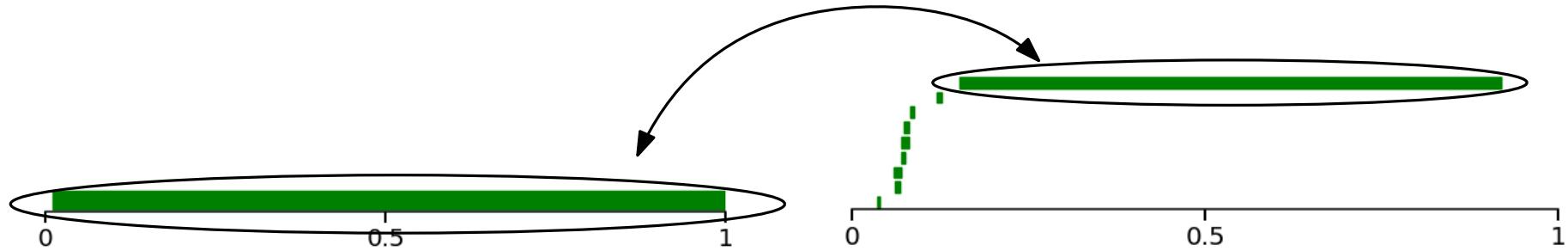


A **partial matching** between the barcodes is a subset $M \subset P \times Q$ such that

- for every $p \in P$, there exists at most one $q \in Q$ such that $(p, q) \in M$,
- for every $q \in Q$, there exists at most one $p \in P$ such that $(p, q) \in M$.

The bars $p \in P$ (resp. $q \in Q$) such that there exists $q \in Q$ (resp. $p \in P$) with $(p, q) \in M$ are said **matched** by M .

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.

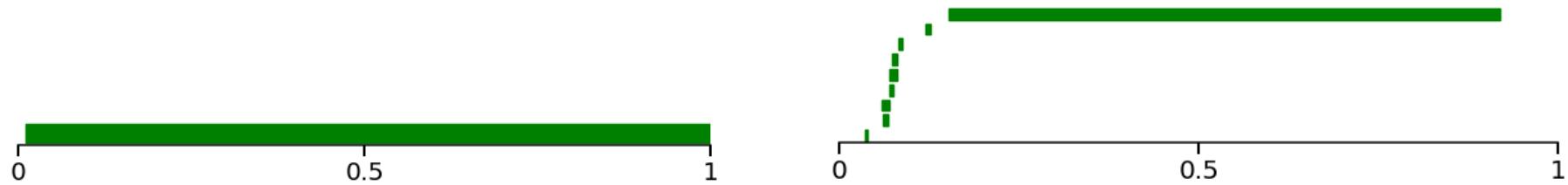


A **partial matching** between the barcodes is a subset $M \subset P \times Q$ such that

- for every $p \in P$, there exists at most one $q \in Q$ such that $(p, q) \in M$,
- for every $q \in Q$, there exists at most one $p \in P$ such that $(p, q) \in M$.

The bars $p \in P$ (resp. $q \in Q$) such that there exists $q \in Q$ (resp. $p \in P$) with $(p, q) \in M$ are said **matched** by M .

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.



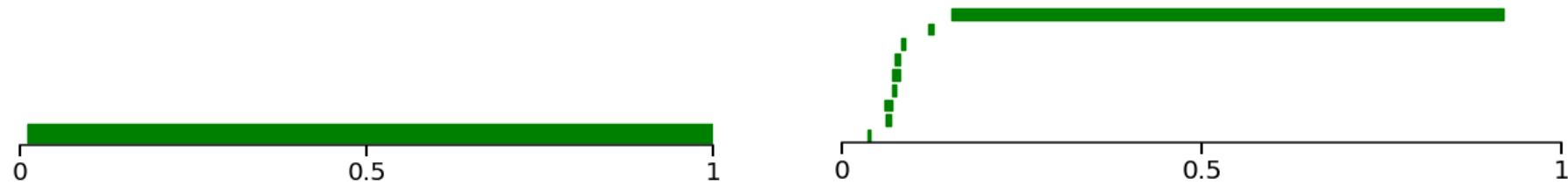
A **partial matching** between the barcodes is a subset $M \subset P \times Q$ such that

- for every $p \in P$, there exists at most one $q \in Q$ such that $(p, q) \in M$,
- for every $q \in Q$, there exists at most one $p \in P$ such that $(p, q) \in M$.

The bars $p \in P$ (resp. $q \in Q$) such that there exists $q \in Q$ (resp. $p \in P$) with $(p, q) \in M$ are said **matched** by M .

If a bar $p \in P$ (resp. $q \in Q$) is not matched by M , we consider that it is matched with the singleton $\bar{p} = [\frac{p_1+p_2}{2}, \frac{p_1+p_2}{2}]$ (resp. $\bar{q} = [\frac{q_1+q_2}{2}, \frac{q_1+q_2}{2}]$).

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.



A **partial matching** between the barcodes is a subset $M \subset P \times Q$ such that

- for every $p \in P$, there exists at most one $q \in Q$ such that $(p, q) \in M$,
- for every $q \in Q$, there exists at most one $p \in P$ such that $(p, q) \in M$.

The bars $p \in P$ (resp. $q \in Q$) such that there exists $q \in Q$ (resp. $p \in P$) with $(p, q) \in M$ are said **matched** by M .

If a bar $p \in P$ (resp. $q \in Q$) is not matched by M , we consider that it is matched with the singleton $\bar{p} = [\frac{p_1+p_2}{2}, \frac{p_1+p_2}{2}]$ (resp. $\bar{q} = [\frac{q_1+q_2}{2}, \frac{q_1+q_2}{2}]$).

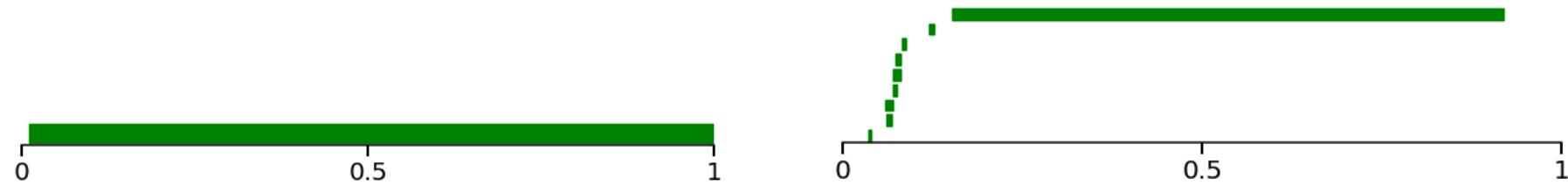
The **cost** of a matched pair (p, q) (resp. (p, \bar{p}) , resp. (\bar{q}, q)) is the sup norm $\|p - q\|_\infty = \sup\{|p_1 - q_1|, |p_2 - q_2|\}$ (resp. $\|p - \bar{p}\|_\infty$, resp. $\|\bar{q} - q\|_\infty$).

The **cost** of the partial matching M , denoted $\text{cost}(M)$, is the supremum of all costs.

Distância bottleneck

22/46 (8/12)

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.



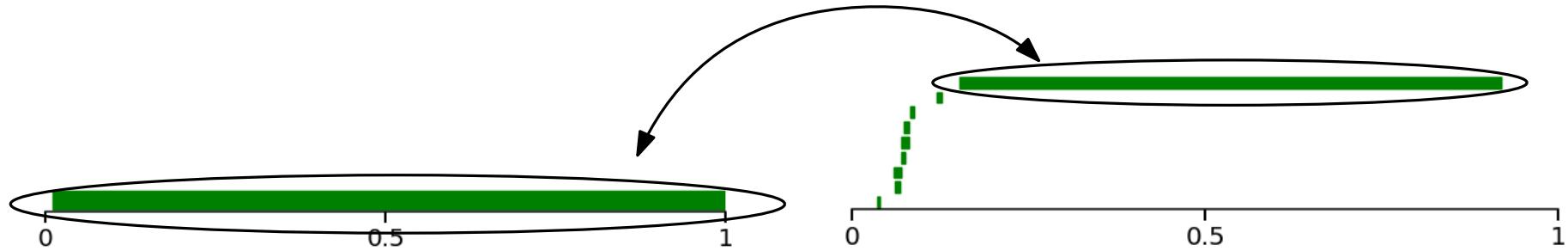
Definition: The **bottleneck distance** between P and Q is defined as the infimum of costs over all the partial matchings:

$$d_b(P, Q) = \inf \{ \text{cost}(M) \mid M \text{ is a partial matching between } P \text{ and } Q \}.$$

Distância bottleneck

22/46 (9/12)

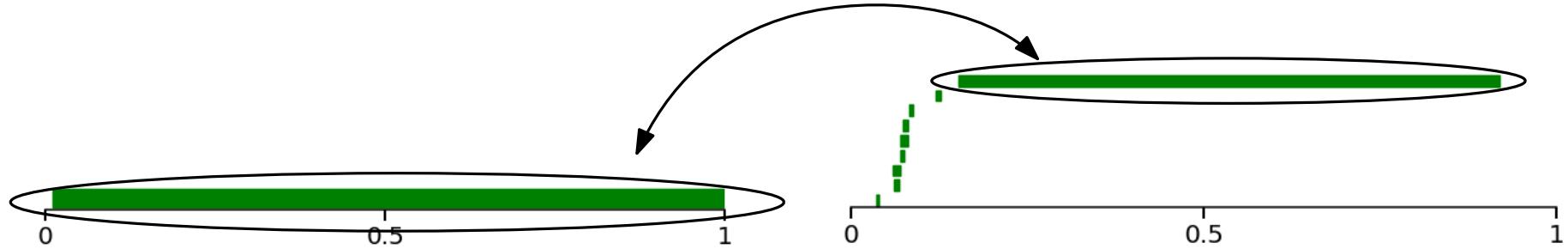
Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.



Definition: The **bottleneck distance** between P and Q is defined as the infimum of costs over all the partial matchings:

$$d_b(P, Q) = \inf \{ \text{cost}(M) \mid M \text{ is a partial matching between } P \text{ and } Q \}.$$

Consider two barcodes P and Q , that is, multisets of intervals $\{(a_i, b_i), i \in \mathcal{I}\}$ of $(\overline{\mathbb{R}^+})^2$ such that $a_i \leq b_i$ for all $i \in \mathcal{I}$.



Definition: The **bottleneck distance** between P and Q is defined as the infimum of costs over all the partial matchings:

$$d_b(P, Q) = \inf \{ \text{cost}(M) \mid M \text{ is a partial matching between } P \text{ and } Q \}.$$

If \mathbb{U} and \mathbb{V} are two decomposable persistence modules, we define their **bottleneck distance** as

$$d_b(\mathbb{U}, \mathbb{V}) = d_b(\text{Barcode}(\mathbb{U}), \text{Barcode}(\mathbb{V})).$$

Distância bottleneck

22/46 (11/12)

Example: Consider $a \leq b$, $a' \leq b'$, and the barcodes $P = \{[a, b]\}$ and $Q = \{[a', b']\}$.



First matching: the empty matching $M = \emptyset$. The intervals are matched to their midpoint, and the cost is

$$\left| (a, b) - \left(\frac{a+b}{2}, \frac{a+b}{2} \right) \right|_{\infty} = \frac{b-a}{2}, \quad \left| (a', b') - \left(\frac{a'+b'}{2}, \frac{a'+b'}{2} \right) \right|_{\infty} = \frac{b'-a'}{2}$$

The total cost is $\text{cost}(M) = \max \left\{ \frac{b-a}{2}, \frac{b'-a'}{2} \right\}$.

Second matching: $M' = \{((a, b), (a', b'))\}$. The intervals are matched together, and the cost of the pair is

$$|(a, b) - (a', b')|_{\infty} = \max\{|a - a'|, |b - b'|\}.$$

which is also $\text{cost}(M')$.

These are the only two partial matchings, and we deduce the bottleneck distance

$$d_b(P, Q) = \min \left\{ \max \left\{ \frac{b-a}{2}, \frac{b'-a'}{2} \right\}, \max\{|a - a'|, |b - b'|\} \right\}.$$

Distância bottleneck

22/46 (12/12)

Example: Consider $a \leq b$, $a' \leq b'$, and the barcodes $P = \{[a, b]\}$ and $Q = \{[a', b']\}$.



We have

$$d_b(P, Q) = \min \left\{ \max \left\{ \frac{b-a}{2}, \frac{b'-a'}{2} \right\}, \max\{|a-a'|, |b-b'|\} \right\}.$$

Example: Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Their barcodes are the sets P and Q of the previous example, from which we deduce

$$d_b(\mathbb{B}[a, b], \mathbb{B}[a', b']) = \min \left\{ \max \left\{ \frac{b-a}{2}, \frac{b'-a'}{2} \right\}, \max\{|a-a'|, |b-b'|\} \right\}.$$

I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

Distância de entrelaçamento

24/46 (1/12)

Consider two persistence modules \mathbb{V} and \mathbb{W} :

$$\cdots \rightarrow V^{t_1} \xrightarrow{v_{t_1}^{t_2}} V^{t_2} \xrightarrow{v_{t_2}^{t_3}} V^{t_3} \xrightarrow{v_{t_3}^{t_4}} V^{t_4} \cdots$$

$$\cdots \rightarrow W^{t_1} \xrightarrow{w_{t_1}^{t_2}} W^{t_2} \xrightarrow{w_{t_2}^{t_3}} W^{t_3} \xrightarrow{w_{t_3}^{t_4}} W^{t_4} \cdots$$

Given $\epsilon \geq 0$, an **ϵ -morphism** between \mathbb{V} and \mathbb{W} is a family of linear maps $\phi = (\phi_t : V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^{s+\epsilon} & \xrightarrow{w_{s+\epsilon}^{t+\epsilon}} & W^{t+\epsilon} \end{array}$$

Distância de entrelaçamento

24/46 (2/12)

Consider two persistence modules \mathbb{V} and \mathbb{W} :

$$\cdots \rightarrow V^{t_1} \xrightarrow{v_{t_1}^{t_2}} V^{t_2} \xrightarrow{v_{t_2}^{t_3}} V^{t_3} \xrightarrow{v_{t_3}^{t_4}} V^{t_4} \cdots$$

$$\cdots \rightarrow W^{t_1} \xrightarrow{w_{t_1}^{t_2}} W^{t_2} \xrightarrow{w_{t_2}^{t_3}} W^{t_3} \xrightarrow{w_{t_3}^{t_4}} W^{t_4} \cdots$$

Given $\epsilon \geq 0$, an **ϵ -morphism** between \mathbb{V} and \mathbb{W} is a family of linear maps $\phi = (\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^{s+\epsilon} & \xrightarrow{w_{s+\epsilon}^{t+\epsilon}} & W^{t+\epsilon} \end{array}$$

An **ϵ -interleaving** between \mathbb{V} and \mathbb{W} is a pair of ϵ -morphisms $(\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ and $(\psi_t: W^t \rightarrow V^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagrams commute for every $t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^t & \xrightarrow{v_t^{t+2\epsilon}} & V^{t+2\epsilon} \\ & \searrow \phi_t \quad \nearrow \psi_{t+\epsilon} & \\ & W^{t+\epsilon} & \end{array}$$

$$\begin{array}{ccc} & V^{t+\epsilon} & \\ & \nearrow \psi_t \quad \searrow \phi_{t+\epsilon} & \\ W^t & \xrightarrow{w_t^{t+2\epsilon}} & W^{t+2\epsilon} \end{array}$$

Distância de entrelaçamento

24/46 (3/12)

Consider two persistence modules \mathbb{V} and \mathbb{W} :

$$\cdots \rightarrow V^{t_1} \xrightarrow{v_{t_1}^{t_2}} V^{t_2} \xrightarrow{v_{t_2}^{t_3}} V^{t_3} \xrightarrow{v_{t_3}^{t_4}} V^{t_4} \cdots$$

$$\cdots \rightarrow W^{t_1} \xrightarrow{w_{t_1}^{t_2}} W^{t_2} \xrightarrow{w_{t_2}^{t_3}} W^{t_3} \xrightarrow{w_{t_3}^{t_4}} W^{t_4} \cdots$$

