

Topological Data Analysis and Clustering Algorithms in Machine Learning

İsmail GÜZEL

Mathematical Engineering - İTÜ

March 13, 2022

Outline

- 1 Introduction
- 2 Hierarchical Clustering
- 3 Topological Data Analysis
- 4 Theoretical Contributions
- 5 Experimental Contributions

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Hierarchical Clustering and Zeroth Persistent Homology

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ORIGINAL PAPER

A new non-archimedean metric on persistent homology

Check for updates

İsmail Güzel¹  · Atabey Kaygun¹ 

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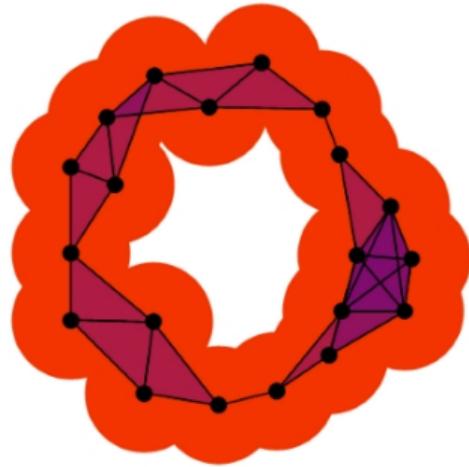
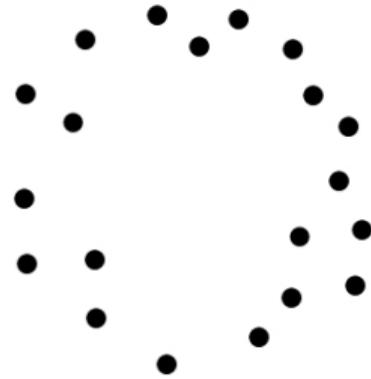
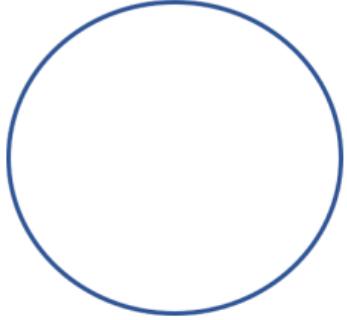
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PERSISTENT HOMOLOGY, MATROIDS AND COBORDISMS

İSMAIL GÜZEL AND ATABEY KAYGUN



"Data has shape, and shape has meaning."

Prof. Gunnar Carlsson

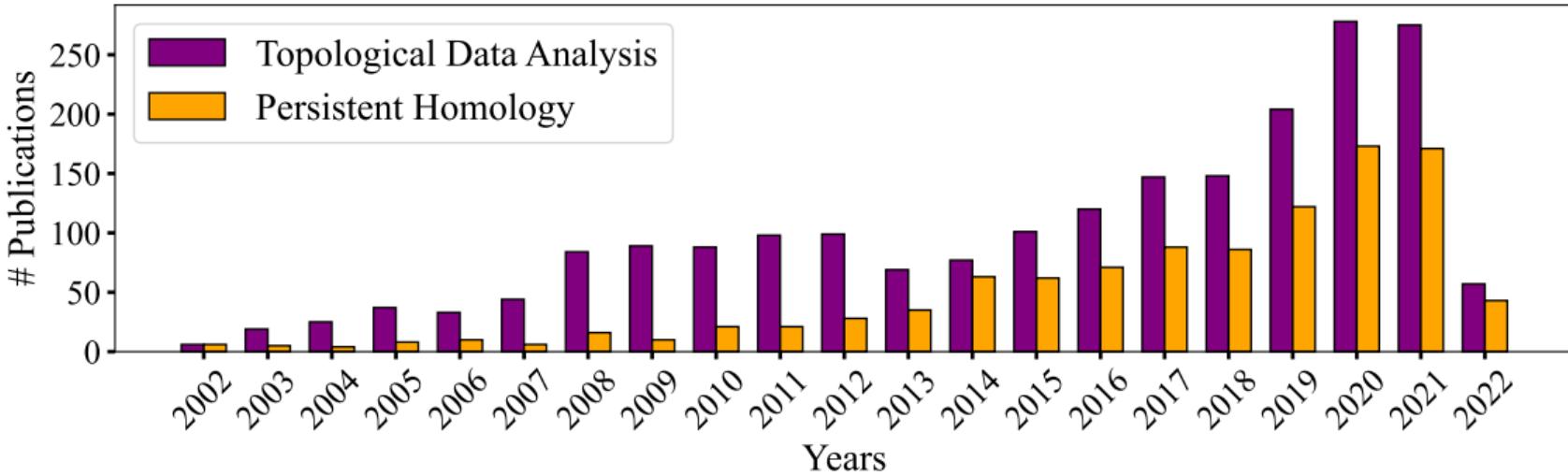
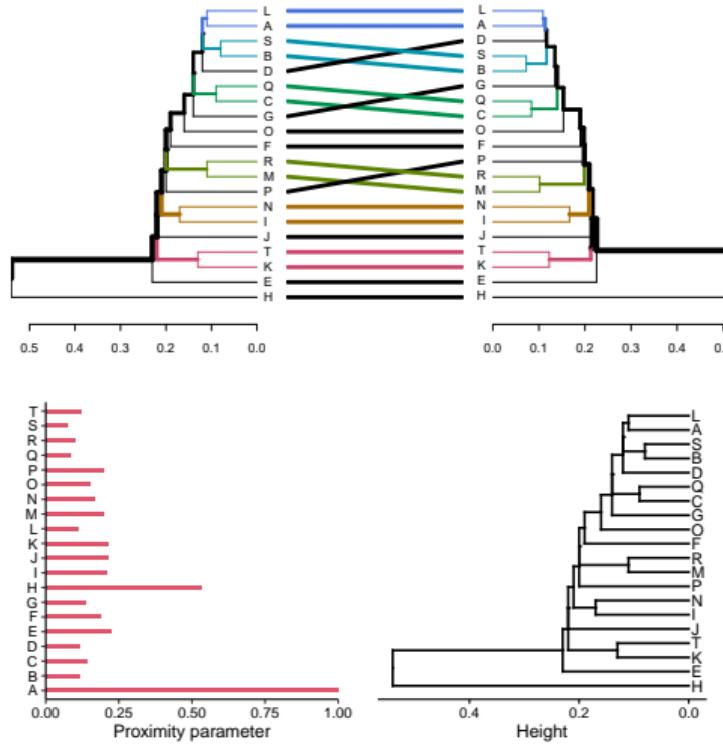


Figure: The data taken from SCOPUS

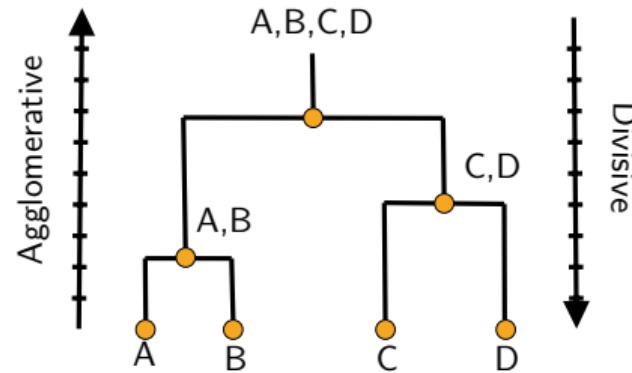
Questions answered

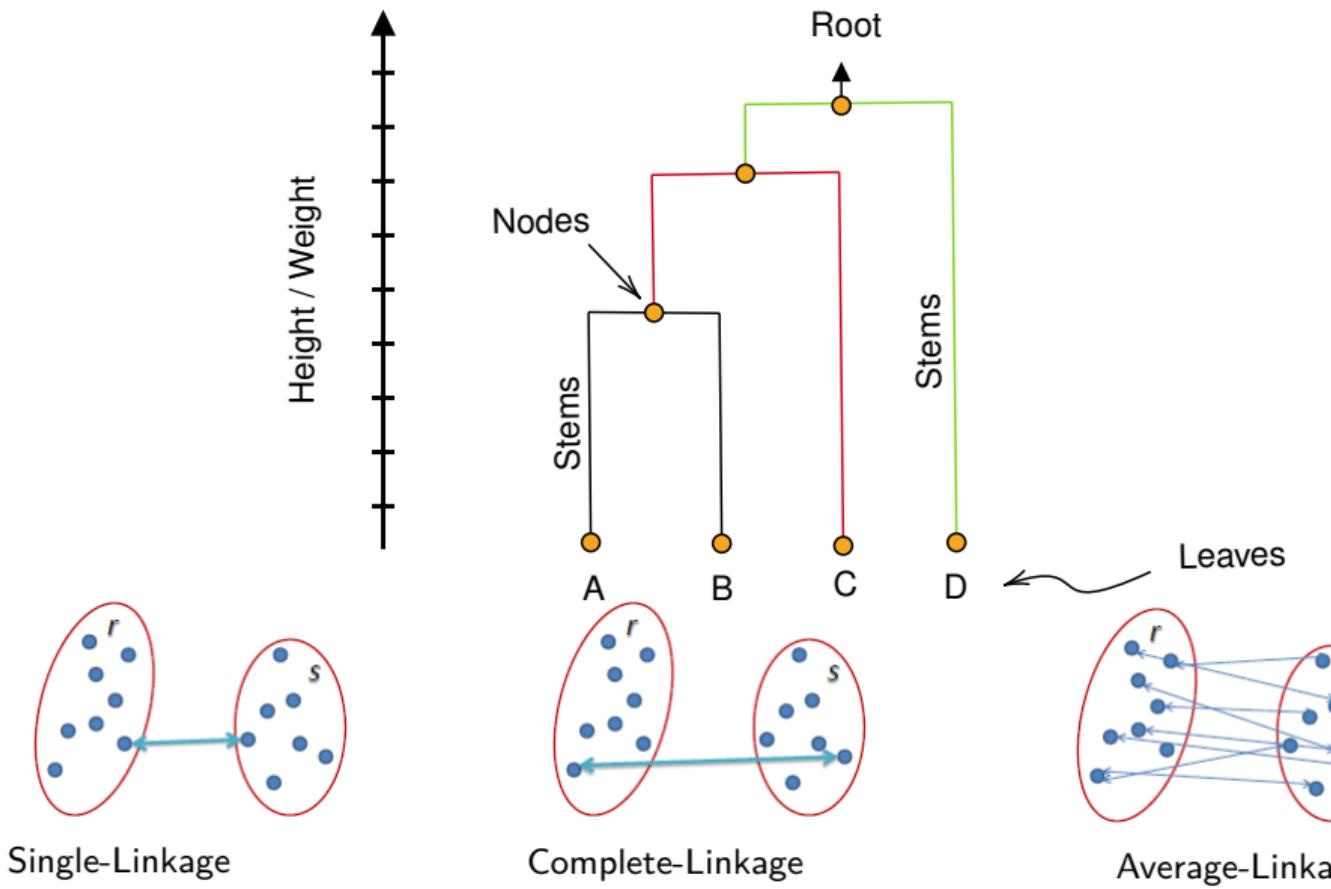
1. What are the similarities and differences between hierarchical clustering and 0-th persistent homology?
2. What is the difference from the persistence barcode?
3. What about higher dimensional persistent homology?
4. What are the experiments on real datasets?