Given $\epsilon \geq 0$, an **ϵ -morphism** between \mathbb{V} and \mathbb{W} is a family of linear maps $\phi = (\phi_t : V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^{s+\epsilon} & \xrightarrow{w_{s+\epsilon}^{t+\epsilon}} & W^{t+\epsilon} \end{array}$$

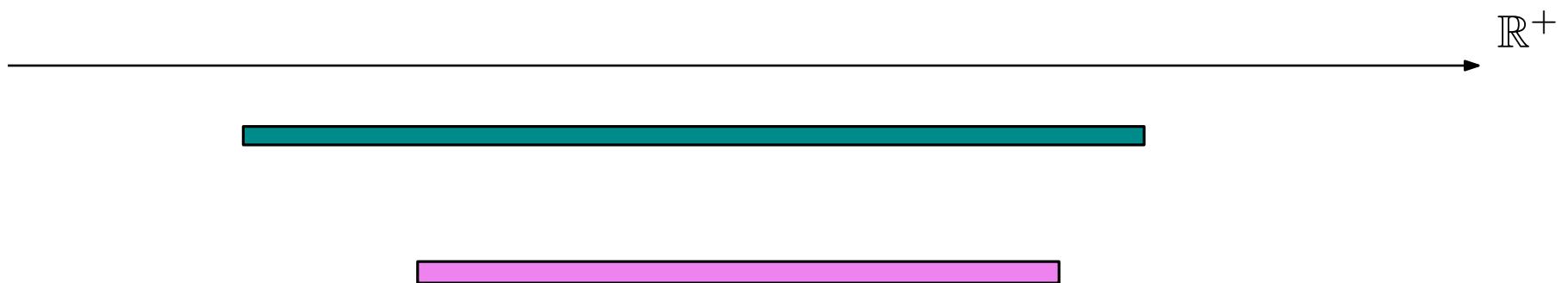
An **ϵ -interleaving** between \mathbb{V} and \mathbb{W} is a pair of ϵ -morphisms $(\phi_t : V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ and $(\psi_t : W^t \rightarrow V^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagrams commute for every $t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^t & \xrightarrow{v_t^{t+2\epsilon}} & V^{t+2\epsilon} \\ \searrow \phi_t & \nearrow \psi_{t+\epsilon} & \\ W^{t+\epsilon} & & \end{array} \quad \begin{array}{ccc} & V^{t+\epsilon} & \\ \nearrow \psi_t & & \searrow \phi_{t+\epsilon} \\ W^t & \xrightarrow{w_t^{t+2\epsilon}} & W^{t+2\epsilon} \end{array}$$

The **interleaving distance** is: $d_i(\mathbb{V}, \mathbb{W}) = \inf\{\epsilon \geq 0 \mid \mathbb{V} \text{ and } \mathbb{W} \text{ are } \epsilon\text{-interleaved}\}$.

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.

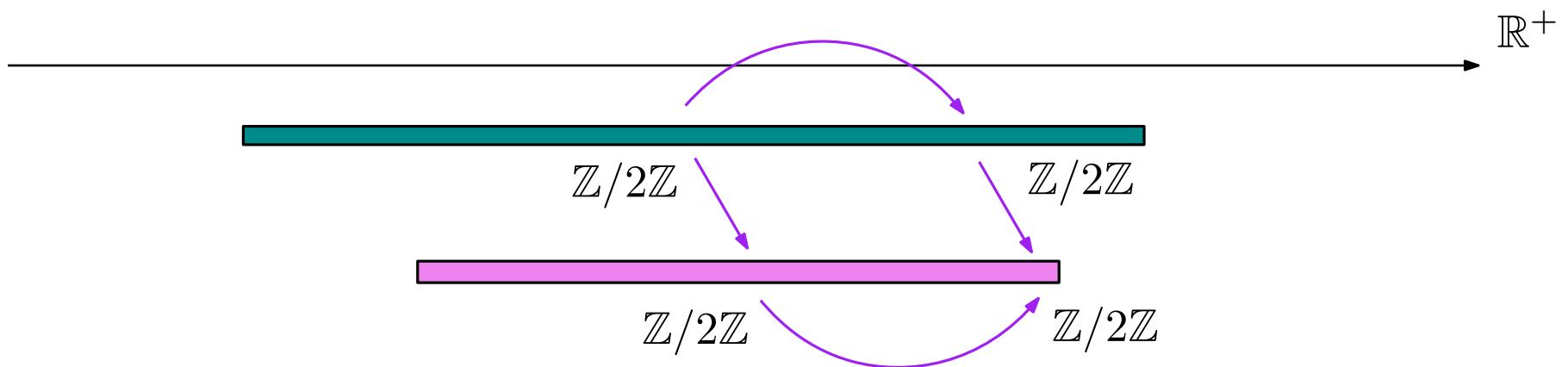


Distância de entrelaçamento

24/46 (5/12)

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.



Given $\epsilon \geq 0$, an ϵ -**morphism** between \mathbb{V} and \mathbb{W} is a family of linear maps $\phi = (\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

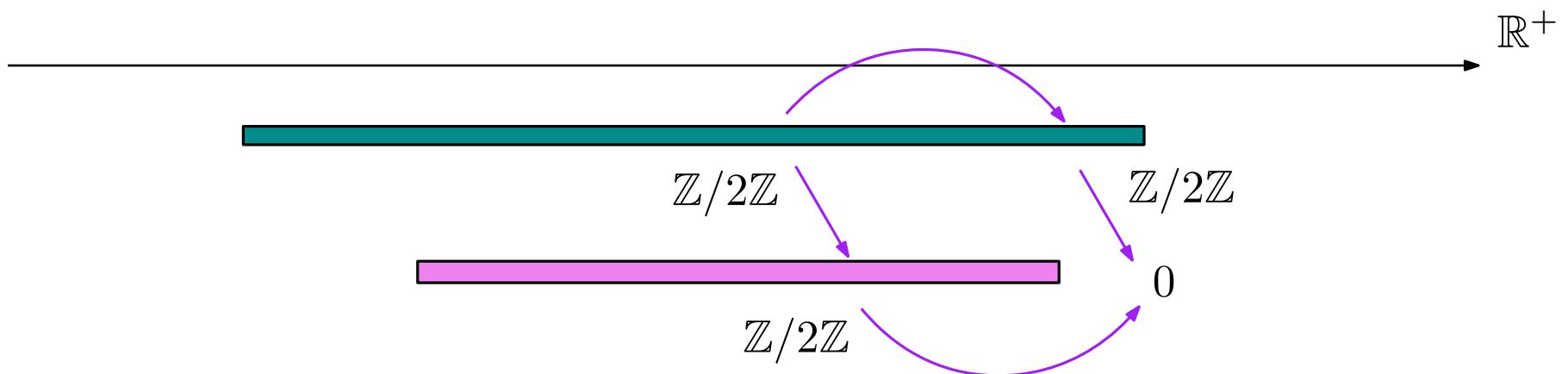
$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^{s+\epsilon} & \xrightarrow{w_{s+\epsilon}^{t+\epsilon}} & W^{t+\epsilon} \end{array}$$

Distância de entrelaçamento

24/46 (6/12)

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.

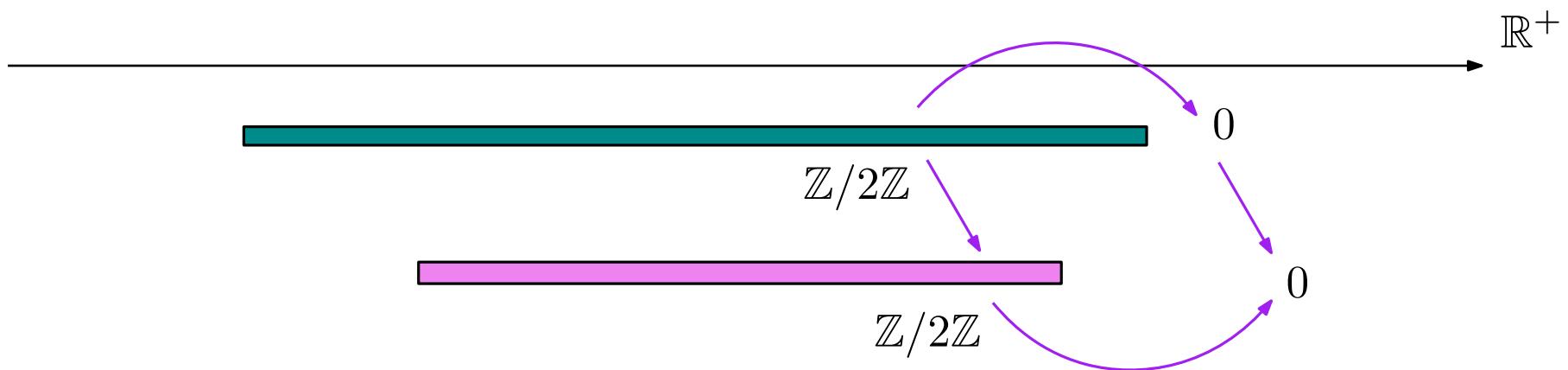


Given $\epsilon \geq 0$, an ϵ -**morphism** between \mathbb{V} and \mathbb{W} is a family of linear maps $\phi = (\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^{s+\epsilon} & \xrightarrow{w_{s+\epsilon}^{t+\epsilon}} & W^{t+\epsilon} \end{array}$$

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.



Given $\epsilon \geq 0$, an ϵ -**morphism** between \mathbb{V} and \mathbb{W} is a family of linear maps $\phi = (\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

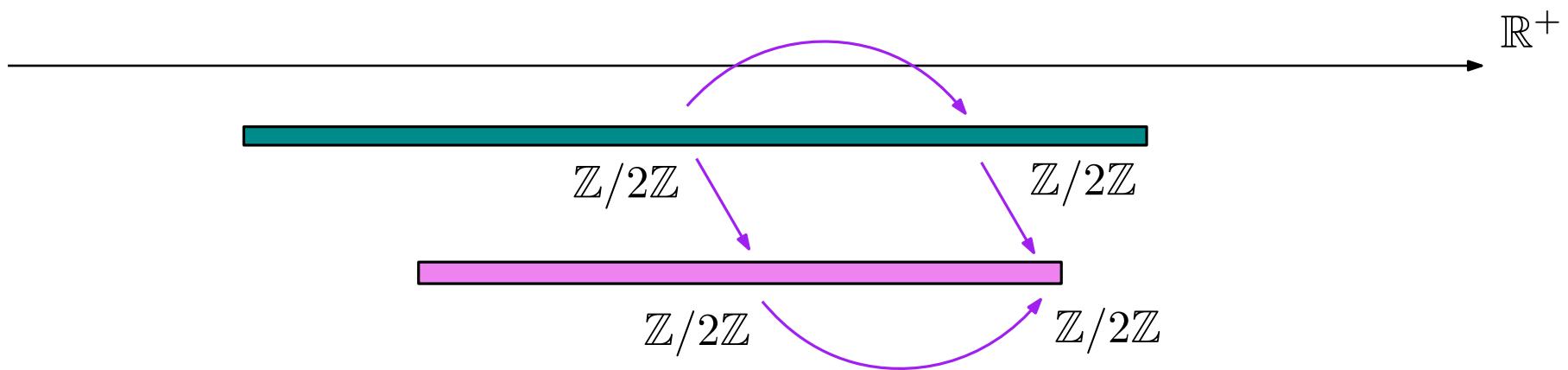
$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^{s+\epsilon} & \xrightarrow{w_{s+\epsilon}^{t+\epsilon}} & W^{t+\epsilon} \end{array}$$

Distância de entrelaçamento

24/46 (8/12)

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.



Given $\epsilon \geq 0$, an ϵ -**morphism** between \mathbb{V} and \mathbb{W} is a family of linear maps $\phi = (\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagram commutes for every $s \leq t \in \mathbb{R}^+$:

$$\begin{array}{ccc} V^s & \xrightarrow{v_s^t} & V^t \\ \downarrow \phi_s & & \downarrow \phi_t \\ W^{s+\epsilon} & \xrightarrow{w_{s+\epsilon}^{t+\epsilon}} & W^{t+\epsilon} \end{array}$$

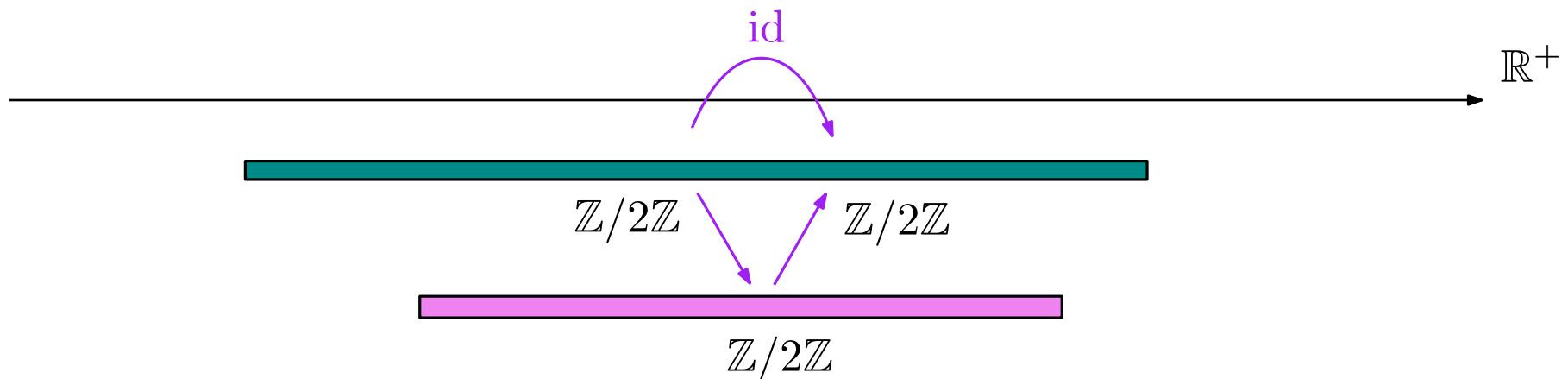
- Only two possibilities for ϕ :
- always the zero map
 - always nonzero when V^t and $W^{t+\epsilon}$ are nonzero

Distância de entrelaçamento

24/46 (9/12)

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.



An **ϵ -interleaving** between \mathbb{V} and \mathbb{W} is a pair of ϵ -morphisms $(\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ and $(\psi_t: W^t \rightarrow V^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagrams commute for every $t \in \mathbb{R}^+$:

$$\begin{array}{ccc}
 V^t & \xrightarrow{v_t^{t+2\epsilon}} & V^{t+2\epsilon} \\
 \phi_t \searrow & \nearrow \psi_{t+\epsilon} & \\
 & W^{t+\epsilon} &
 \end{array}
 \quad
 \begin{array}{ccc}
 & V^{t+\epsilon} & \\
 \psi_t \nearrow & \downarrow w_t^{t+2\epsilon} & \searrow \phi_{t+\epsilon} \\
 W^t & \xrightarrow{w_t^{t+2\epsilon}} & W^{t+2\epsilon}
 \end{array}$$

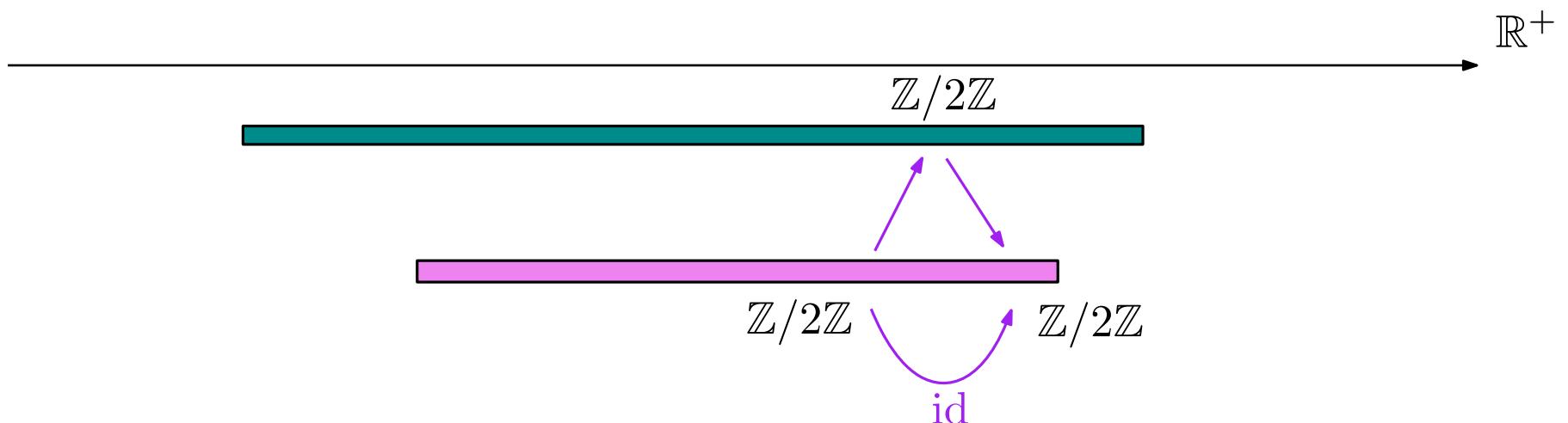
→ $\psi_{t+\epsilon} \circ \phi_t$ must be nonzero when $[t, t + \epsilon] \subset [a, b]$

Distância de entrelaçamento

24/46 (10/12)

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.