Hierarchical Clustering

- Unsupervised machine learning algorithm
- Aim: divide the data set into disjoint subsets such that
 - homogeneous in cluster
 - heterogeneous between clusters
- The metric structure of ambient space from data set.
- A nice tree-based representation, called a *dendrogram*



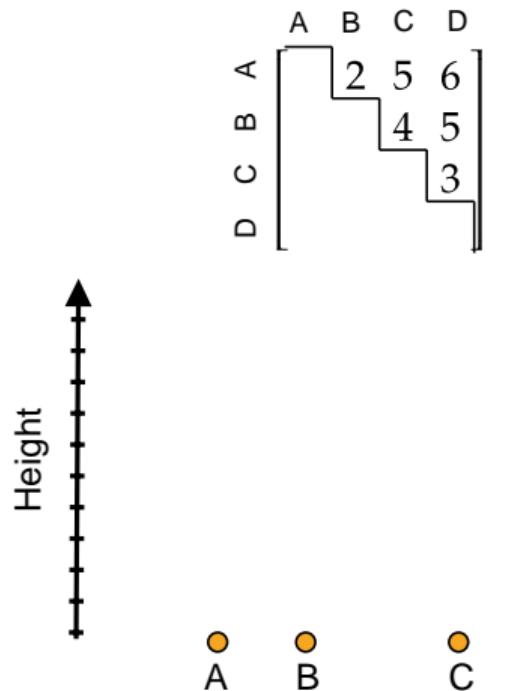


Single-Linkage

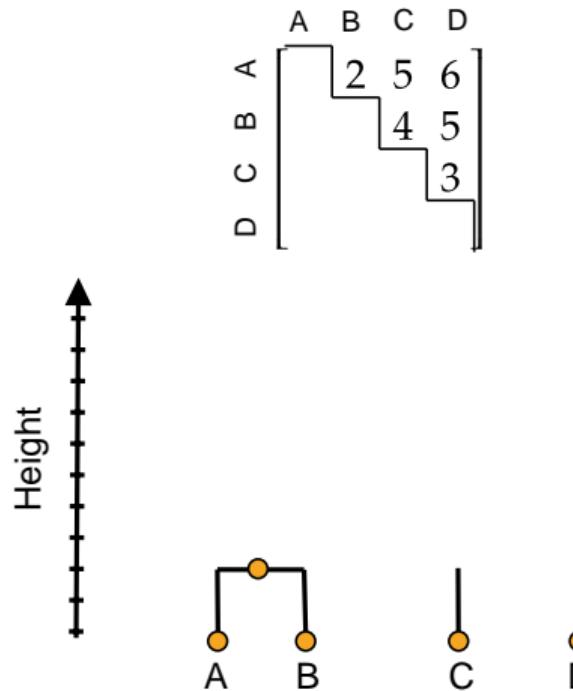
Complete-Linkage

Average-Linkage

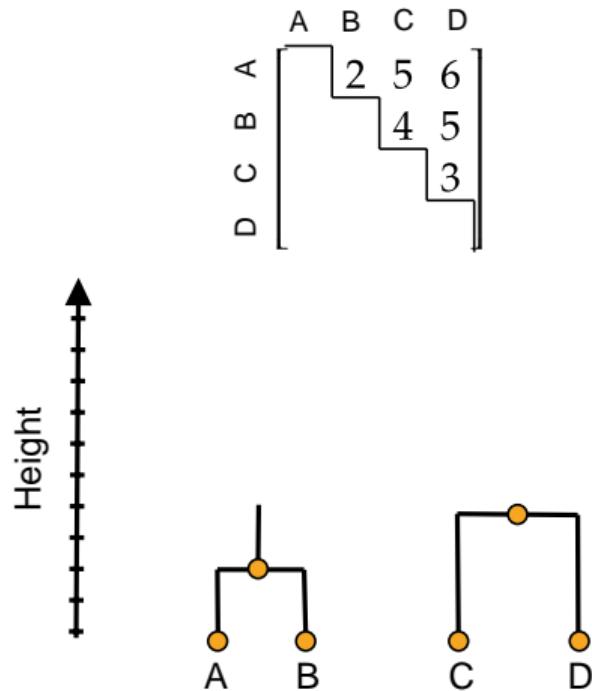
Cophenetic Matrix



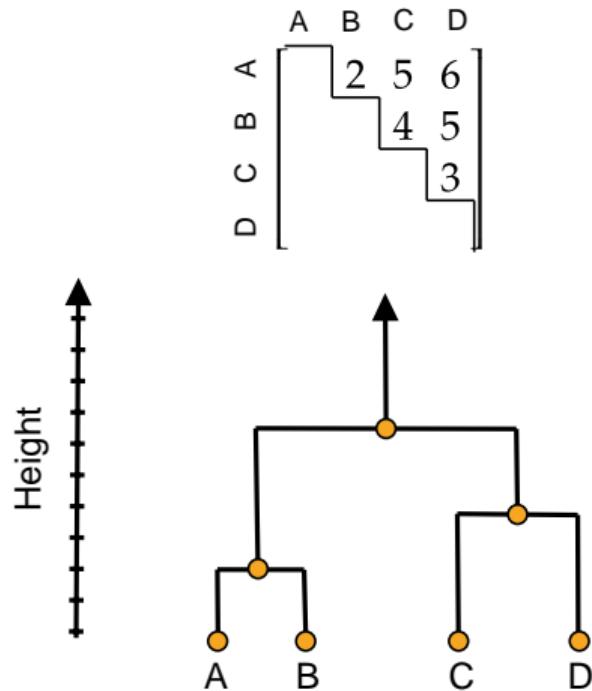
Cophenetic Matrix



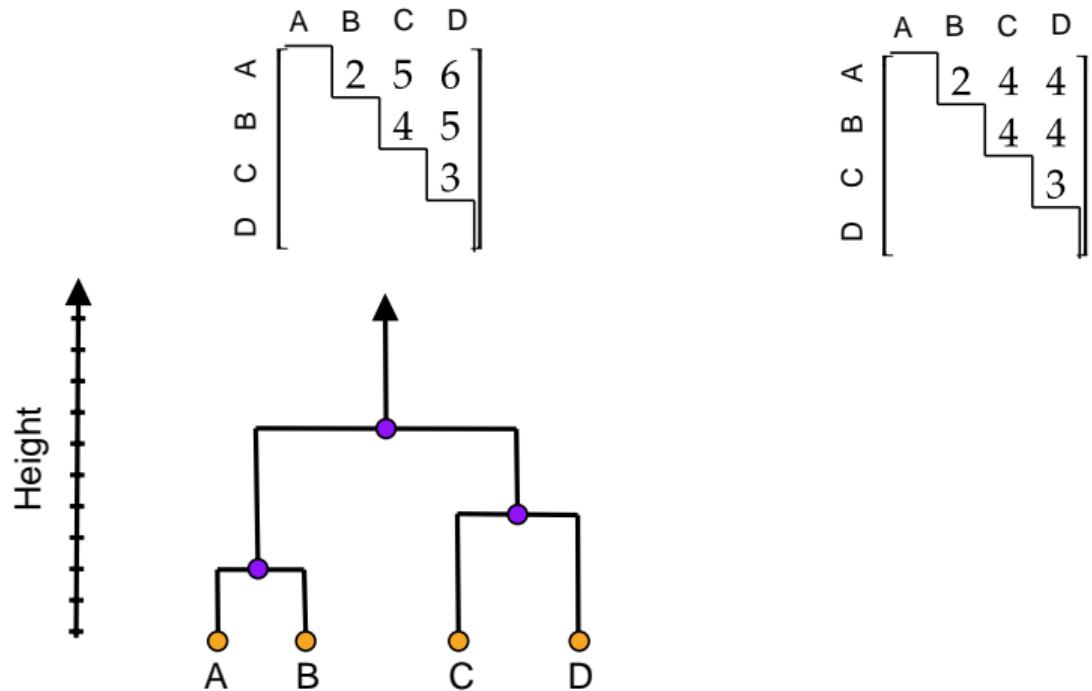
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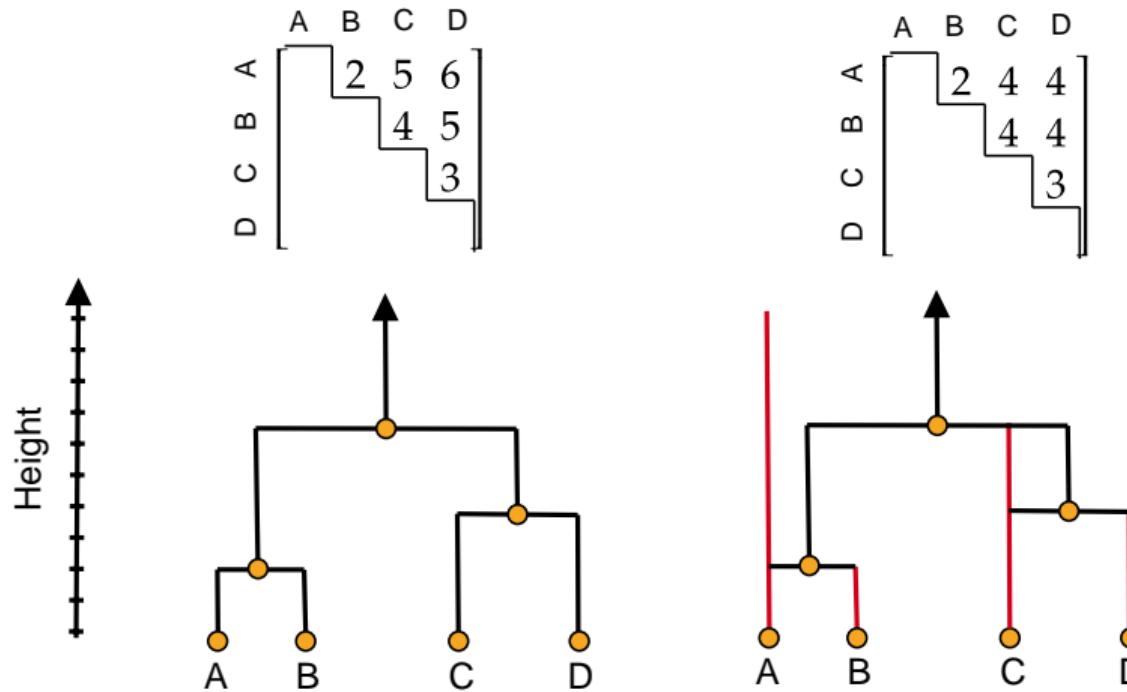
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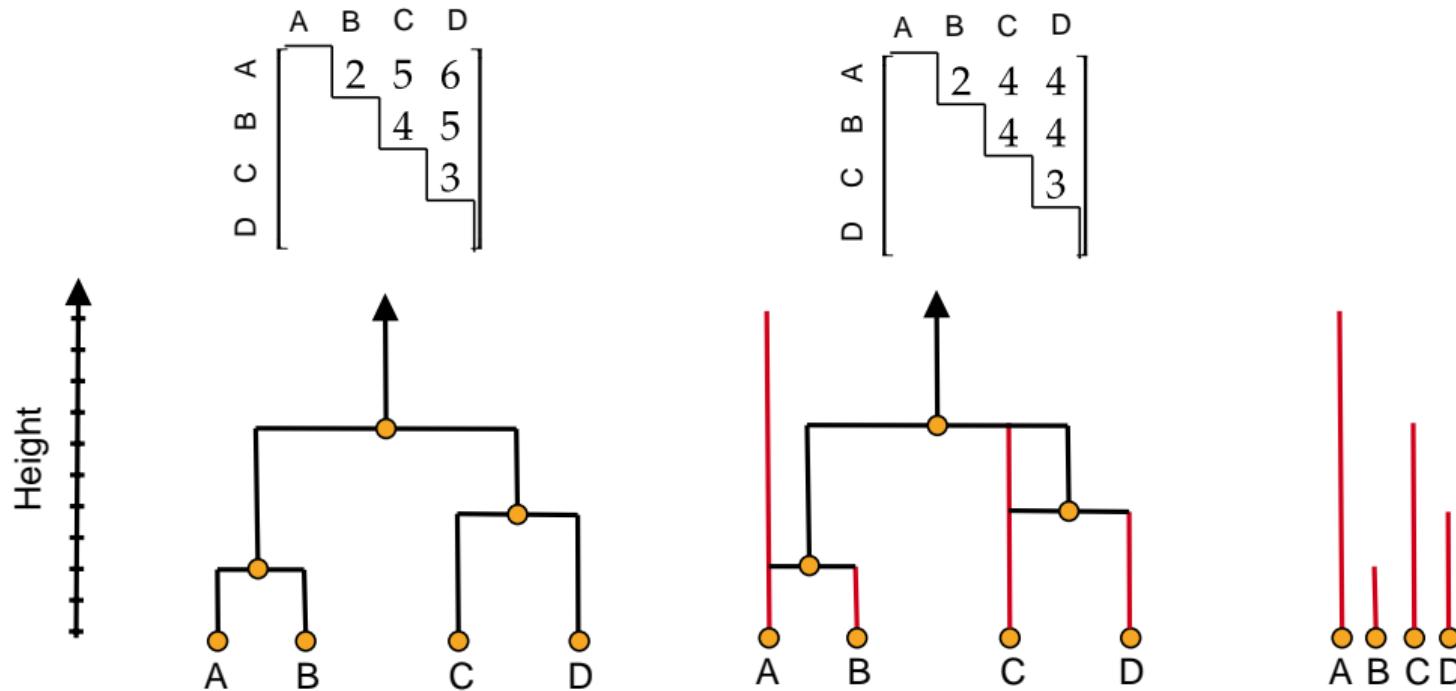
Cophenetic Matrix