An ϵ -**interleaving** between \mathbb{V} and \mathbb{W} is a pair of ϵ -morphisms $(\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ and $(\psi_t: W^t \rightarrow V^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagrams commute for every $t \in \mathbb{R}^+$:

$$\begin{array}{ccc}
 V^t & \xrightarrow{v_t^{t+2\epsilon}} & V^{t+2\epsilon} \\
 \phi_t \searrow & \nearrow \psi_{t+\epsilon} & \\
 & W^{t+\epsilon} &
 \end{array}
 \quad
 \begin{array}{ccc}
 & V^{t+\epsilon} & \\
 \psi_t \nearrow & \downarrow w_t^{t+2\epsilon} & \searrow \phi_{t+\epsilon} \\
 W^t & \xrightarrow{w_t^{t+2\epsilon}} & W^{t+2\epsilon}
 \end{array}$$

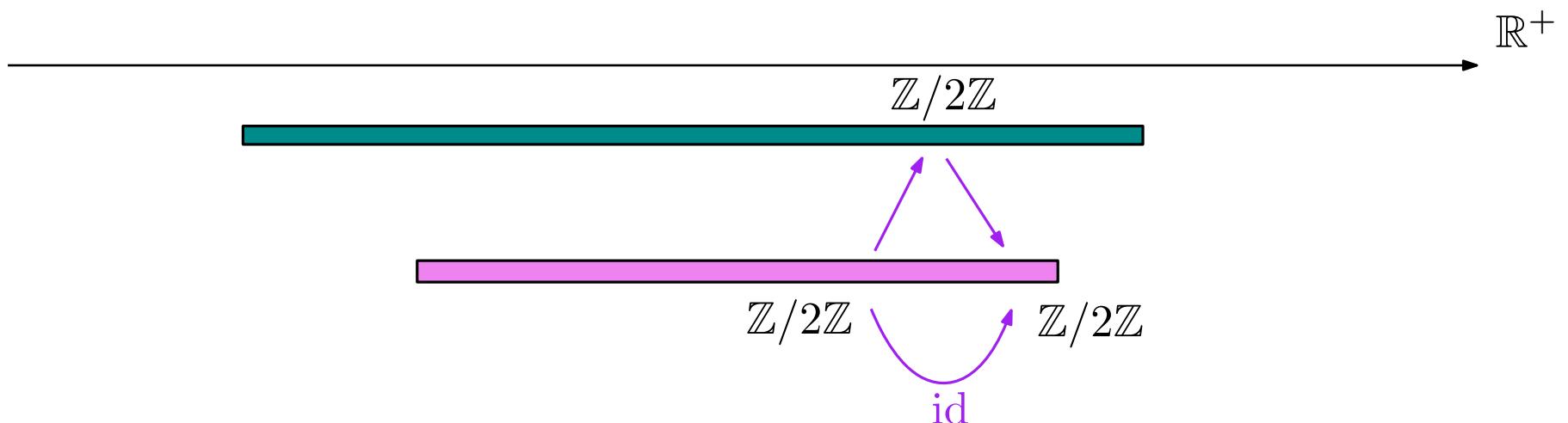
→ $\psi_{t+\epsilon} \circ \phi_t$ must be nonzero when $[t, t + \epsilon] \subset [a, b]$

Distância de entrelaçamento

24/46 (11/12)

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.



An ϵ -**interleaving** between \mathbb{V} and \mathbb{W} is a pair of ϵ -morphisms $(\phi_t: V^t \rightarrow W^{t+\epsilon})_{t \in \mathbb{R}^+}$ and $(\psi_t: W^t \rightarrow V^{t+\epsilon})_{t \in \mathbb{R}^+}$ such that the following diagrams commute for every $t \in \mathbb{R}^+$:

$$\begin{array}{ccc}
 V^t & \xrightarrow{v_t^{t+2\epsilon}} & V^{t+2\epsilon} \\
 \searrow \phi_t & \nearrow \psi_{t+\epsilon} & \\
 & W^{t+\epsilon} &
 \end{array}
 \quad
 \begin{array}{ccc}
 & V^{t+\epsilon} & \\
 \nearrow \psi_t & & \searrow \phi_{t+\epsilon} \\
 W^t & \xrightarrow{w_t^{t+2\epsilon}} & W^{t+2\epsilon}
 \end{array}$$

→ $\psi_{t+\epsilon} \circ \phi_t$ must be nonzero when $[t, t + \epsilon] \subset [a, b]$
 $\phi_{t+\epsilon} \circ \psi_t$ must be nonzero when $[t, t + \epsilon] \subset [a', b']$

Example: Let $a, a', b, b' \in \mathbb{R}^+$ such that $a \leq b$ and $a' \leq b'$. Consider the interval-modules $\mathbb{B}[a, b]$ and $\mathbb{B}[a', b']$.

Let us find an ϵ -interleaving.



- Only two possibilities for ϕ : • always the zero map
• always nonzero when V^t and $W^{t+\epsilon}$ are nonzero
- $\psi_{t+\epsilon} \circ \phi_t$ must be nonzero when $[t, t + \epsilon] \subset [a, b]$
 $\phi_{t+\epsilon} \circ \psi_t$ must be nonzero when $[t, t + \epsilon] \subset [a', b']$

We deduce that either

- $|a - b| \leq 2\epsilon$ and $|a' - b'| \leq 2\epsilon$, or
- $|a - a'| \leq \epsilon$ and $|b - b'| \leq \epsilon$

Conclusion: $d_i(\mathbb{B}[a, b], \mathbb{B}[a', b']) = \min \left\{ \max \left\{ \frac{b-a}{2}, \frac{b'-a'}{2} \right\}, \max\{|a-a'|, |b-b'|\} \right\}$

Theorem (Chazal, de Silva, Glisse, Oudot, 2009): If the persistence modules \mathbb{U} and \mathbb{V} are interval-decomposable, then $d_i(\mathbb{U}, \mathbb{V}) = d_b(\mathbb{U}, \mathbb{V})$.

→ **Stability:** $d_i(\mathbb{U}, \mathbb{V}) \geq d_b(\mathbb{U}, \mathbb{V})$

→ **Converse stability:** $d_i(\mathbb{U}, \mathbb{V}) \leq d_b(\mathbb{U}, \mathbb{V})$

Teorema de isometria

25/46 (2/3)

Theorem (Chazal, de Silva, Glisse, Oudot, 2009): If the persistence modules \mathbb{U} and \mathbb{V} are interval-decomposable, then $d_i(\mathbb{U}, \mathbb{V}) = d_b(\mathbb{U}, \mathbb{V})$.

Proof: Let us write the decomposition of the persistence modules in intervals:

$$\mathbb{V} \simeq \bigoplus_{I \in \mathcal{I}} \mathbb{B}[I]$$

$$\mathbb{W} \simeq \bigoplus_{J \in \mathcal{J}} \mathbb{B}[J]$$

Suppose that we have a ϵ -partial matching $M \subset \mathcal{I} \times \mathcal{J}$. This gives a matching of some intervals (I, J) , where $I = (a, b)$ and $J = (a', b')$, such that $|a - a'| \leq \epsilon$ and $|b - b'| \leq \epsilon$.

We can build an ϵ -interleaving between $\mathbb{B}[I]$ and $\mathbb{B}[J]$, that we denote $(\phi_{(I,J)}, \psi_{(I,J)})$.

Some intervals I (resp. J) are not matched, in which case their length is not greater than 2ϵ , and we can build an ϵ -interleaving with the zero persistence module. We denote this interleaving $(\phi_{(I,0)}, \psi_{(I,0)})$ (resp. $(\phi_{(0,J)}, \psi_{(0,J)})$).

Now, let us consider the sums of all these linear maps:

$$\bar{\phi} = \bigoplus_{\substack{(I,J) \text{ matched}}} \phi_{(I,J)} + \bigoplus_{\substack{I \text{ not matched}}} \phi_{(I,0)}, \quad \bar{\psi} = \bigoplus_{\substack{(I,J) \text{ matched}}} \psi_{(I,J)} + \bigoplus_{\substack{J \text{ not matched}}} \phi_{(0,J)}$$

————— ➔ $(\bar{\phi}, \bar{\psi})$ is an ϵ -interleaving ————— ➔ $d_i(\mathbb{U}, \mathbb{V}) \leq d_b(\mathbb{U}, \mathbb{V})$

Theorem (Chazal, de Silva, Glisse, Oudot, 2009): If the persistence modules \mathbb{U} and \mathbb{V} are interval-decomposable, then $d_i(\mathbb{U}, \mathbb{V}) = d_b(\mathbb{U}, \mathbb{V})$.

→ **Stability:** $d_i(\mathbb{U}, \mathbb{V}) \geq d_b(\mathbb{U}, \mathbb{V})$

→ **Converse stability:** $d_i(\mathbb{U}, \mathbb{V}) \leq d_b(\mathbb{U}, \mathbb{V})$

The stability part is more difficult.

A first strategy uses the interpolation lemma, and concludes with the box lemma.

Interpolation lemma: If \mathbb{U} and \mathbb{V} are δ -interleaved, then there exists a family of persistence modules $(\mathbb{U}_t)_{t \in [0, \delta]}$ such that $\mathbb{U}_0 = \mathbb{U}$, $\mathbb{U}_\delta = \mathbb{V}$ and $d_i(\mathbb{U}_s, \mathbb{U}_t) \leq |s - t|$ for every $s, t \in [0, \delta]$.

Another proof builds an explicit partial matching from an interleaving (Bauer, Lesnick, 2013).

I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

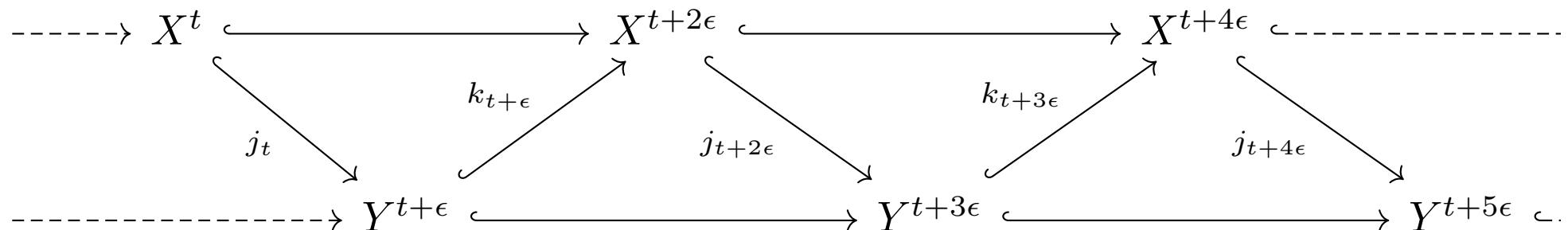
De volta aos espessamentos

27/46 (1/8)

Let X and Y be two subsets of \mathbb{R}^n . Define $\epsilon = d_H(X, Y)$ (Hausdorff distance).

We have seen that $X \subset Y^\epsilon$ and $Y \subset X^\epsilon$. We even have that $X^t \subset Y^{t+\epsilon}$ and $Y^t \subset X^{t+\epsilon}$ for all $t \geq 0$.

By denoting j and k these inclusions, we have a commutative diagram



De volta aos espessamentos

27/46 (2/8)

Let X and Y be two subsets of \mathbb{R}^n . Define $\epsilon = d_H(X, Y)$ (Hausdorff distance).

We have seen that $X \subset Y^\epsilon$ and $Y \subset X^\epsilon$. We even have that $X^t \subset Y^{t+\epsilon}$ and $Y^t \subset X^{t+\epsilon}$ for all $t \geq 0$.

By denoting j and k these inclusions, we have a commutative diagram

$$\begin{array}{ccccccc}
 & \dashrightarrow & X^t & \xrightarrow{\hspace{2cm}} & X^{t+2\epsilon} & \xrightarrow{\hspace{2cm}} & X^{t+4\epsilon} & \dashleftarrow \\
 & \searrow j_t & & \nearrow k_{t+\epsilon} & & \nearrow k_{t+3\epsilon} & & \searrow j_{t+4\epsilon} \\
 & \dashrightarrow & Y^{t+\epsilon} & \xrightarrow{\hspace{2cm}} & Y^{t+3\epsilon} & \xrightarrow{\hspace{2cm}} & Y^{t+5\epsilon} & \dashleftarrow
 \end{array}$$

This also gives inclusions between Čech complexes:

$$\begin{array}{ccccccc}
 \check{\text{C}}\text{ech}^t(X) & \xrightarrow{\hspace{2cm}} & \check{\text{C}}\text{ech}^{t+2\epsilon}(X) & \xrightarrow{\hspace{2cm}} & \check{\text{C}}\text{ech}^{t+4\epsilon}(X) & \dashleftarrow & \\
 \searrow j_t & & \nearrow k_{t+\epsilon} & & \nearrow k_{t+3\epsilon} & & \searrow j_{t+4\epsilon} \\
 & \dashrightarrow & \check{\text{C}}\text{ech}^{t+\epsilon}(Y) & \xrightarrow{\hspace{2cm}} & \check{\text{C}}\text{ech}^{t+3\epsilon}(Y) & \xrightarrow{\hspace{2cm}} & \check{\text{C}}\text{ech}^{t+5\epsilon}(Y) & \dashleftarrow
 \end{array}$$

[...] This also gives inclusions between Čech complexes:

$$\begin{array}{ccccccc}
 \check{\text{C}}\text{ech}^t(X) & \xrightarrow{\quad} & \check{\text{C}}\text{ech}^{t+2\epsilon}(X) & \xrightarrow{\quad} & \check{\text{C}}\text{ech}^{t+4\epsilon}(X) & \xleftarrow{\quad} & \dots \\
 \downarrow j_t & & \nearrow k_{t+\epsilon} & & \nearrow k_{t+3\epsilon} & & \nearrow j_{t+4\epsilon} \\
 \dots & \rightarrow & \check{\text{C}}\text{ech}^{t+\epsilon}(Y) & \xrightarrow{\quad} & \check{\text{C}}\text{ech}^{t+3\epsilon}(Y) & \xrightarrow{\quad} & \check{\text{C}}\text{ech}^{t+5\epsilon}(Y)
 \end{array}$$

Now, we apply the i^{th} homology functor.

$$\begin{array}{ccccccc}
 H_i(\check{\text{C}}\text{ech}^t(X)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+2\epsilon}(X)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+4\epsilon}(X)) & \xleftarrow{\quad} & \dots \\
 \downarrow (j_t)_* & & \nearrow (k_{t+\epsilon})_* & & \nearrow (k_{t+3\epsilon})_* & & \nearrow (j_{t+4\epsilon})_* \\
 \dots & \rightarrow & H_i(\check{\text{C}}\text{ech}^{t+\epsilon}(Y)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+3\epsilon}(Y)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+5\epsilon}(Y))
 \end{array}$$