Cophenetic Matrix



Cophenetic Matrix



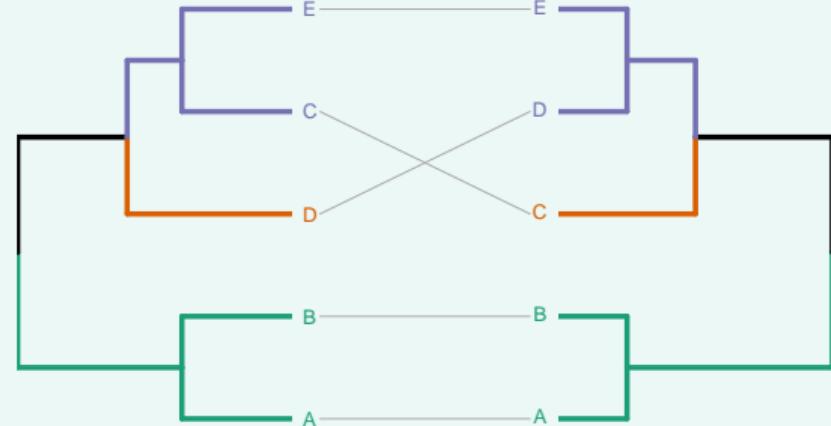
Compare Dendograms

- Two Dendograms
- Two cophenetic matrix

Mantel Test

- Non-parametric Test
- Like correlation coefficient
- Mantel statistic $r \in [-1, 1]$

Tanglegrams



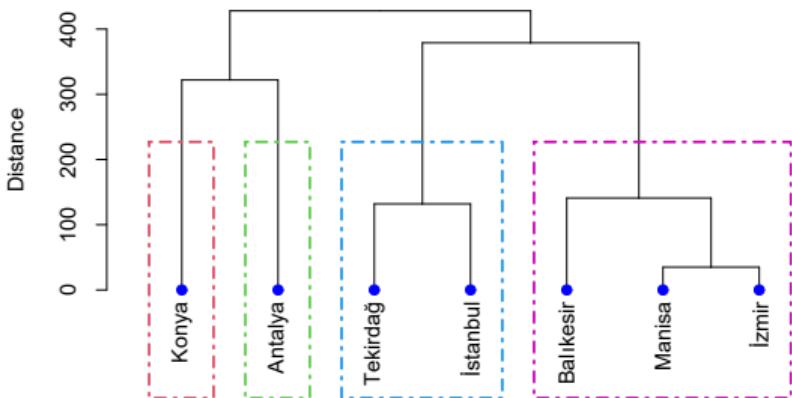
Entanglement values ranges 0 and 1.

Silhouette scores

Silhouette scores

$$s(x) = \frac{b(x) - a(x)}{\max(a(x), b(x))},$$

$$a(x) = d(x, U(x)) \quad \text{and} \quad b(x) = \min_{C \neq U(x)} d(x, C).$$



	Tekirdağ	İstanbul	Balıkesir	Manisa	İzmir	Konya	Antalya
Tekirdağ	0						
İstanbul	132	0					
Balıkesir	379	390	0				
Manisa	511	529	141	0			
İzmir	506	564	176	35	0		
Konya	794	662	551	534	550	0	
Antalya	850	718	505	428	444	322	0

Points	Cohesion	Separation	Silhouette
Tekirdağ	132	465.3	0.72
İstanbul	132	494.5	0.73
Balıkesir	158.5	384.5	0.59
Manisa	88	428	0.79
İzmir	105.5	444	0.76
Konya	0	322	1
Antalya	0	322	1

Overall silhouette score is $s \approx 0.80$

Simplicial Technology

k -simplex

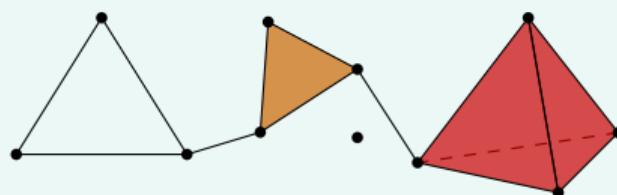
0-simplex
vertex

1-simplex
edge

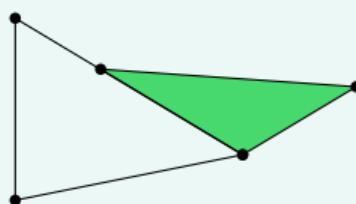
2-simplex
triangle

3-simplex
tetrahedron

Simplicial complex



Simplicial complex



Not simplicial complex

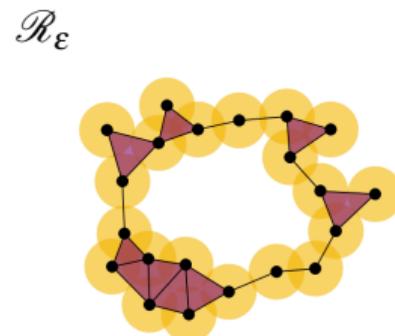
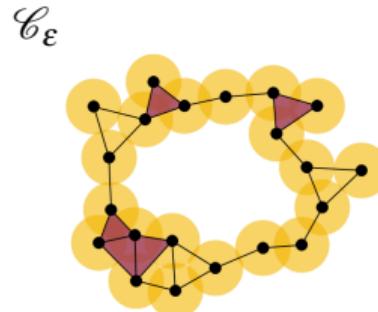
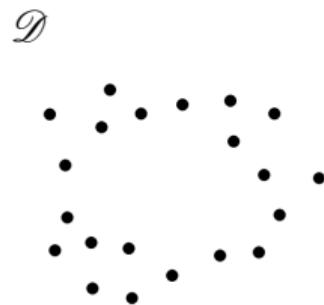
Point Cloud to Complex

Čech Complex

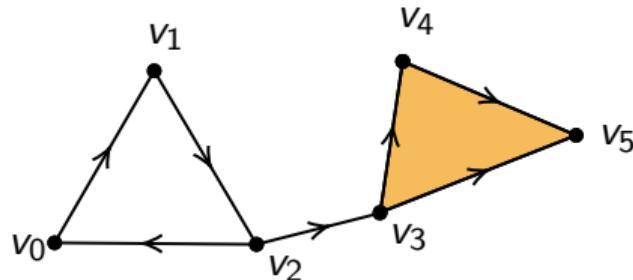
$$\mathcal{C}_\varepsilon = \left\{ \sigma \subseteq \mathcal{D} \quad | \quad \bigcap_{x \in \sigma} B_\varepsilon(x) \neq \emptyset \right\}, \quad B_\varepsilon(x) = \{y \mid d(x, y) < \varepsilon\}$$

Vietoris Rips Complex

$$\mathcal{R}_\varepsilon = \{ \sigma \subset \mathcal{D} \quad | \quad \|x - y\| \leq \varepsilon, \text{ for all } x, y \in \sigma \}$$



Example: A Simplicial complex \mathcal{K}



$$\begin{aligned}\mathcal{C}_0 &= \{v_0, v_1, v_2, v_3, v_4, v_5\}, \\ \mathcal{C}_1 &= \{[v_0, v_1], [v_1, v_2], [v_2, v_0], [v_2, v_3], \\ &\quad [v_3, v_4], [v_4, v_5], [v_3, v_5]\} , \\ \mathcal{C}_2 &= \{[v_3, v_4, v_5]\}.\end{aligned}$$

$$0 \xrightarrow{\partial_3} \mathcal{C}_2 \xrightarrow{\partial_2} \mathcal{C}_1 \xrightarrow{\partial_1} \mathcal{C}_0 \xrightarrow{\partial_0} 0$$