De volta aos espessamentos

27/46 (4/8)

[...] This also gives inclusions between Čech complexes:

$$\begin{array}{ccccccc}
 \check{\text{C}}\text{ech}^t(X) & \longrightarrow & \check{\text{C}}\text{ech}^{t+2\epsilon}(X) & \longrightarrow & \check{\text{C}}\text{ech}^{t+4\epsilon}(X) & \dashrightarrow & \\
 \downarrow j_t & & \nearrow k_{t+\epsilon} & & \nearrow k_{t+3\epsilon} & & \\
 \dashrightarrow & \check{\text{C}}\text{ech}^{t+\epsilon}(Y) & \longrightarrow & \check{\text{C}}\text{ech}^{t+3\epsilon}(Y) & \longrightarrow & \check{\text{C}}\text{ech}^{t+5\epsilon}(Y) &
 \end{array}$$

Now, we apply the i^{th} homology functor.

$$\begin{array}{ccccccc}
 H_i(\check{\text{C}}\text{ech}^t(X)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+2\epsilon}(X)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+4\epsilon}(X)) & \dashrightarrow & \\
 \downarrow (j_t)_* & \nearrow (k_{t+\epsilon})_* & \downarrow (j_{t+2\epsilon})_* & \nearrow (k_{t+3\epsilon})_* & \downarrow (j_{t+4\epsilon})_* & \nearrow & \\
 \dashrightarrow & H_i(\check{\text{C}}\text{ech}^{t+\epsilon}(Y)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+3\epsilon}(Y)) & \longrightarrow & H_i(\check{\text{C}}\text{ech}^{t+5\epsilon}(Y)) &
 \end{array}$$

→ persistence module of Čech complex of X

→ persistence module of Čech complex of Y

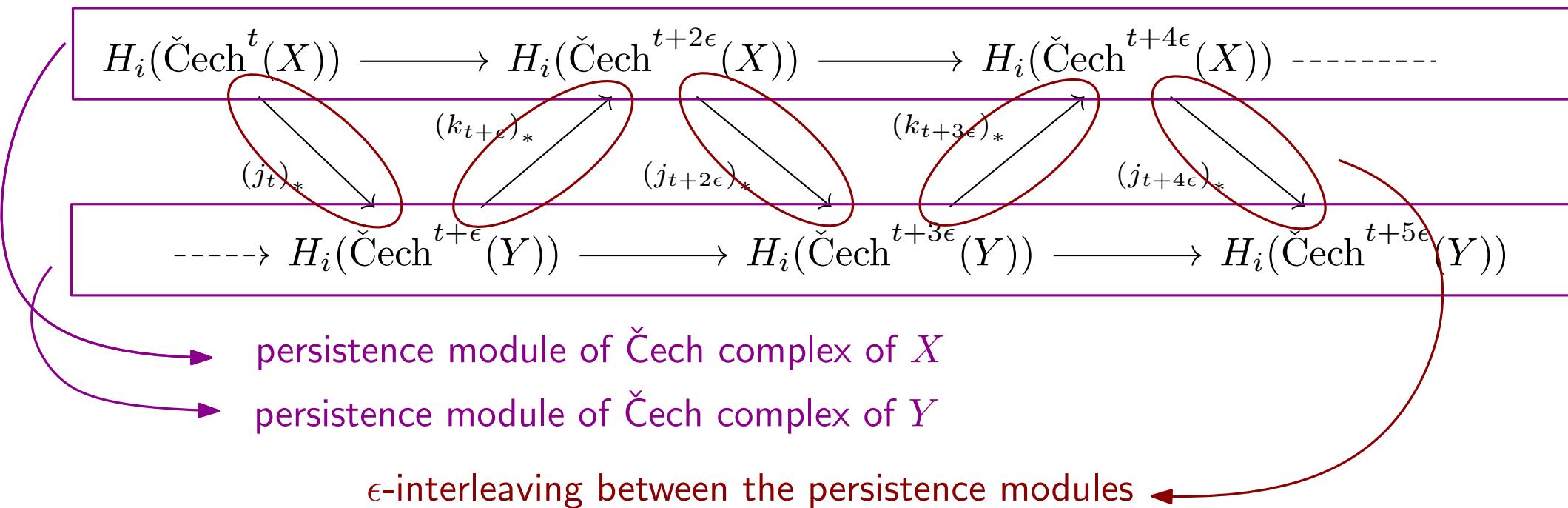
De volta aos espessamentos

27/46 (5/8)

[...] This also gives inclusions between Čech complexes:

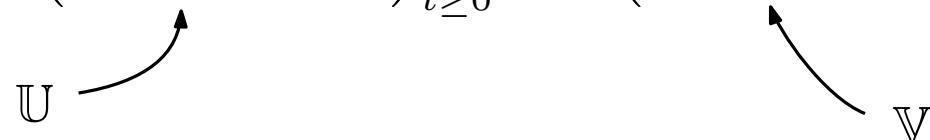
$$\begin{array}{ccccccc}
 \check{\text{Cech}}^t(X) & \xrightarrow{\quad} & \check{\text{Cech}}^{t+2\epsilon}(X) & \xrightarrow{\quad} & \check{\text{Cech}}^{t+4\epsilon}(X) & \xleftarrow{\quad} & \dots \\
 \downarrow j_t & & \nearrow k_{t+\epsilon} & & \nearrow k_{t+3\epsilon} & & \nearrow j_{t+4\epsilon} \\
 \dots & \rightarrow & \check{\text{Cech}}^{t+\epsilon}(Y) & \xrightarrow{\quad} & \check{\text{Cech}}^{t+3\epsilon}(Y) & \xrightarrow{\quad} & \check{\text{Cech}}^{t+5\epsilon}(Y)
 \end{array}$$

Now, we apply the i^{th} homology functor.



[...]

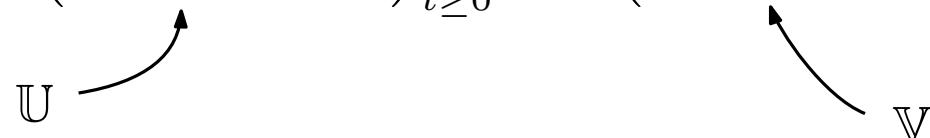
Hence the persistence modules $\left(H_i(\check{\text{Cech}}^t(X))\right)_{t \geq 0}$ and $\left(H_i(\check{\text{Cech}}^t(Y))\right)_{t \geq 0}$ are ϵ -interleaved.



Hence $d_i(\mathbb{U}, \mathbb{V}) \leq \epsilon$.

[...]

Hence the persistence modules $\left(H_i(\check{\text{Cech}}^t(X))\right)_{t \geq 0}$ and $\left(H_i(\check{\text{Cech}}^t(Y))\right)_{t \geq 0}$ are ϵ -interleaved.

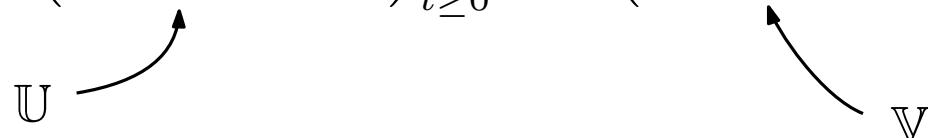


Hence $d_i(\mathbb{U}, \mathbb{V}) \leq \epsilon$.

We use the isometry theorem: $d_b(\mathbb{U}, \mathbb{V}) = d_i(\mathbb{U}, \mathbb{V}) \leq \epsilon$.

[...]

Hence the persistence modules $\left(H_i(\check{\text{Cech}}^t(X))\right)_{t \geq 0}$ and $\left(H_i(\check{\text{Cech}}^t(Y))\right)_{t \geq 0}$ are ϵ -interleaved.



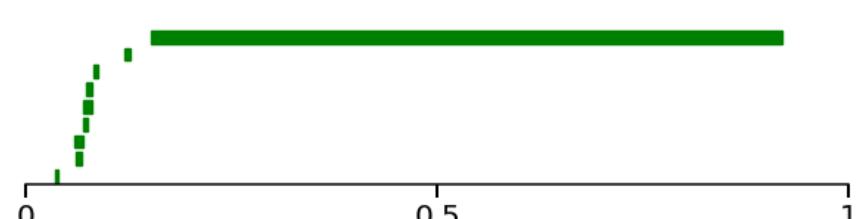
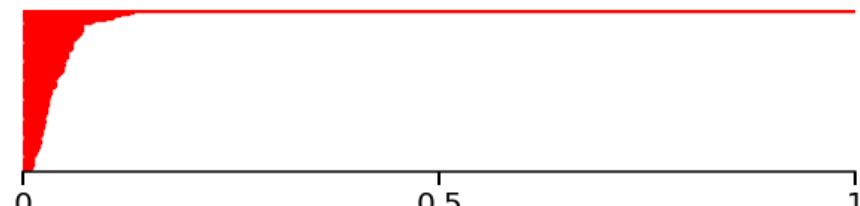
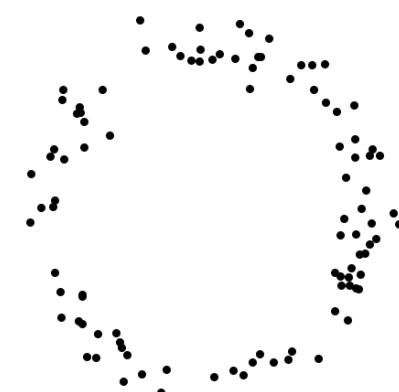
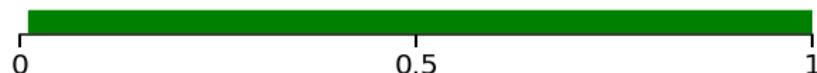
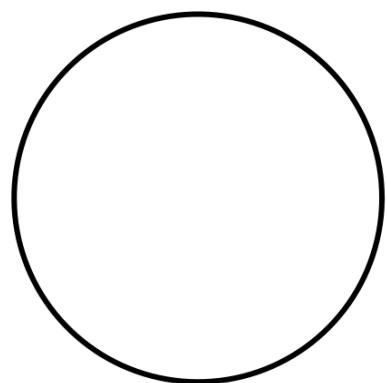
Hence $d_i(\mathbb{U}, \mathbb{V}) \leq \epsilon$.

We use the isometry theorem: $d_b(\mathbb{U}, \mathbb{V}) = d_i(\mathbb{U}, \mathbb{V}) \leq \epsilon$.

Theorem (Cohen-Steiner, Edelsbrunner, Harer, 2005): Let X and Y be two subsets of \mathbb{R}^n . Consider their Čech (resp. Rips) filtrations, and the corresponding i^{th} homology persistence modules, \mathbb{U} and \mathbb{V} . Suppose that they are interval-decomposables. Then $d_b(\mathbb{U}, \mathbb{V}) \leq d_H(X, Y)$.

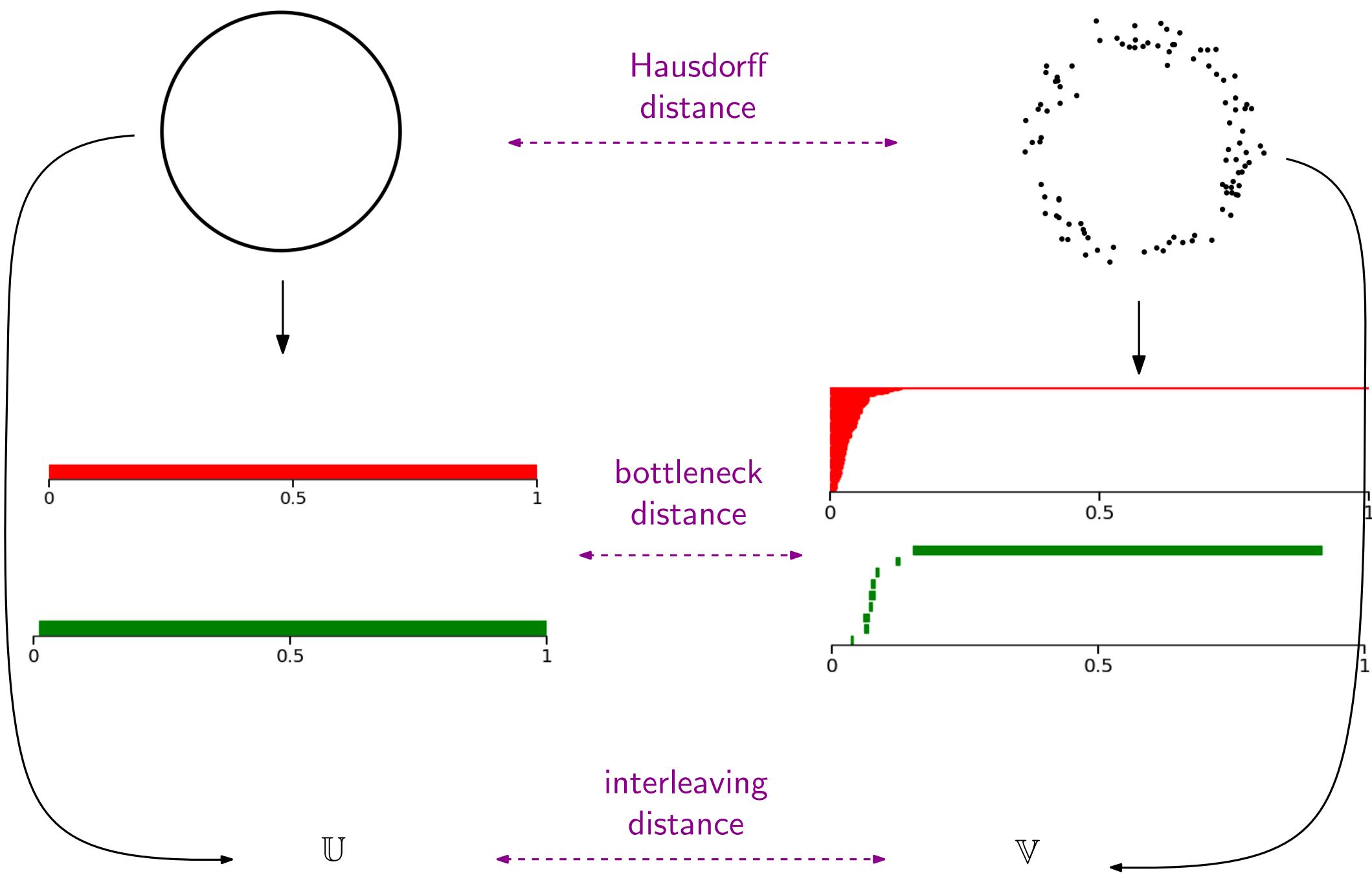
Resumo

28/46 (1/2)



Resumo

28/46 (2/2)



I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

Topological inference I

30/46 (1/2)