Homology

The k^{th} homology group of \mathcal{K} is defined by

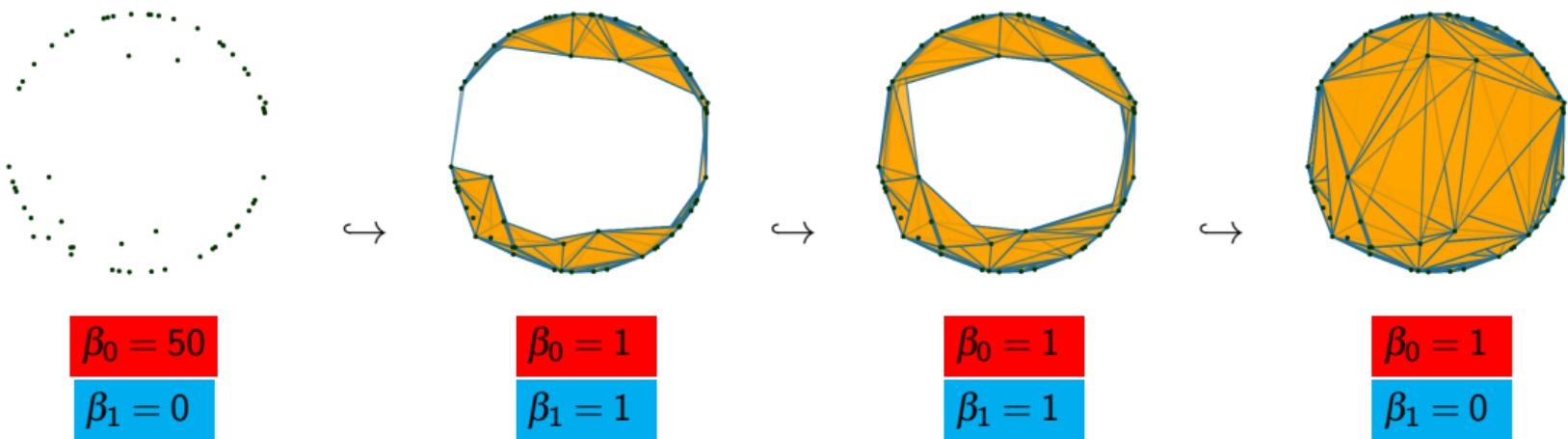
$$H_k(K) := \frac{\ker(\partial_k)}{\text{im}(\partial_{k+1})}$$

$$\beta_0 = \dim(\ker(\partial_0)) - \dim(\text{im}(\partial_1)) = 6 - 5 = 1$$

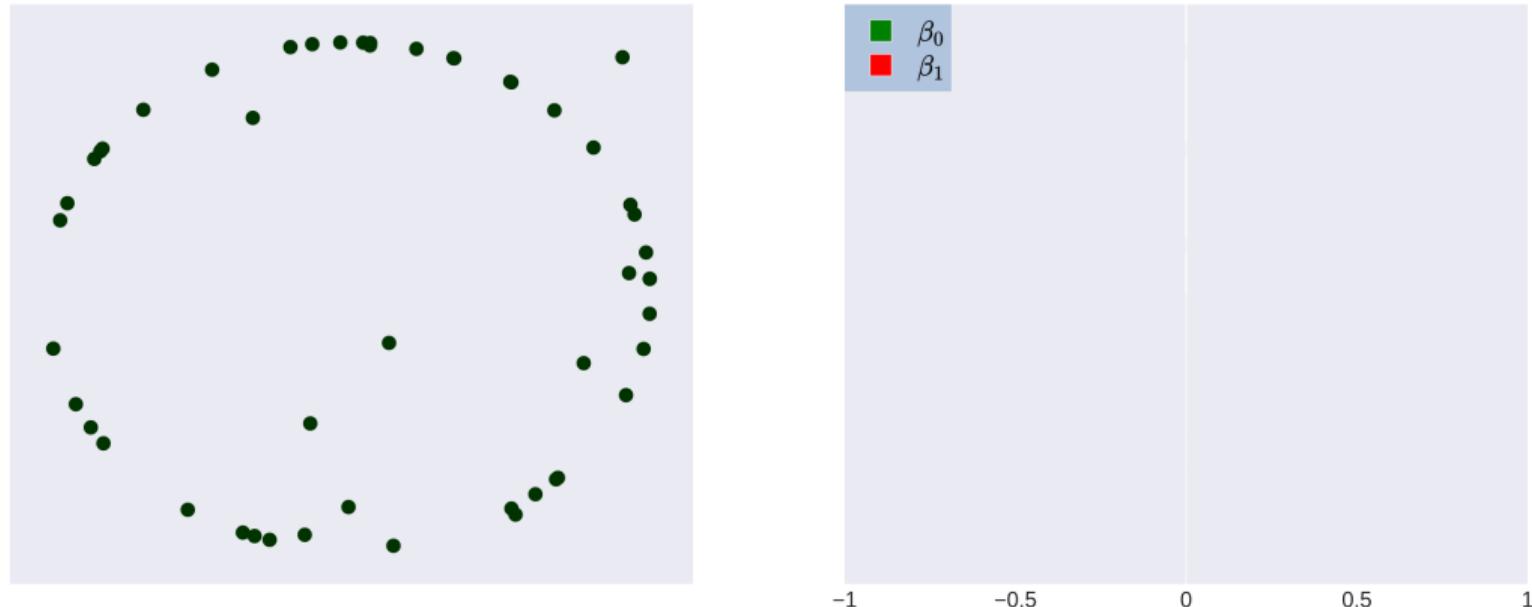
$$\beta_1 = \dim(\ker(\partial_1)) - \dim(\text{im}(\partial_2)) = 2 - 1 = 1$$

Vietoris-Rips Filtration

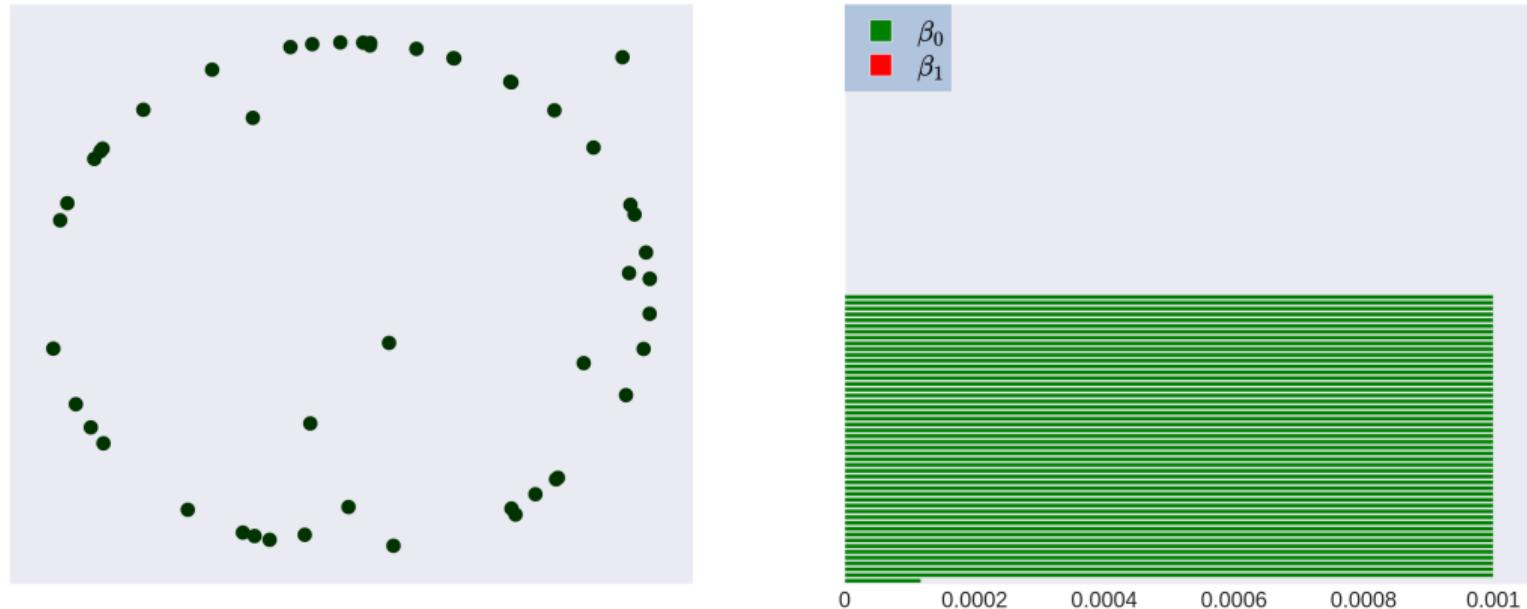
Filtration of a simplicial complex K is a collection of subcomplexes $\mathbb{K} = \{K_\varepsilon : \varepsilon \in \mathbb{R}^+\}$ that satisfy $K_{\varepsilon_1} \subseteq K_{\varepsilon_2}$ whenever $\varepsilon_1 \leq \varepsilon_2$.



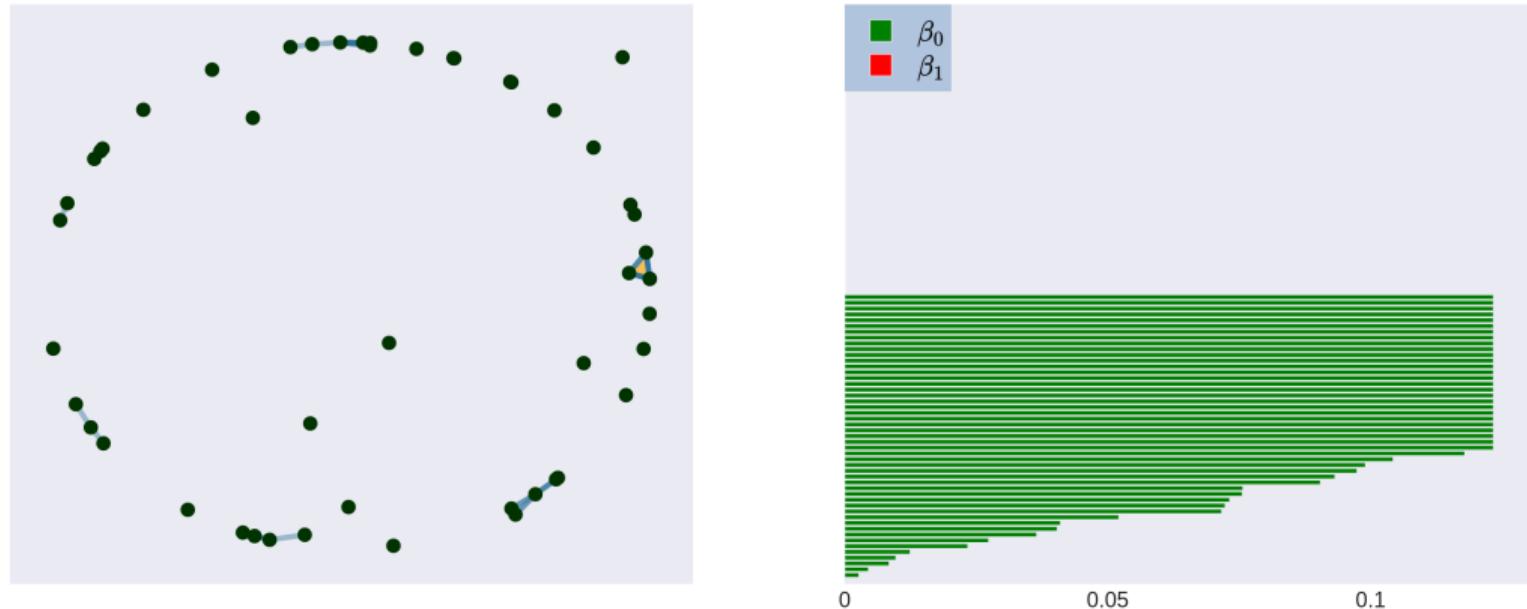
Persistent Homology and Barcodes



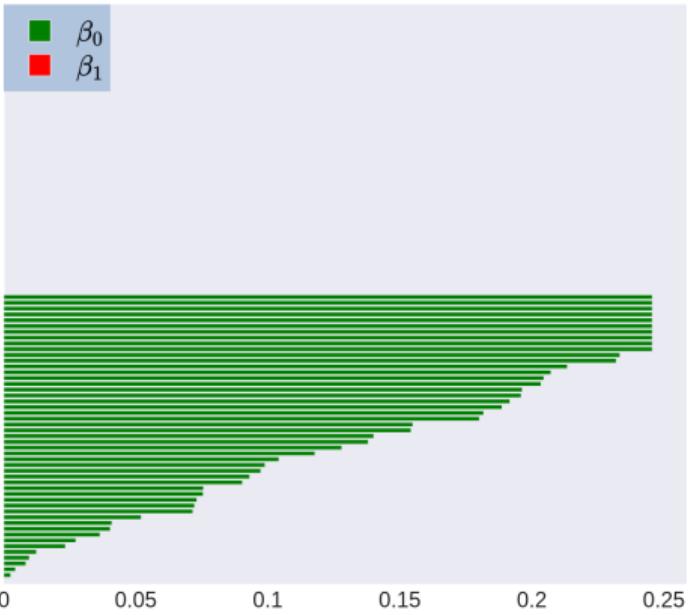
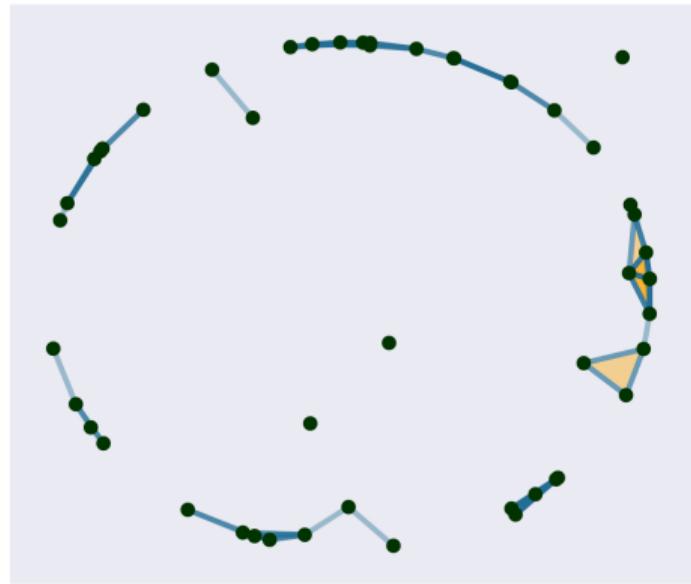
Persistent Homology and Barcodes



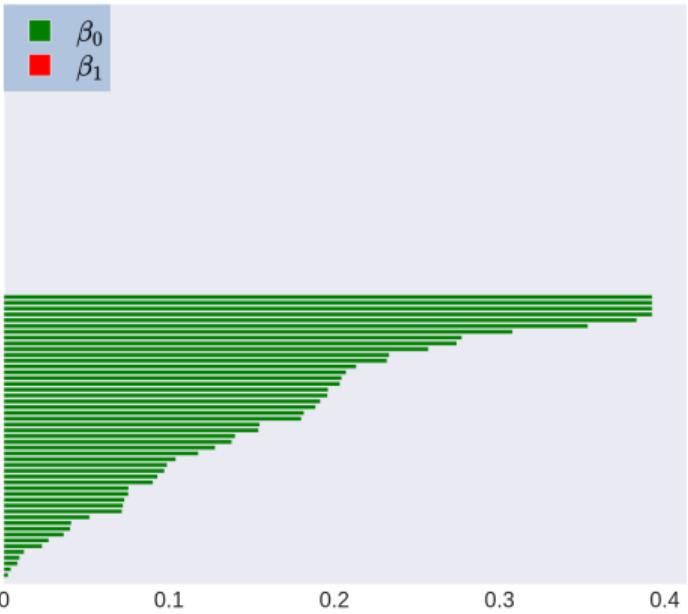
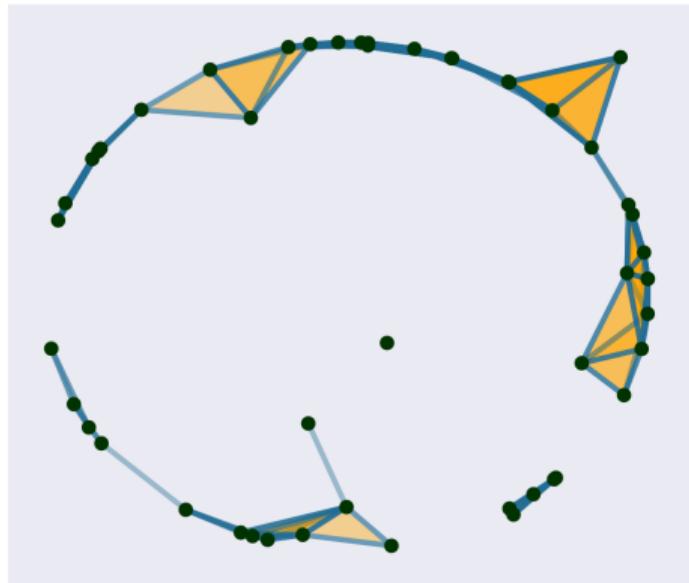
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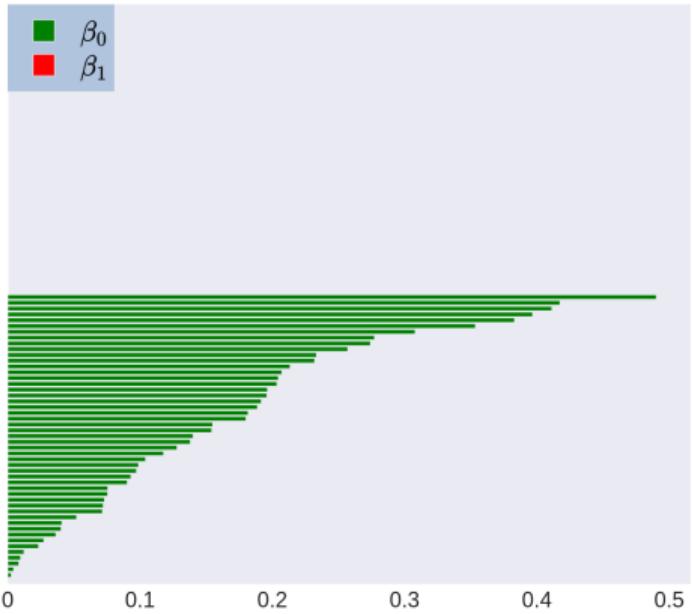
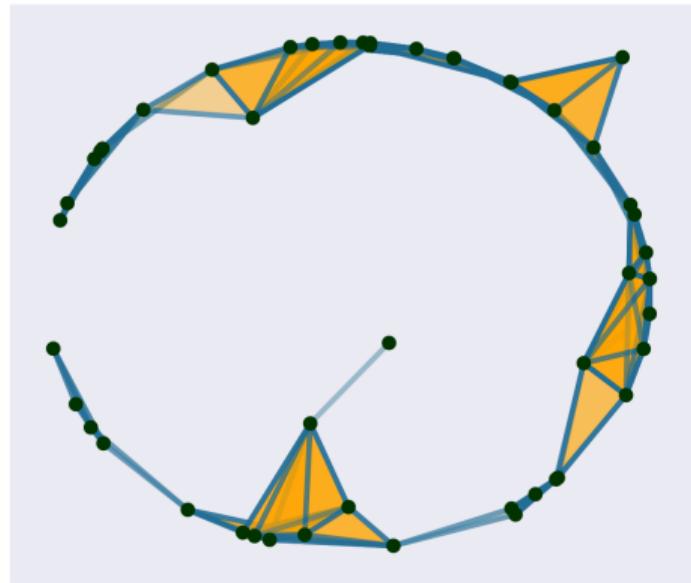
Persistent Homology and Barcodes



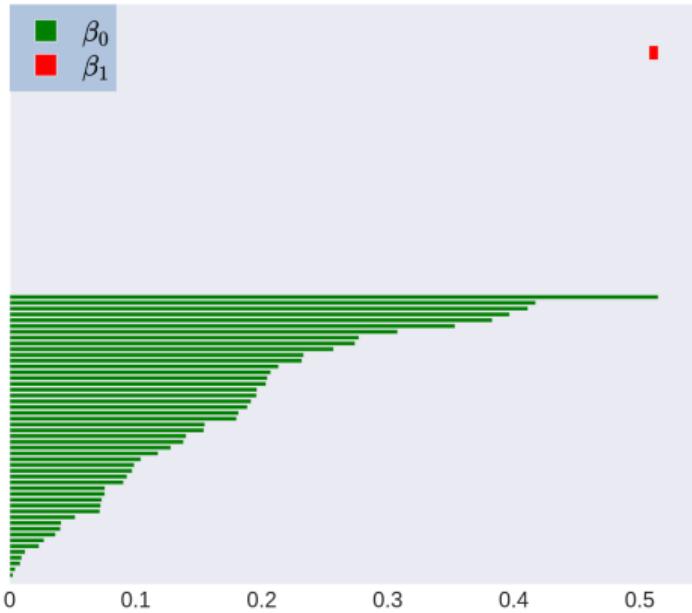
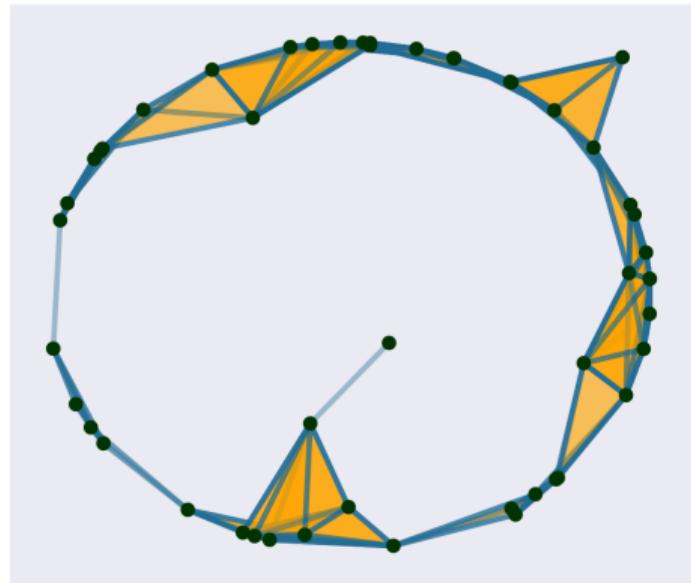
Persistent Homology and Barcodes