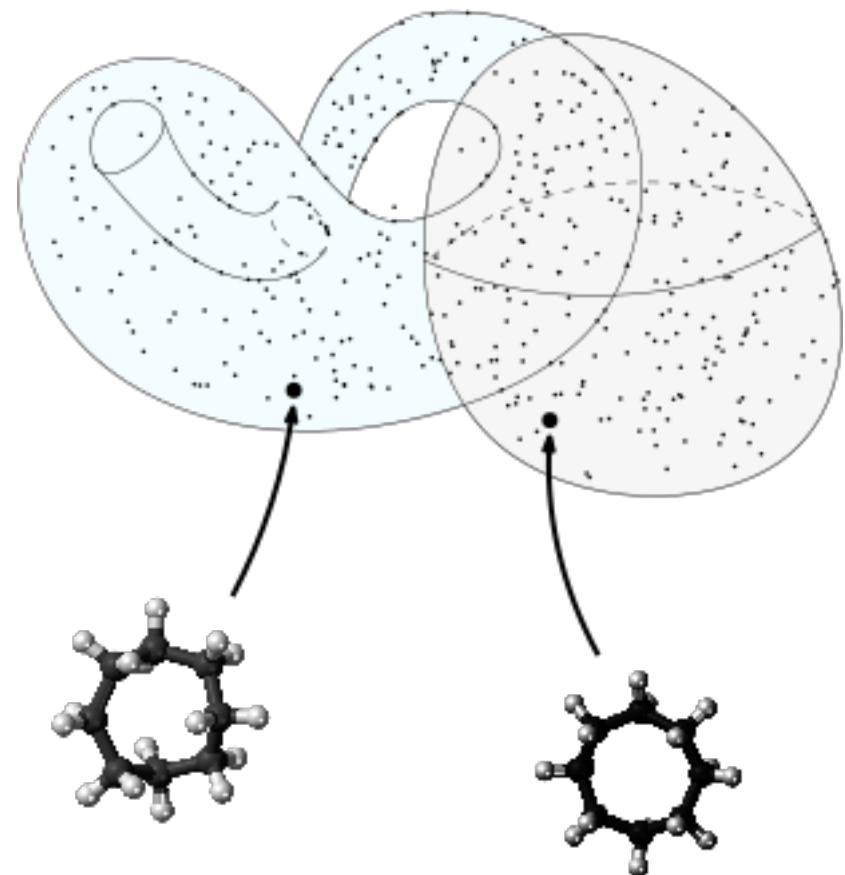
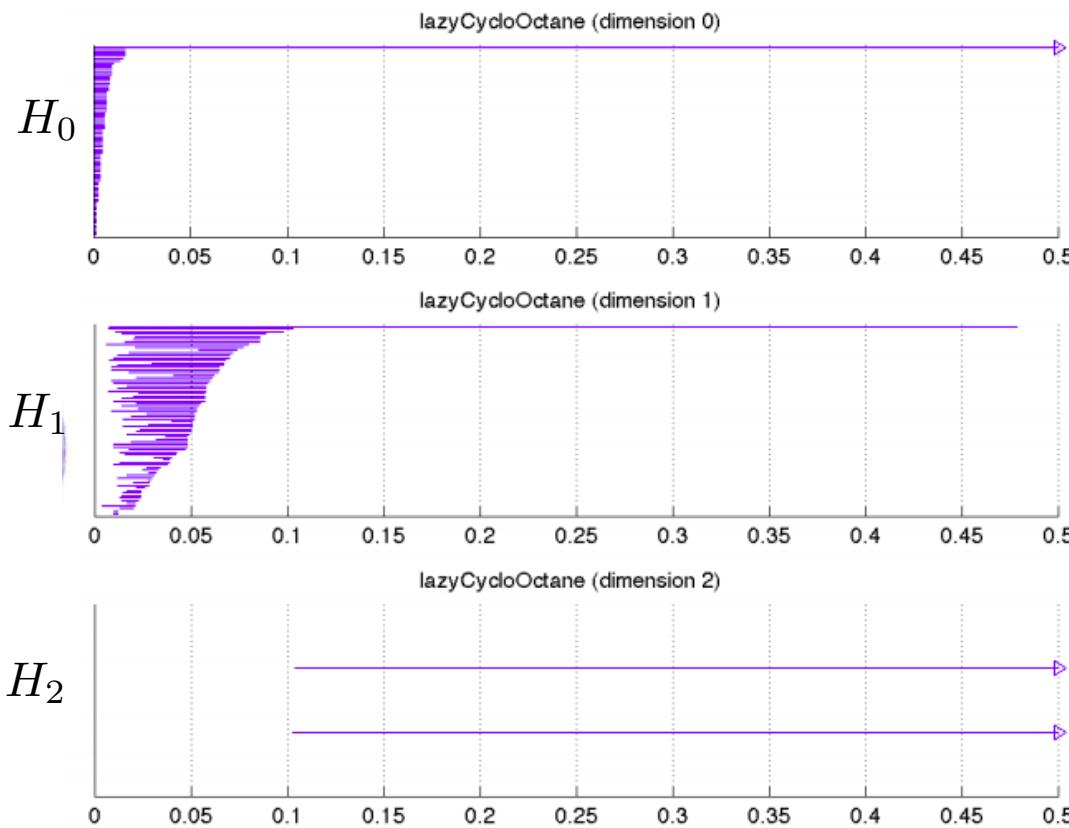
S. Martin, A. Thompson, E. A. Coutsias, and J-P. Watson, [Topology of cyclo-octane energy landscape](#), 2010

https://www.researchgate.net/publication/44697030_Topology_of_Cyclooctane_Energy_Landscape

The cyclo-octane molecule C_8H_{16} contains 24 atoms.

By generating many of these molecules, we obtain a point cloud in \mathbb{R}^{72} ($3 \times 24 = 72$).

We obtain the barcodes:



Topological inference I

30/46 (2/2)

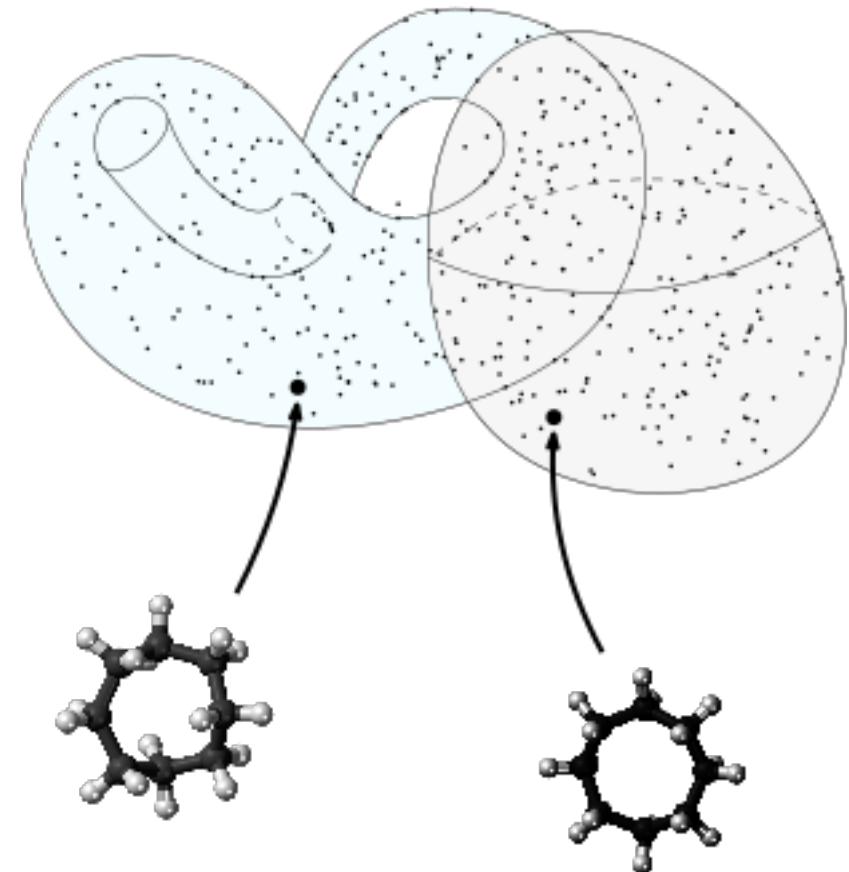
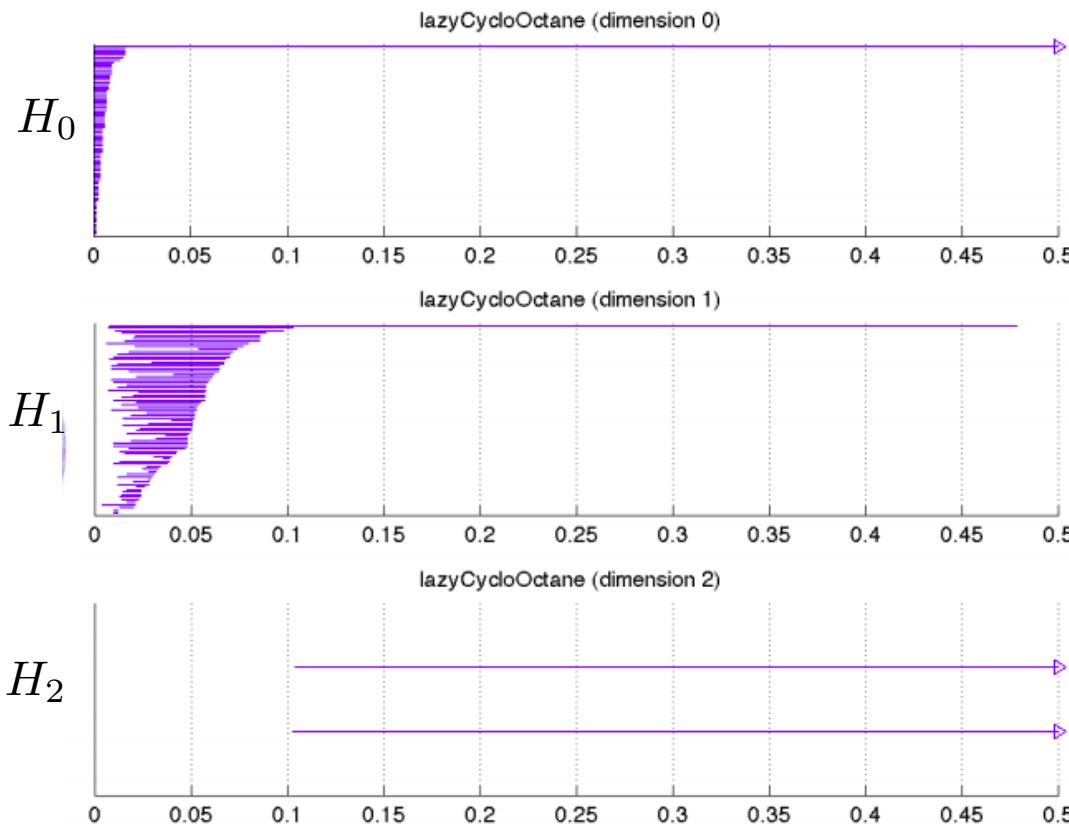
S. Martin, A. Thompson, E. A. Coutsias, and J-P. Watson, [Topology of cyclo-octane energy landscape](#), 2010

https://www.researchgate.net/publication/44697030_Topology_of_Cyclooctane_Energy_Landscape

The cyclo-octane molecule C_8H_{16} contains 24 atoms.

By generating many of these molecules, we obtain a point cloud in \mathbb{R}^{72} ($3 \times 24 = 72$).

We obtain the barcodes:



We deduce : $H_0 = \mathbb{Z}/2\mathbb{Z}$, $H_1 = \mathbb{Z}/2\mathbb{Z}$, $H_2 = (\mathbb{Z}/2\mathbb{Z})^2$

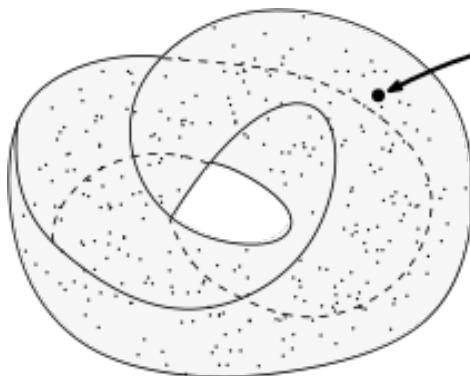
Topological inference II

31/46 (1/2)

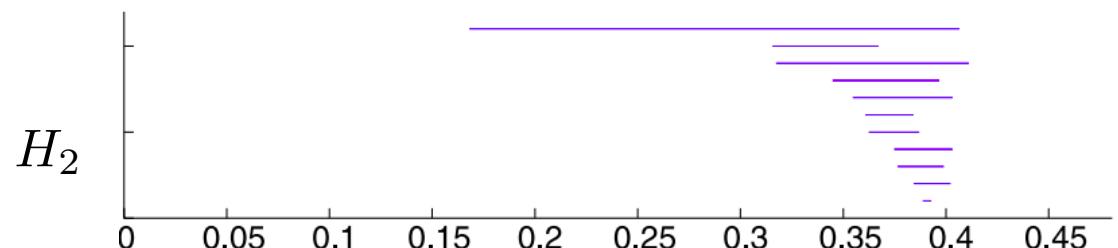
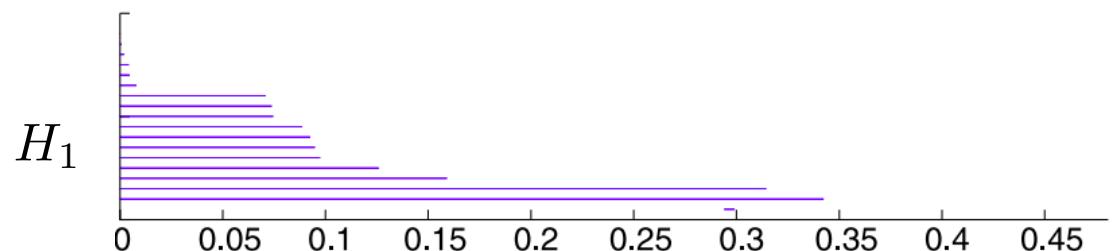
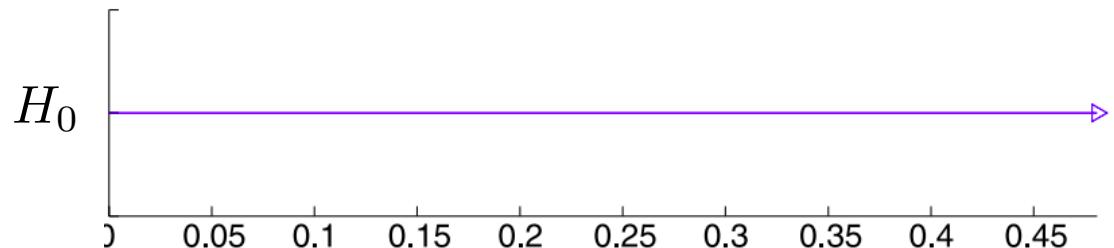
G. Carlsson, T. Ishkhanov, V. de Silva, and A. Zomorodian, [On the Local Behavior of Spaces of Natural Images](#), 2008

<https://link.springer.com/article/10.1007/s11263-007-0056-x>

From a large collection of natural images, the authors extract 3×3 patches. Since it consists of 9 pixels, each of these patches can be seen as a 9-dimensional vector, and the whole set as a point cloud in \mathbb{R}^9 .



We get the barcodes:



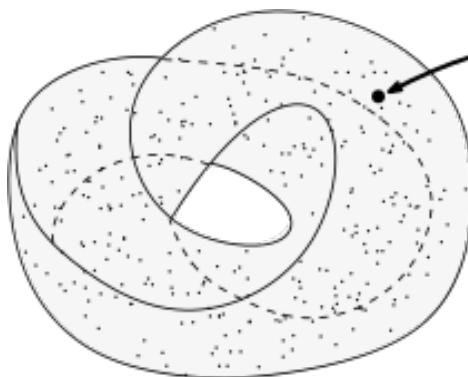
Topological inference II

31/46 (2/2)

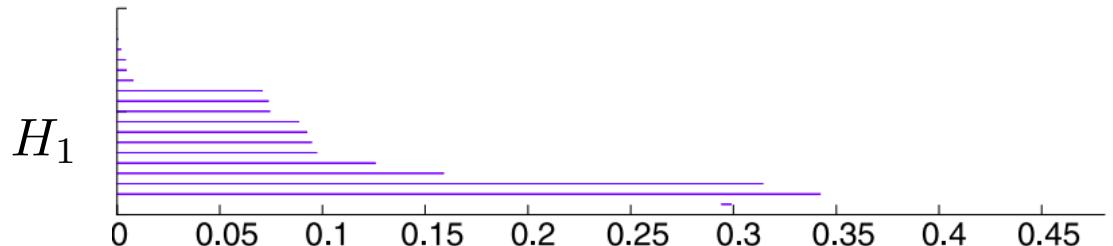
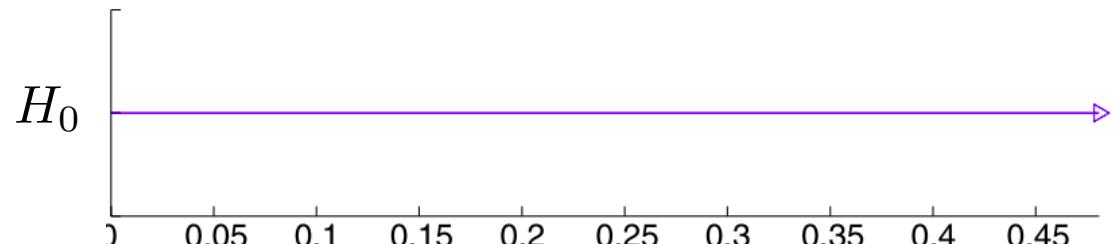
G. Carlsson, T. Ishkhanov, V. de Silva, and A. Zomorodian, [On the Local Behavior of Spaces of Natural Images](#), 2008

<https://link.springer.com/article/10.1007/s11263-007-0056-x>

From a large collection of natural images, the authors extract 3×3 patches. Since it consists of 9 pixels, each of these patches can be seen as a 9-dimensional vector, and the whole set as a point cloud in \mathbb{R}^9 .