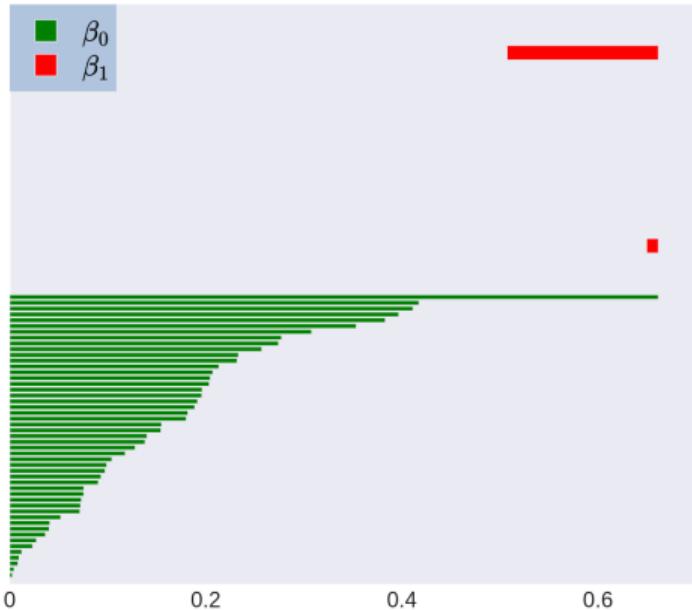
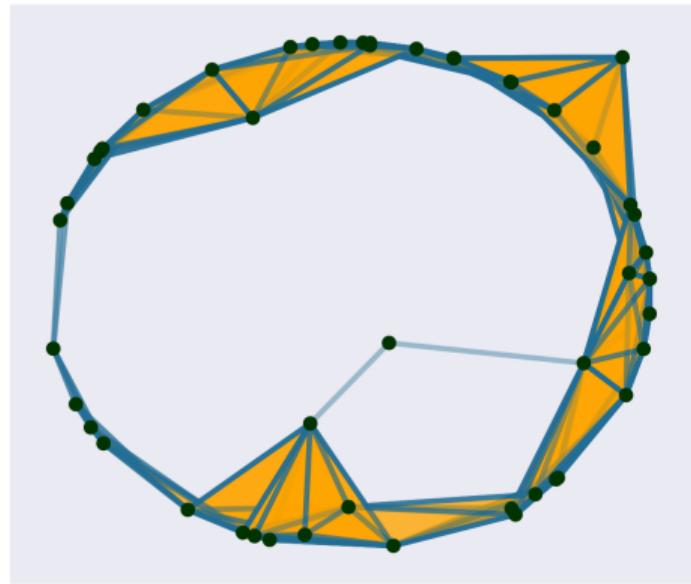
Persistent Homology and Barcodes



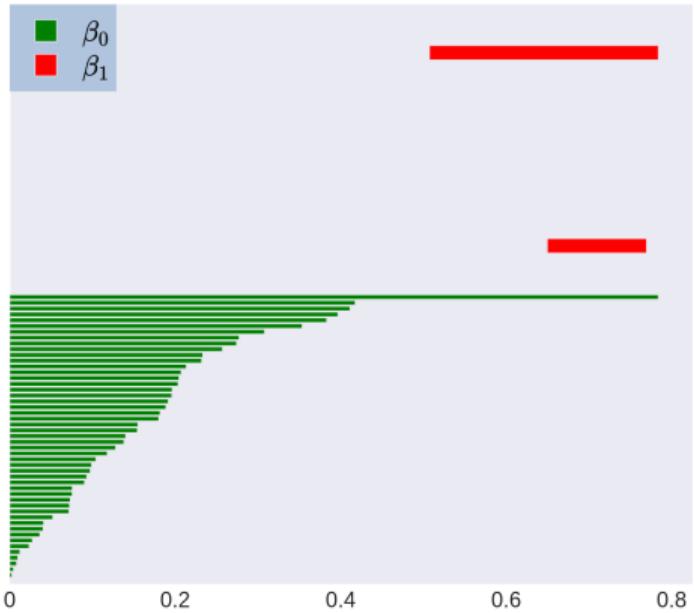
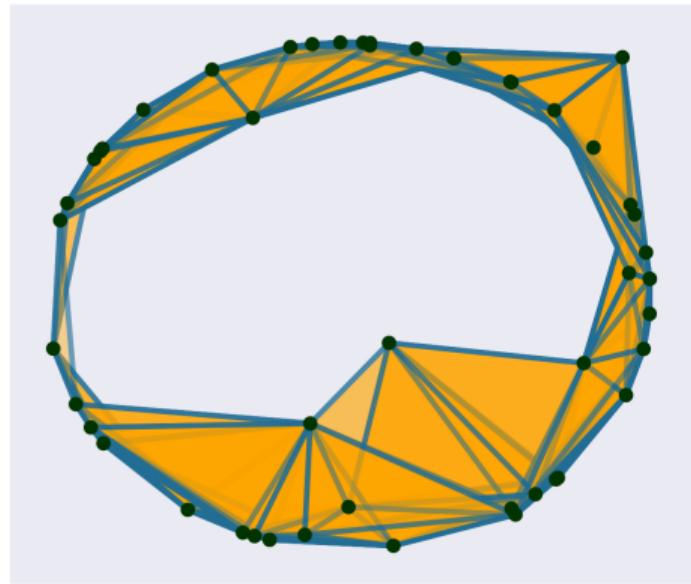
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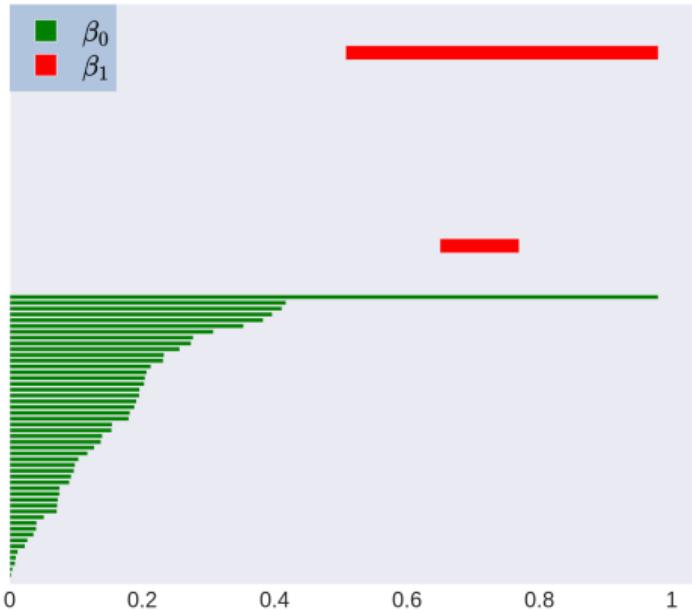
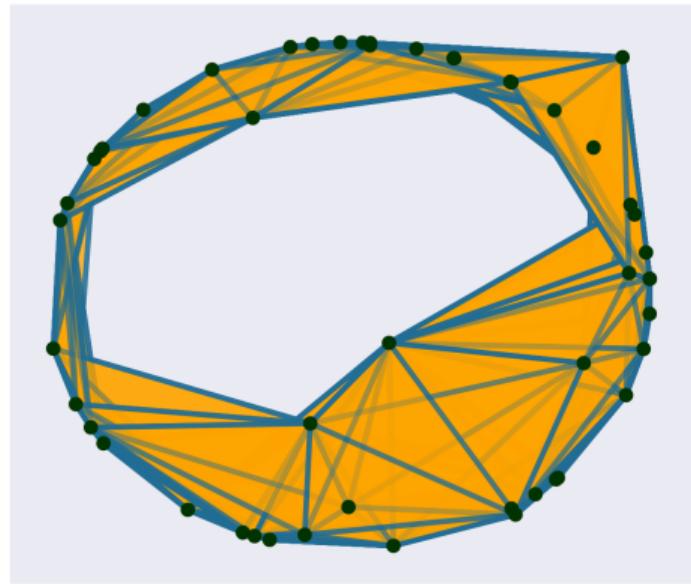
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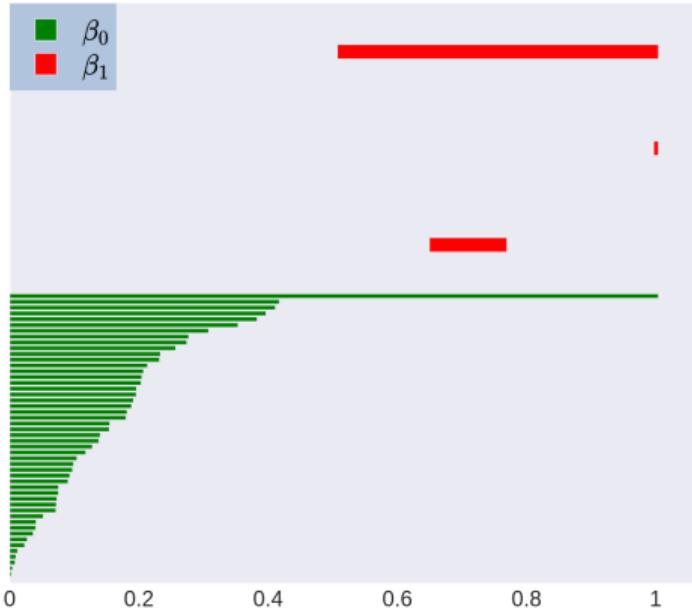
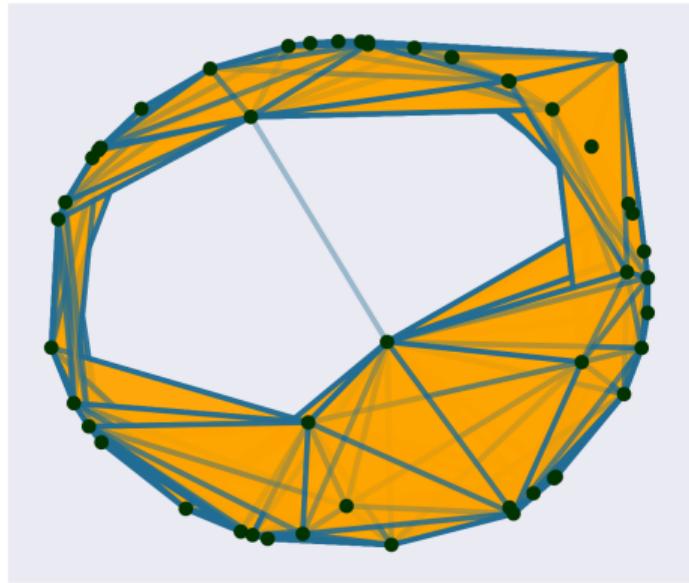
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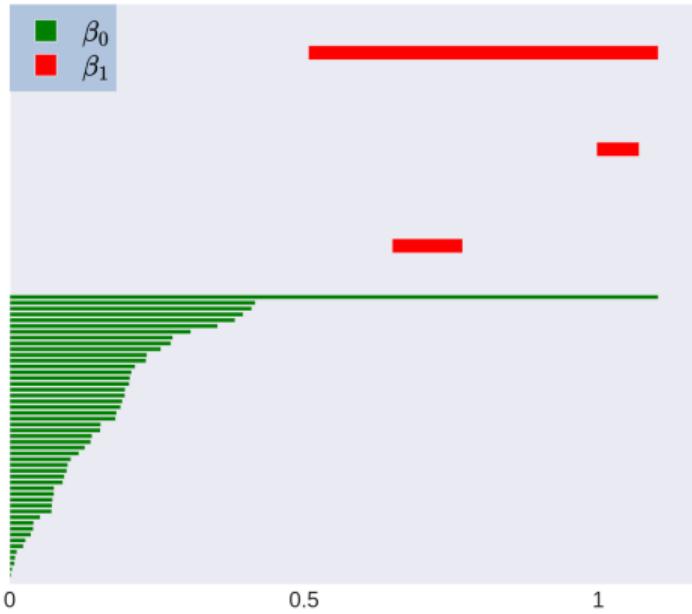
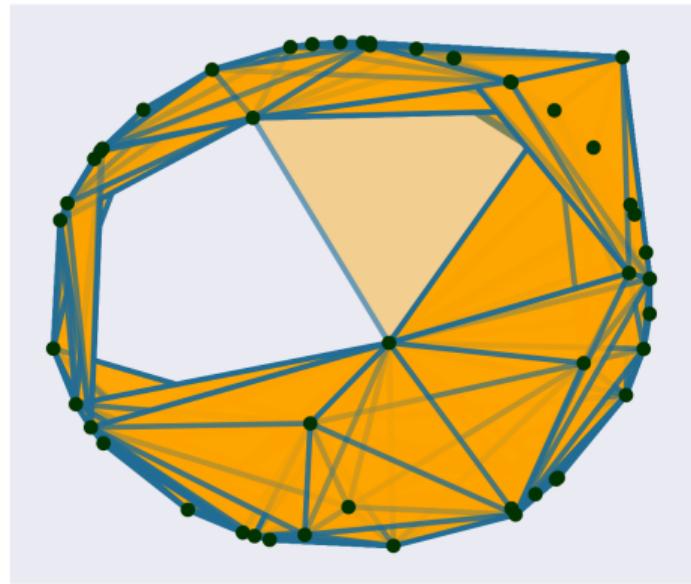
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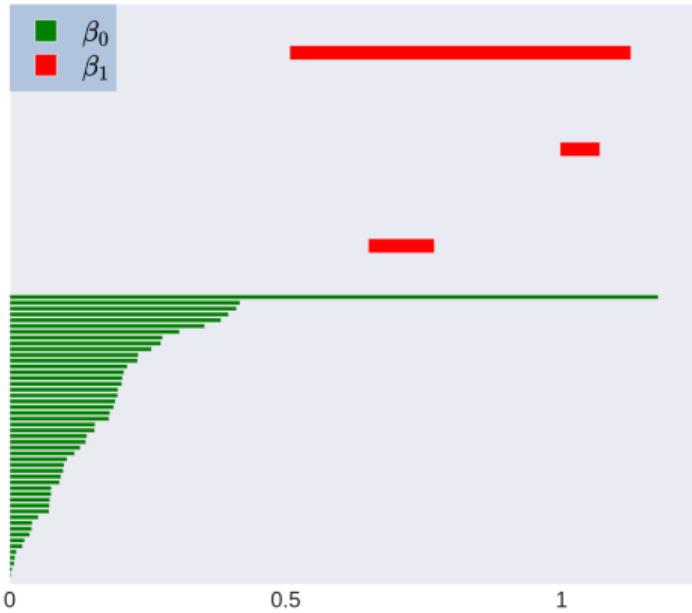
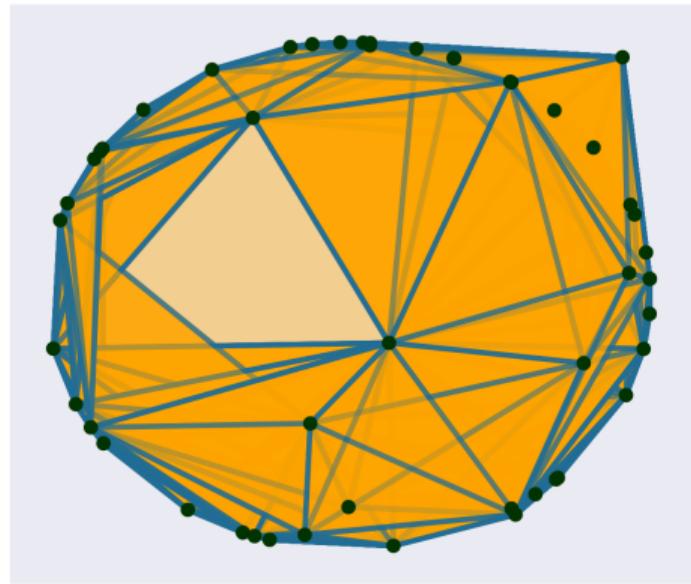
Persistent Homology and Barcodes



Persistent Homology and Barcodes



Persistent Homology and Barcodes



Persistent Homology

Filtered Vietoris-Rips complex

$$R_{\varepsilon_1} \hookrightarrow R_{\varepsilon_2} \hookrightarrow \cdots \hookrightarrow R_{\varepsilon_i} \hookrightarrow \cdots \hookrightarrow R_{\varepsilon_j} \hookrightarrow \cdots \cdots \hookrightarrow R_{\varepsilon_{max}}$$

After applying the homology functor,

$$H_k(R_{\varepsilon_1}) \rightarrow H_k(R_{\varepsilon_2}) \rightarrow \cdots \rightarrow H_k(R_{\varepsilon_i}) \rightarrow \cdots \rightarrow H_k(R_{\varepsilon_j}) \rightarrow \cdots \rightarrow H_k(R_{\varepsilon_{max}})$$

For every pair $\varepsilon_i, \varepsilon_j$

$$\psi_{\varepsilon_i, \varepsilon_j}^k : H_k(R_{\varepsilon_i}) \rightarrow H_k(R_{\varepsilon_j})$$

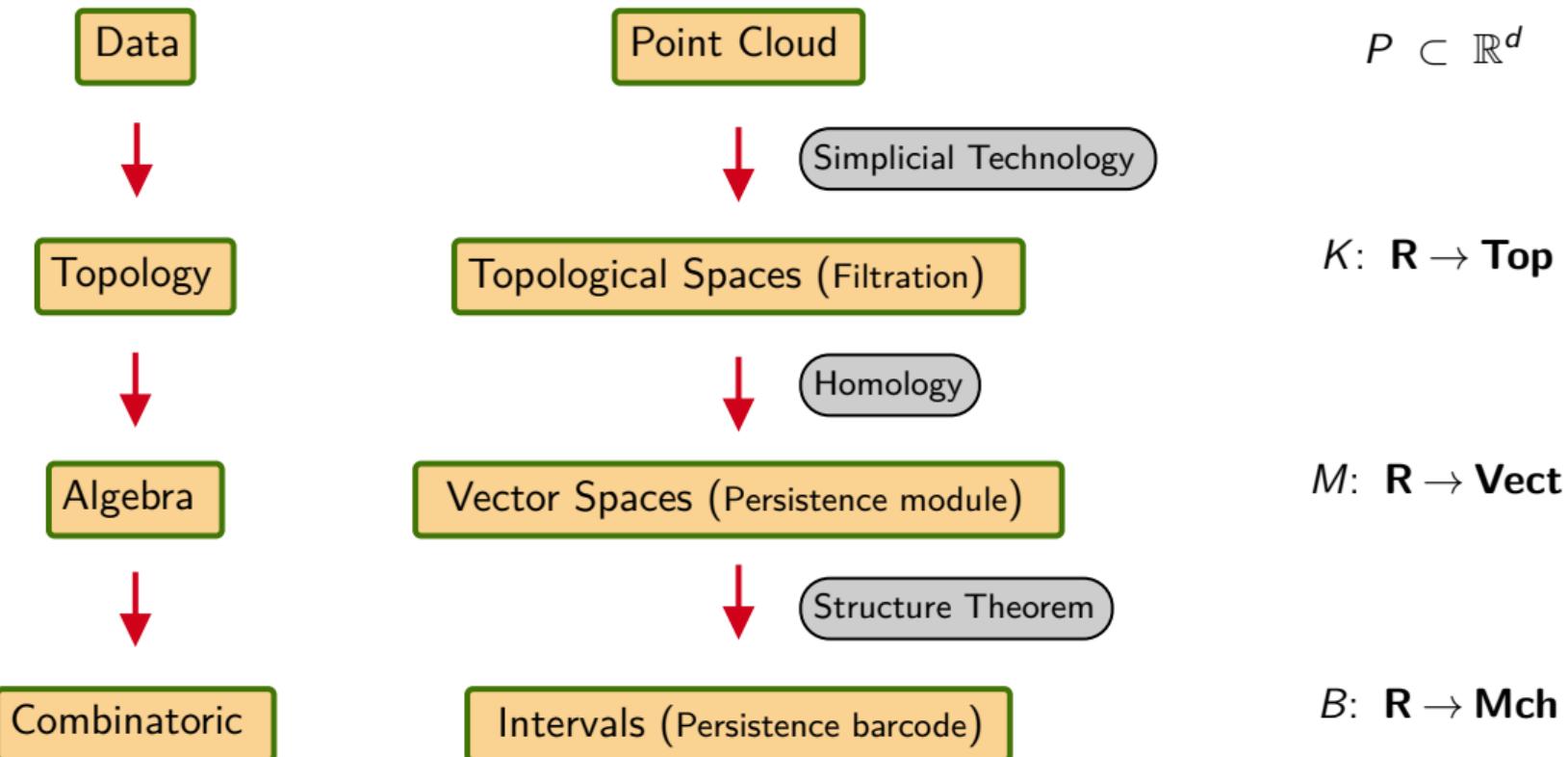
Definition

The k -th persistent homology group is

$$PH_k := \text{im } \psi_{\varepsilon_i, \varepsilon_j}^k.$$

Go Forward

Persistent Homology - Big Picture



Relation between homology classes

How do we connect points ?



Relation between homology classes

How do we connect points ?



$$\partial_1 = [0]$$

$$\partial_0 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Relation between homology classes

How do we connect points ?

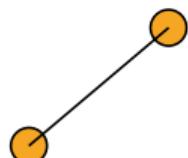
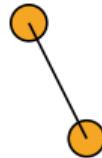


$$\partial_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

$$\partial_0 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Relation between homology classes

How do we connect points ?



$$\partial_1 = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \\ 0 & 1 \end{bmatrix}$$

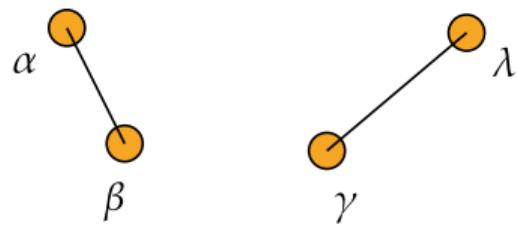
$$\partial_0 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Relation between homology classes

How do we connect points ?