We get the barcodes:



We deduce:

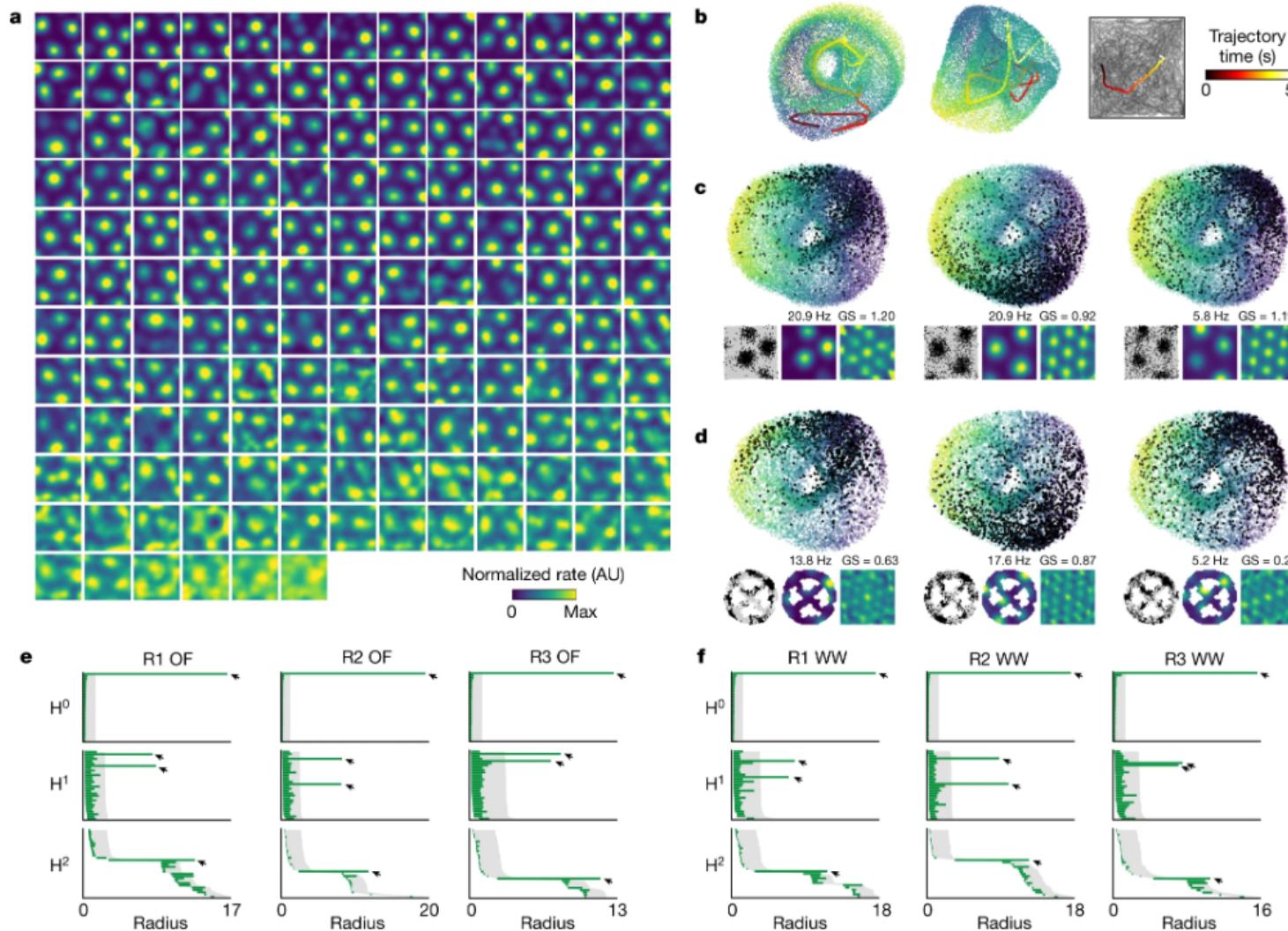
$$\begin{aligned} H_0 &= \mathbb{Z}/2\mathbb{Z}, \\ H_1 &= (\mathbb{Z}/2\mathbb{Z})^2, \\ H_2 &= \mathbb{Z}/2\mathbb{Z}. \end{aligned}$$

Inferência topológica III

32/46

[Richard J. Gardner et al, Toroidal topology of population activity in grid cells , 2022]

The authors recorded spikes from rat grid-cells, and applied dimensionality reduction to the firing matrix. By applying persistent homology, they observed the homology of a torus.

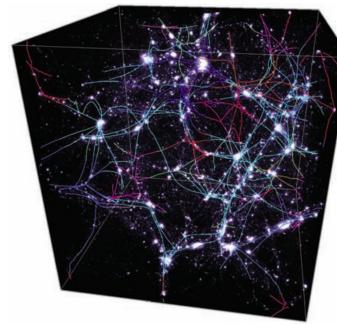
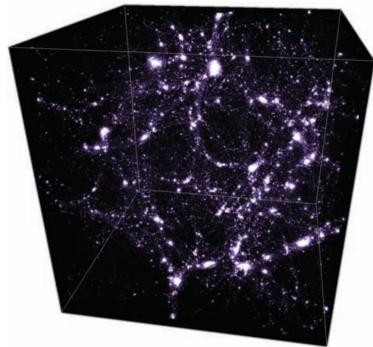


Multiscale analysis I

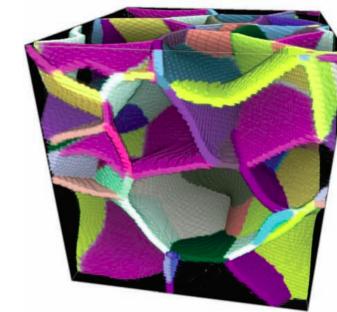
33/46

T. Sousbie, The persistent cosmic web and its filamentary structure, 2011

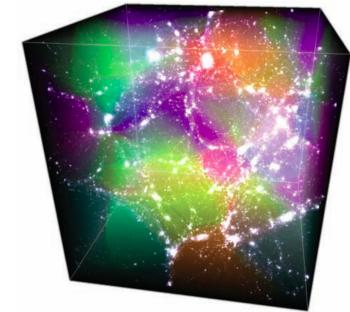
<https://www.giss.nasa.gov/staff/mway/cluster/sousbie2011mnras.pdf>



seen as an object
of dimension 1



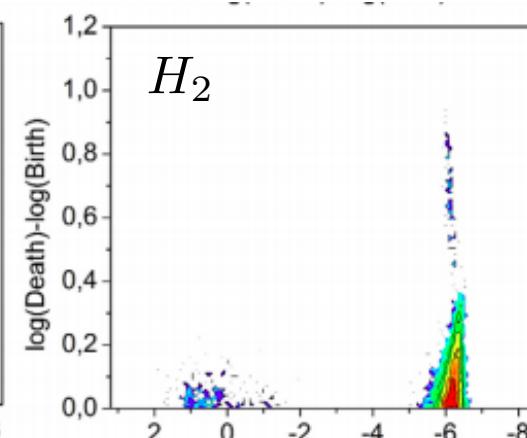
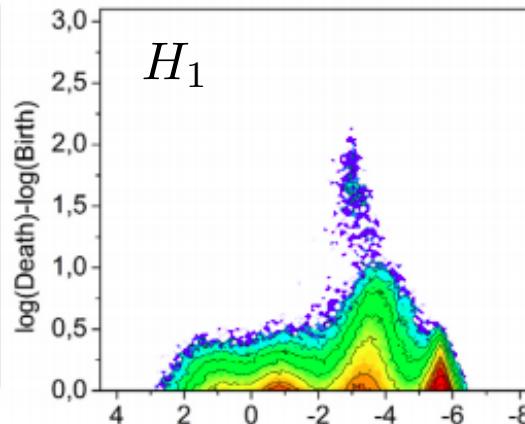
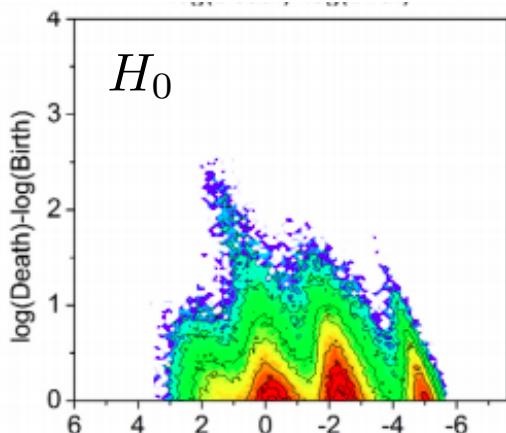
of dimension 2



of dimension 3

P. Pranav, H. Edelsbrunner, R. de Weygaert, G. Vegter, M. Kerber, B. Jones and M. Wintraecken, The topology of the cosmic web in terms of persistent Betti numbers, 2016

<https://arxiv.org/pdf/1608.04519.pdf>



Average persistence
diagrams (log-scale)
for a Voronoï
evolution model

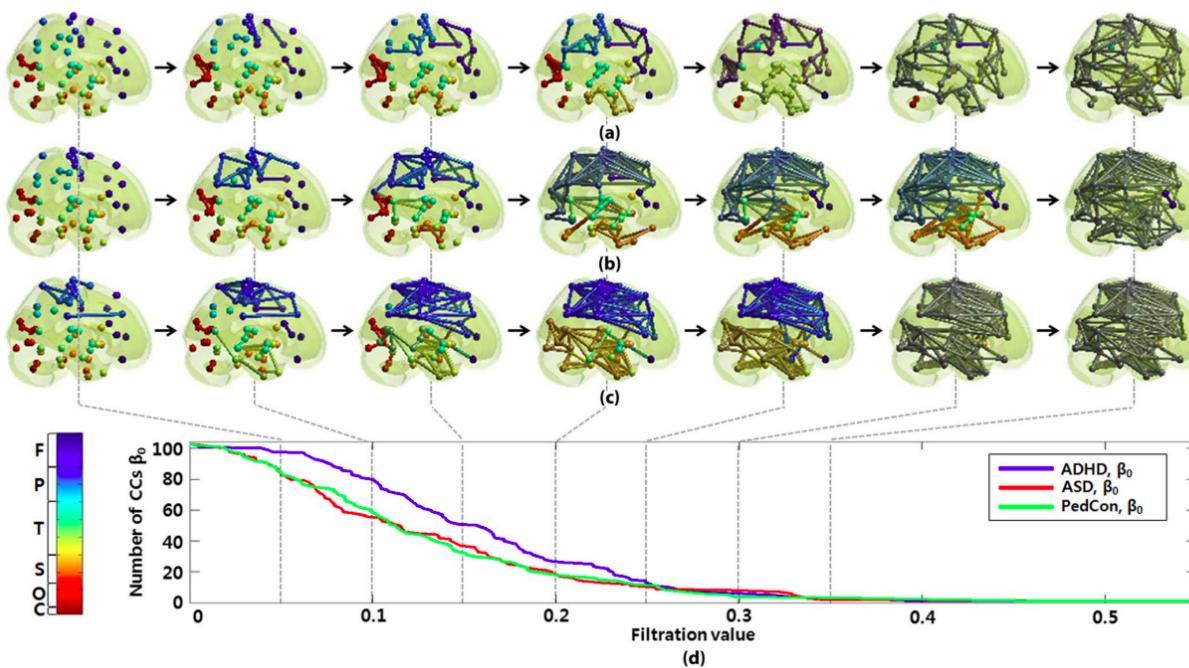
Multiscale analysis II

34/46

Hyekyoung Lee, Hyejin Kang, Moo K Chung, Bung-Nyun Kim, Dong Soo Lee,
Persistent brain network homology from the perspective of dendrogram, 2012

<http://pages.stat.wisc.edu/~mchung/papers/lee.2012.TMI.pdf>

→ H_0 -persistent homology induces a hierarchical clustering



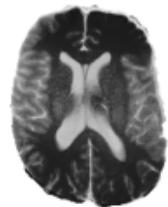
In collaboration with François.

Glioblastoma is the most common brain tumor, diffuse, whose medical diagnostic is difficult to establish.

In this context, the problem of **segmentation** consists in labelling the three regions that form the tumor (edema, necrotic core and enhancing tumor).

We can use **cubic persistent homology**, defined for images.

original image



$t = 0.2$



$t = 0.4$



$t = 0.5$



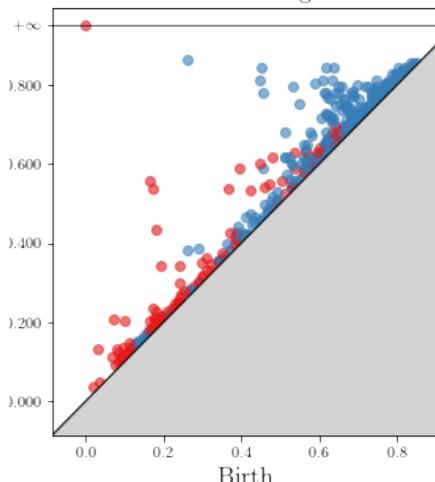
$t = 0.6$



$t = 0.7$



Persistence diagram



I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

3 - Variations on persistent homology

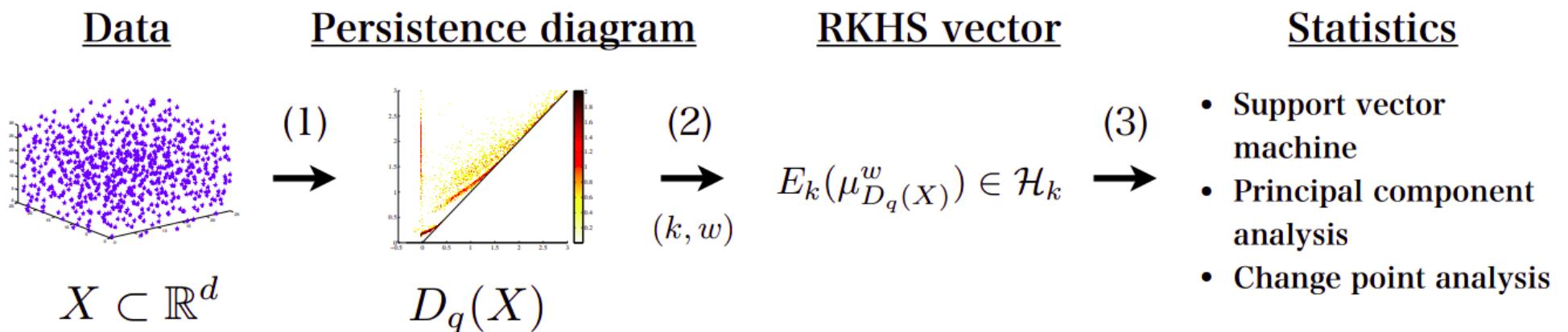
Mathieu Carrière, Marco Cuturi, Steve Oudot, Sliced Wasserstein Kernel for Persistence Diagrams, 2017

<https://arxiv.org/abs/1706.03358>

Genki Kusano, Kenji Fukumizu, Yasuaki Hiraoka, Kernel Method for Persistence Diagrams via Kernel Embedding and Weight Factor, 2018

<https://www.jmlr.org/papers/volume18/17-317/17-317.pdf>

- Barcodes are not subsets of some Euclidean space, hence usual machine learning methods cannot be used directly



Topological layer in Neural Networks

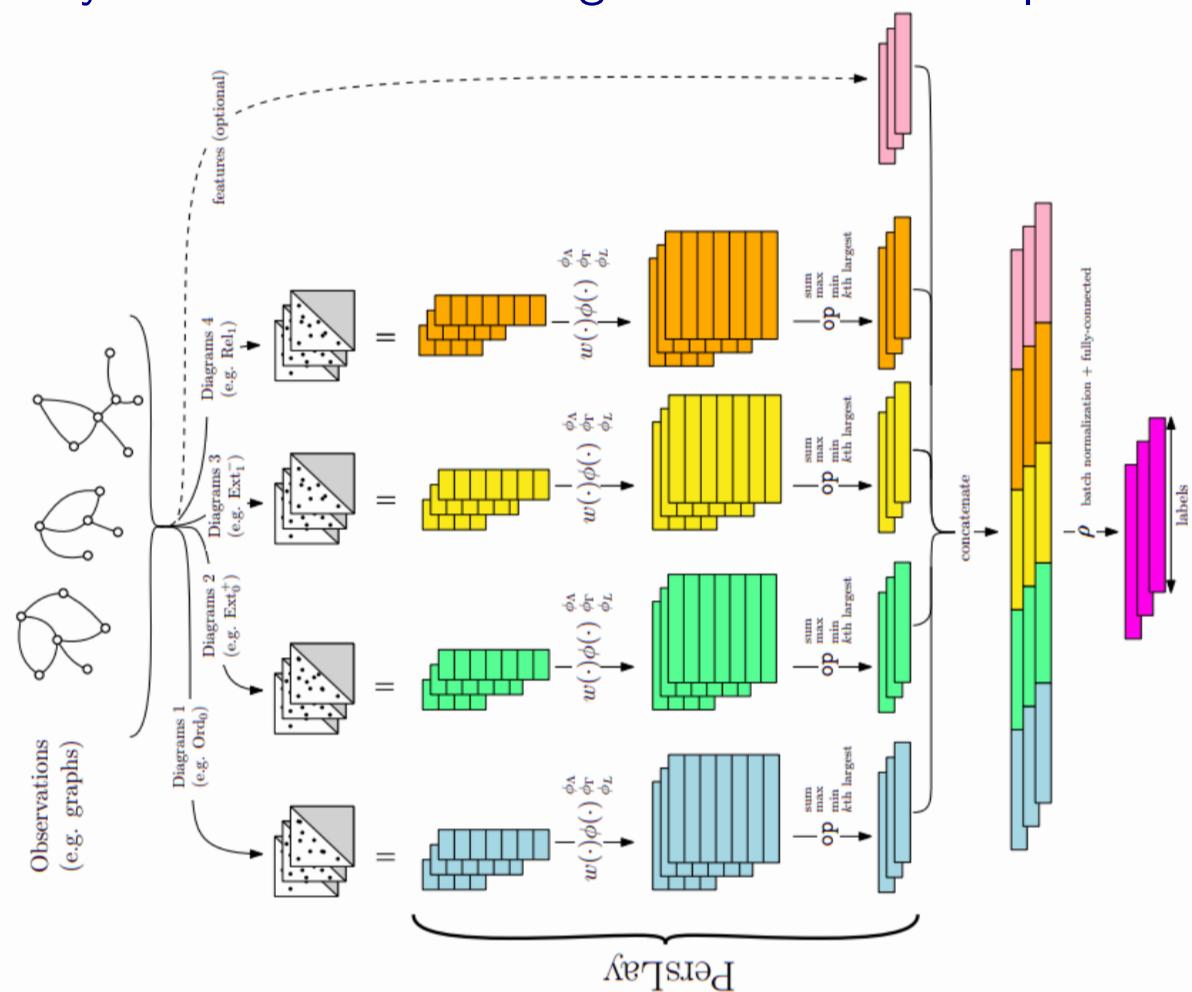
38/46

Rickard Brüel-Gabrielsson, Bradley J. Nelson, Anjan Dwaraknath, Primoz Skraba, Leonidas J. Guibas, Gunnar Carlsson, [A Topology Layer for Machine Learning](#), 2019

<https://arxiv.org/abs/1905.12200>

Mathieu Carrière, Frédéric Chazal, Yuichi Ike, Théo Lacombe, Martin Royer, Yuhei Umeda, [PersLay: A Neural Network Layer for Persistence Diagrams and New Graph Topological Signatures](#), 2019

<https://arxiv.org/abs/1904.09378>



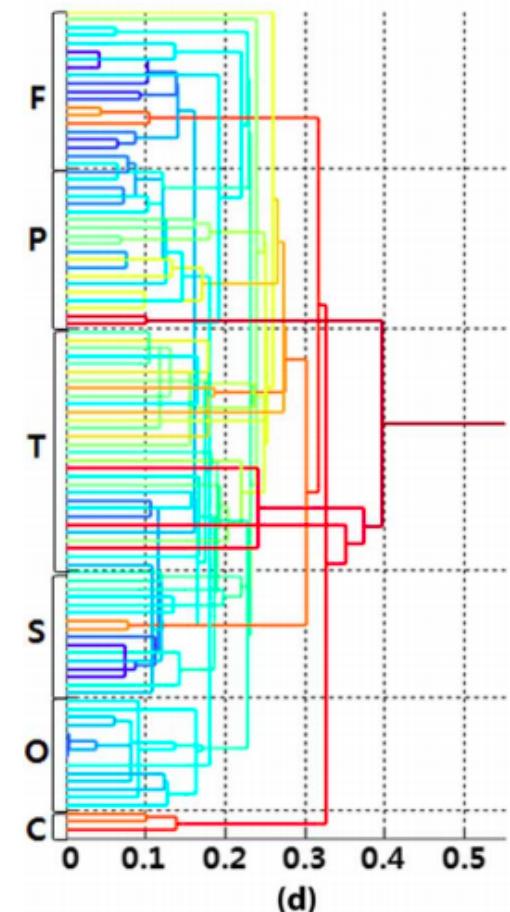
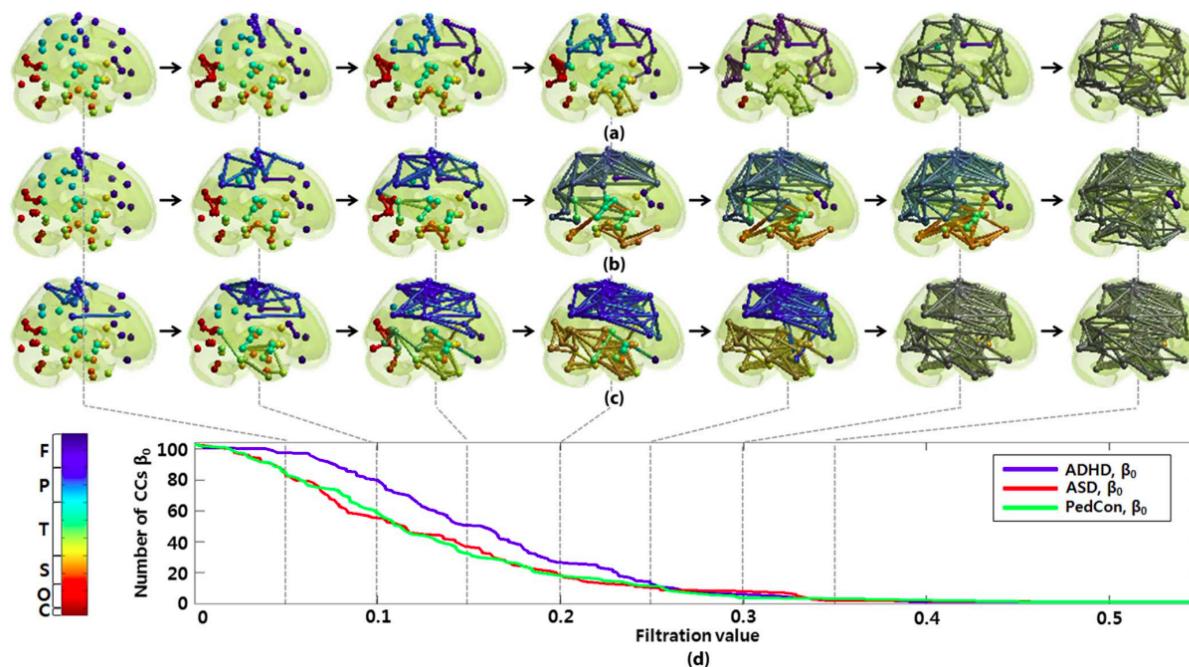
Hierarchical clustering

39/46

Hyekyoung Lee, Hyejin Kang, Moo K Chung, Bung-Nyun Kim, Dong Soo Lee,
Persistent brain network homology from the perspective of dendrogram, 2012

<http://pages.stat.wisc.edu/~mchung/papers/lee.2012.TMI.pdf>

→ H_0 -persistent homology induces a hierarchical clustering



Classification

40/46

Frédéric Chazal, Steve Oudot, Primoz Skraba, Leonidas J. Guibas, Persistence-Based Clustering in Riemannian Manifolds, 2011

<https://geometrica.saclay.inria.fr/team/Fred.Chazal/papers/cgos-pbc-09/cgos-pbcrm-11.pdf>

Chunyuan Li, Maks Ovsjanikov, Frederic Chazal, Persistence-based Structural Recognition, 2014

<https://geometrica.saclay.inria.fr/team/Fred.Chazal/papers/loc-pbsr-14/CVPR2014.pdf>

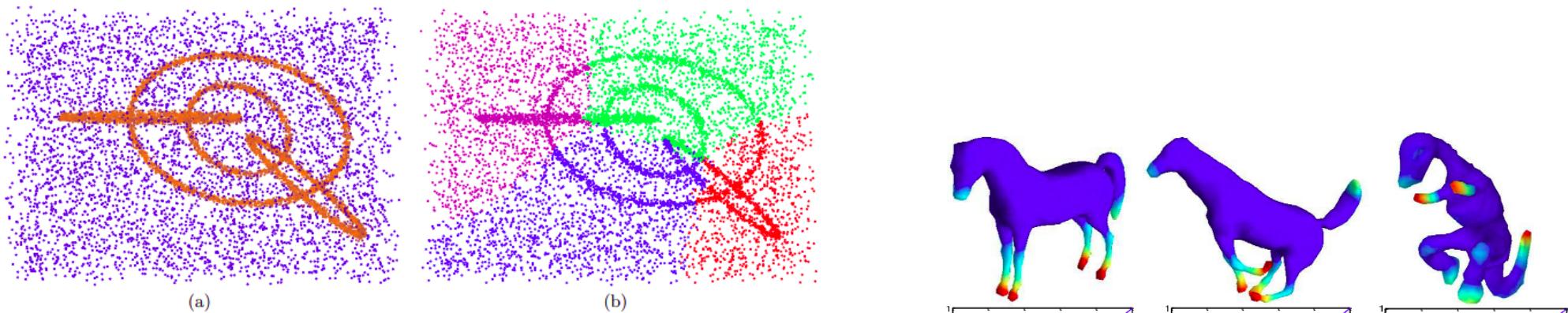
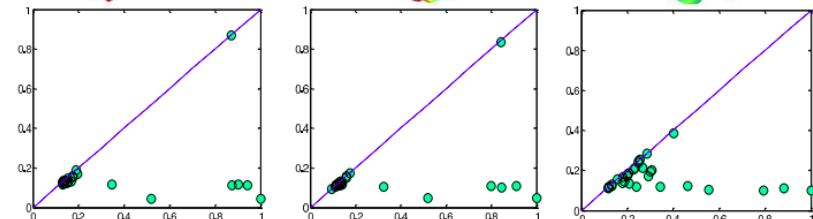
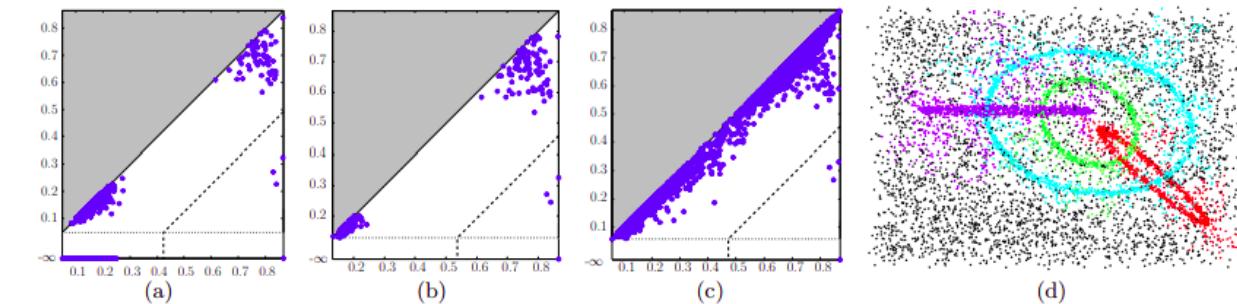


Figure 7: (a) The rings data set with the estimated density function. (b) The result obtained using spectral clustering.

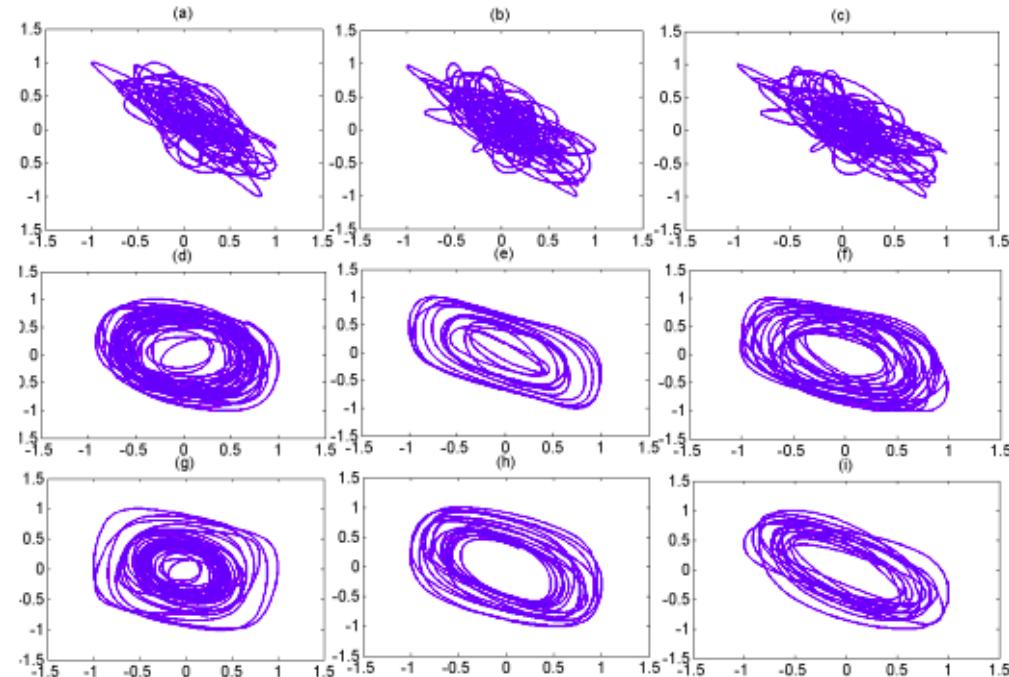
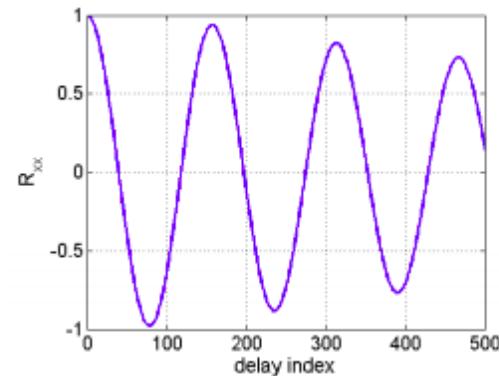
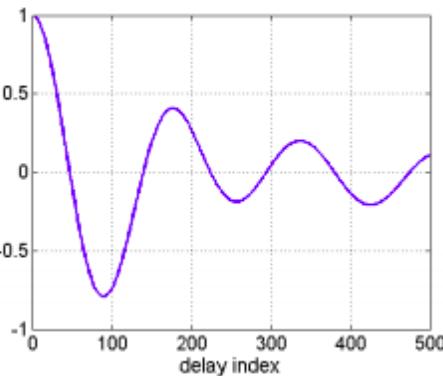


Saba Emrani, Thanos Gentimis, Hamid Krim [Persistent Homology of Delay Embeddings and its Application to Wheeze Detection](#), 2014

https://www.researchgate.net/publication/260523931_Persistent_Homology_of_Delay_EMBEDDINGS_and_its_Application_to_Wheeze_Detection

→ a time series (x_1, x_2, x_3, \dots) does not contain topology...
 turn it into a point cloud of \mathbb{R}^n via **time delay embedding**!

$$X = \{\bar{x}_1, \bar{x}_2, \bar{x}_3, \dots\} \subset \mathbb{R}^n \text{ where } \bar{x}_k = (x_k, x_{k+1}, \dots, x_{k+n-1})$$



I - Decomposition of persistence modules

1 - Simplicial filtrations

2 - Persistence modules

3 - Barcodes

II - Stability of persistence modules

1 - Bottleneck distance

2 - Interleaving distance

3 - Stability

III - Persistent homology in practice

1 - Data analysis

2 - Machine learning

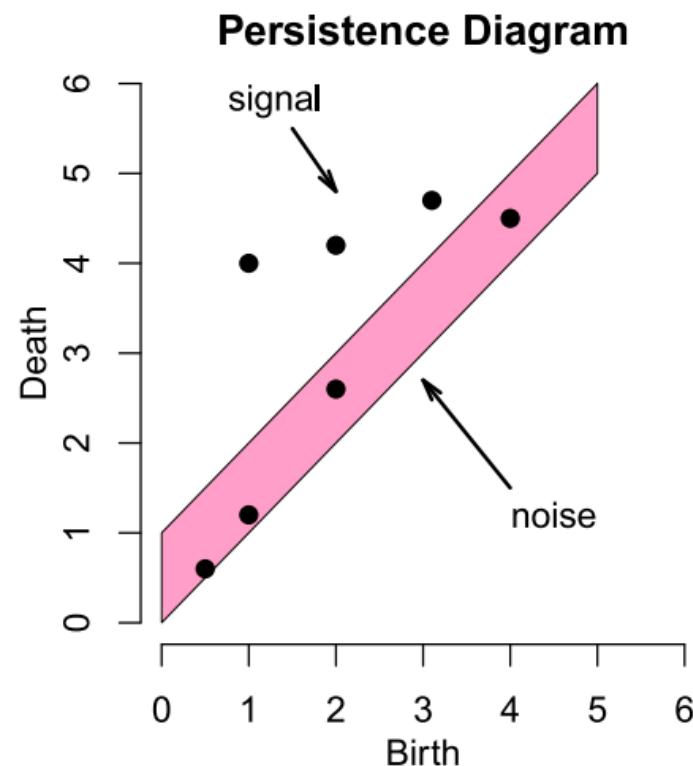
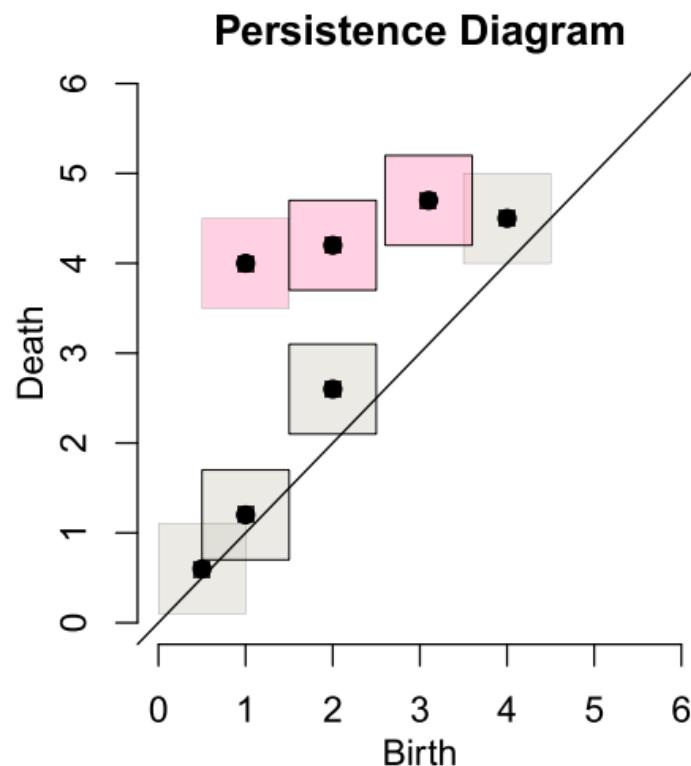
3 - Variations on persistent homology

Statistical aspects of persistent homology 43/46

Brittany Terese Fasy, Fabrizio Lecci, Alessandro Rinaldo, Larry Wasserman, Sivaraman Balakrishnan and Aarti Singh, [Confidence sets for persistence diagrams, 2014](#)

<https://arxiv.org/pdf/1303.7117.pdf>

- Given a barcode, how to determine statistically what is noise and what is not?



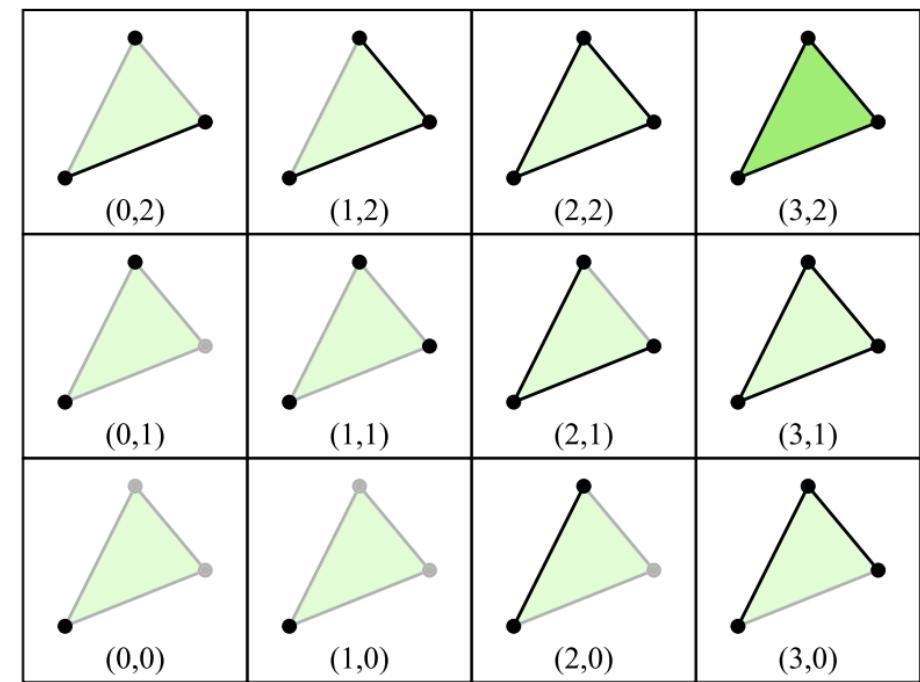
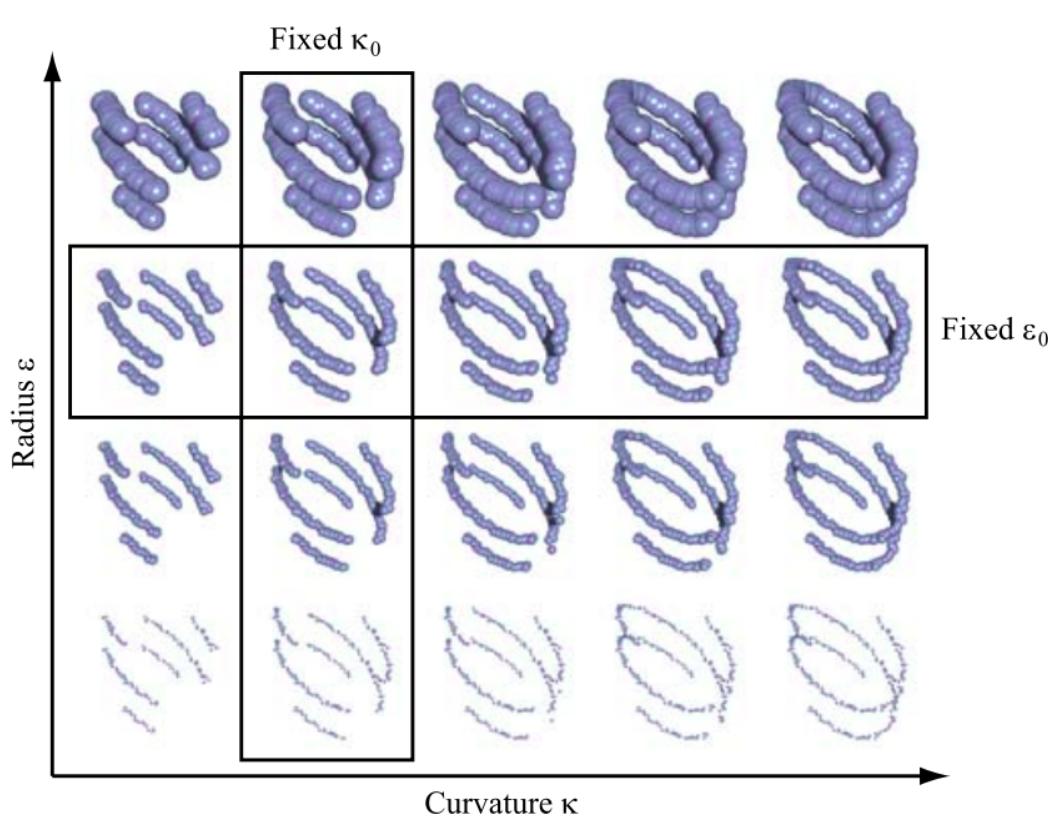
Higher-dimensional persistence

44/46

Gunnar Carlsson, Afra Zomorodian, The Theory of Multidimensional Persistence, 2009

<https://link.springer.com/article/10.1007/s00454-009-9176-0>

→ What if our filtration is not indexed only by $t \in \mathbb{R}^+$?



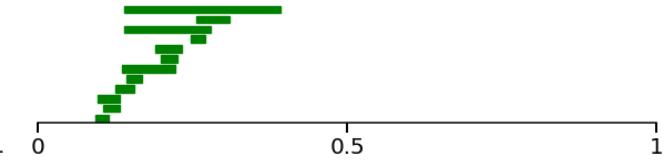
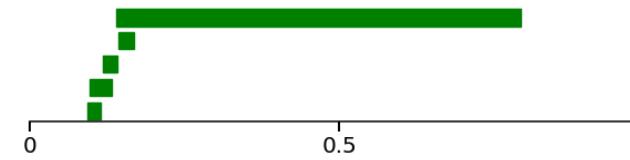
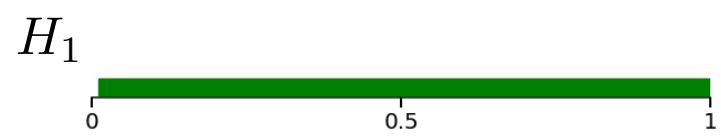
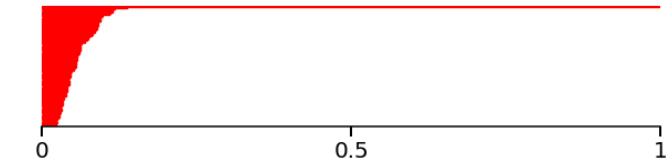
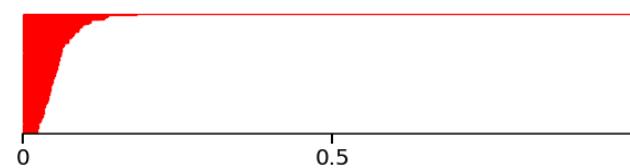
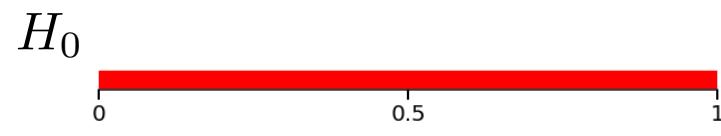
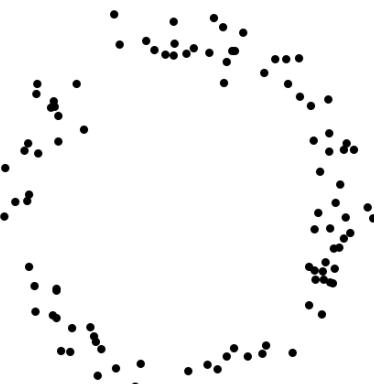
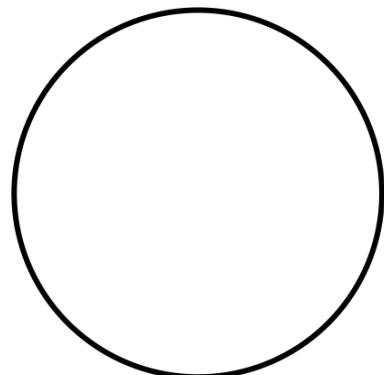
Wasserstein stability

45/46 (1/2)

Hirokazu Anai, Frédéric Chazal, Marc Glisse, Yuichi Ike, Hiroya Inakoshi, Raphaël T.,
Yuhei Umeda, [DTM-based filtrations](#), 2020

<https://arxiv.org/abs/1811.04757>

→ When our dataset is not close to an underlying object in **Hausdorff distance**

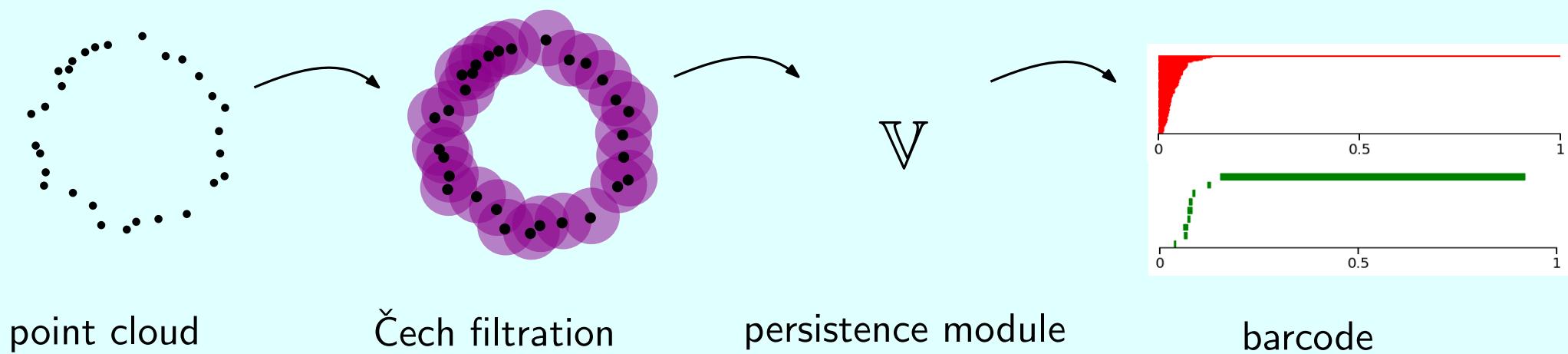


Wasserstein stability

45/46 (2/2)

Conclusão

Persistent homology allows for a **multi-scale** and **stable** estimation of the homology of the datasets.



Allows to analyze data from a new perspective.

A course about TDA: <https://raphaeltinarrage.github.io/EMAp.html>