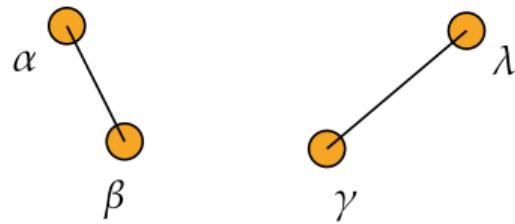
$$D = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

Rank = 2



Relation between homology classes

How do we connect points ?



$$D = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

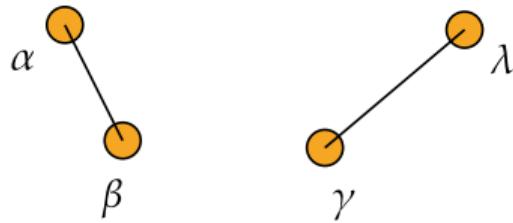
Rank = 2

$$D_{\alpha,\beta} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix}$$

Rank = 2

Relation between homology classes

How do we connect points ?



$$D = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

Rank = 2

$$D_{\alpha,\beta} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix}$$

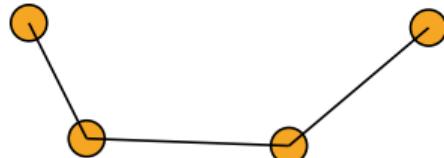
Rank = 2

$$D_{\alpha,\gamma} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

Rank = 3

Relation between homology classes

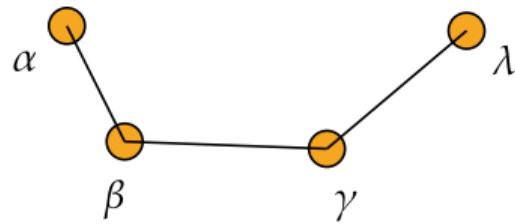
How do we connect points ?



$$\partial_1 = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 1 & 0 \end{bmatrix} \quad \partial_0 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Relation between homology classes

How do we connect points ?



$$D = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix}$$

Rank = 3

$$D_{\beta,\gamma} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \end{bmatrix}$$

Rank = 3

$$D_{\alpha,\gamma} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

Rank = 3

Homological Distance

- ① Given a point cloud P , construct the Rips Complex

$$R_\varepsilon(P) = \{\sigma \subset P \mid \|x - y\| \leq \varepsilon, \text{ for all } x, y \in \sigma\}$$

- ② $R_{\varepsilon_1} \subseteq R_{\varepsilon_2}$ for a given pair ε_1 and ε_2 values with $\varepsilon_1 \leq \varepsilon_2$
- ③ For the pair $\varepsilon_1 \leq \varepsilon_2$ we have the natural map in homology,

$$\psi_{\varepsilon_1, \varepsilon_2}^k : H_k(R_{\varepsilon_1}) \rightarrow H_k(R_{\varepsilon_2})$$

- ④ Take two cycles $\alpha, \beta \in H_k(\mathcal{R}_{\varepsilon_1})$
- ⑤ $\alpha' := \psi_{\varepsilon_1, \varepsilon_2}^k(\alpha)$ and $\beta' := \psi_{\varepsilon_1, \varepsilon_2}^k(\beta)$ by padding α and β with suitable number of 0's.

Homological Distance

- Add α' , β' and α', β' together to the differential matrix \mathcal{D}^t at ε_2 .
- Calculate rank of the differential matrices.
- Check whether $\psi_{\varepsilon_1, \varepsilon_2}^k(\alpha)$ and $\psi_{\varepsilon_1, \varepsilon_2}^k(\beta)$ are linearly dependent.

Homological Distance

k -th homological cophenetic distance $D_k(\alpha, \beta)$ between homology classes α and β is defined as

$$\inf \left\{ \eta - \varepsilon \geq 0 \mid \psi_{\varepsilon, \eta}^k(\alpha), \psi_{\varepsilon, \eta}^k(\beta) \text{ non-zero and lin. dep.} \right\}$$

Experiments

A synthetic point cloud

- $D \in \mathbb{R}^2$ and $|D| = 20$,
- Uniform distribution over $[0, 1)$,
- Labeled with the first 20 letters.

Input: A point cloud D , $|D| = 20$ and
a list $\varepsilon = \{\varepsilon_1 = 0, 0.05, \dots, 0.95, \varepsilon_{max} = 1\}$.

Output: Dendograms

begin

$\text{HomDist} = []_{20 \times 20}$

$\mathcal{R}_\varepsilon(P) \leftarrow \text{Vietoris-Rips filtration} ;$

for each ε **do**

for every $\alpha_i, \alpha_j \in H_0(\mathcal{R}_\varepsilon)$ **do**

$\text{HomDist}_{i,j} \leftarrow \inf\{\varepsilon \mid \text{check lin. dep.}\} ;$

end

end

$E(D) \leftarrow \text{EuclideanDist}(D);$

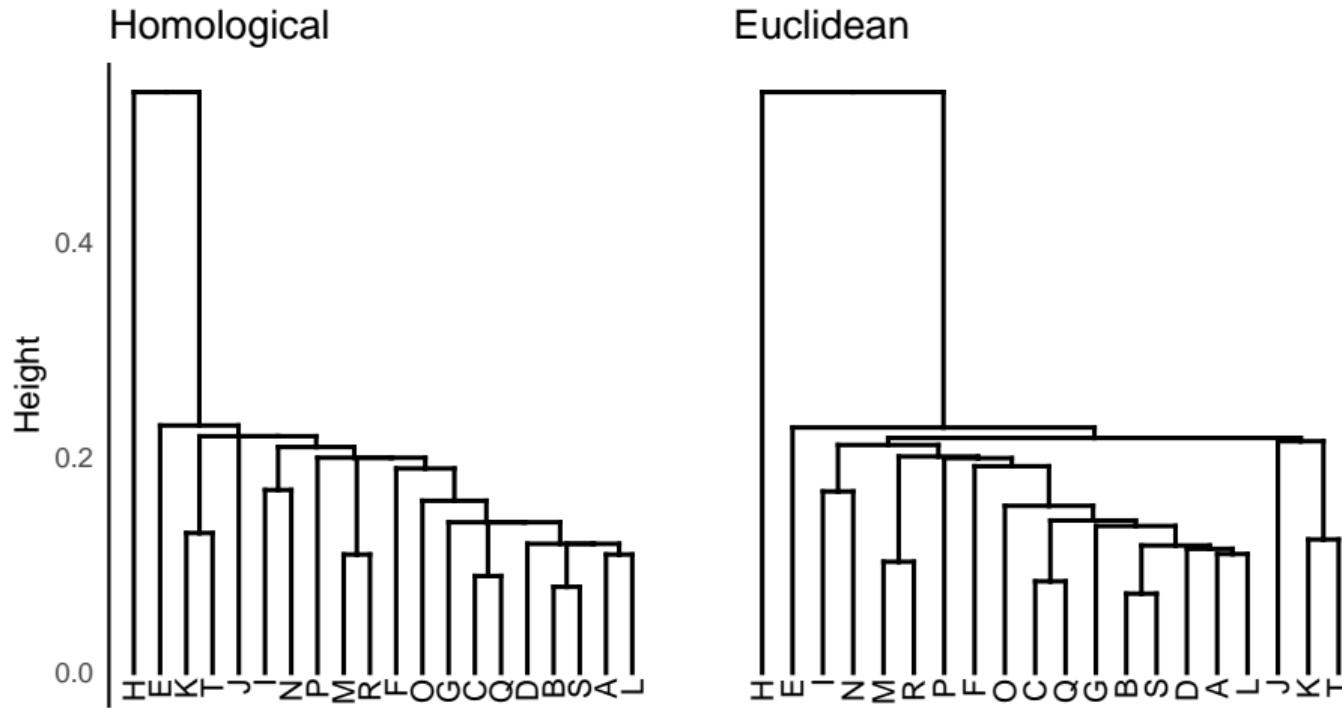
$Dend_1 \leftarrow \text{HierarchicalClustering}(\text{HomDist}(D)) ;$

$Dend_2 \leftarrow \text{HierarchicalClustering}(E(D)) ;$

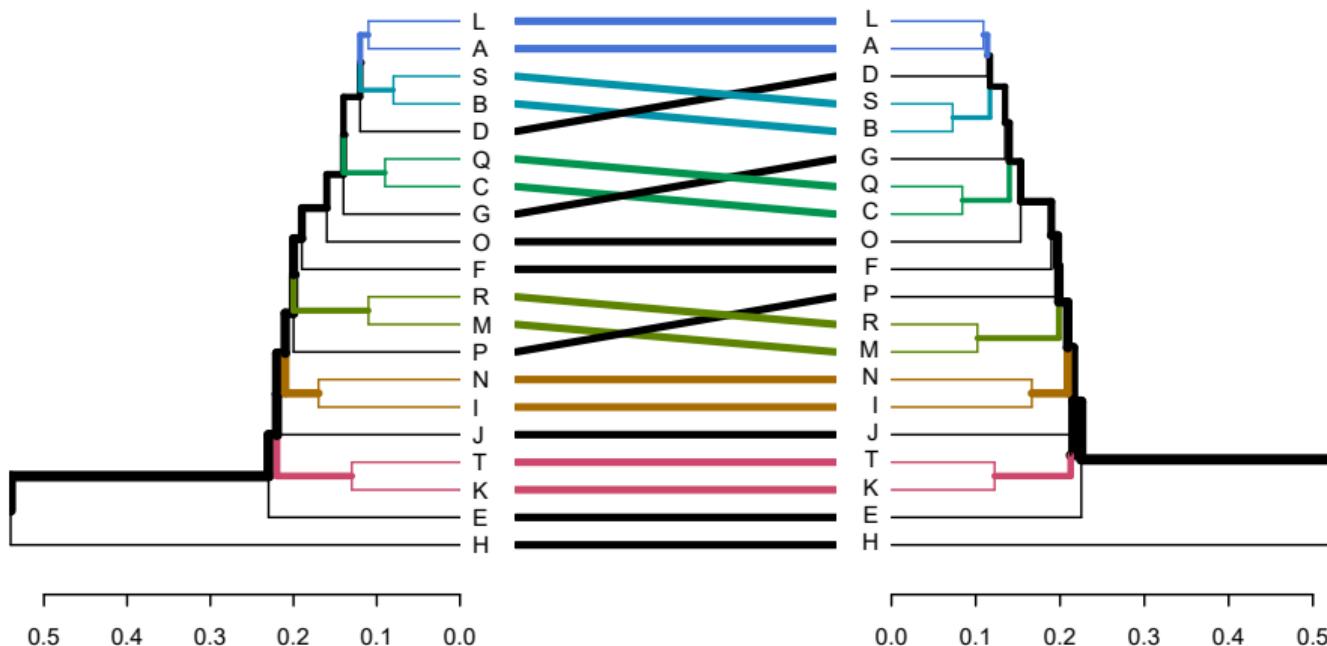
$\text{Compare}(Dend_1, Dend_2)$

end

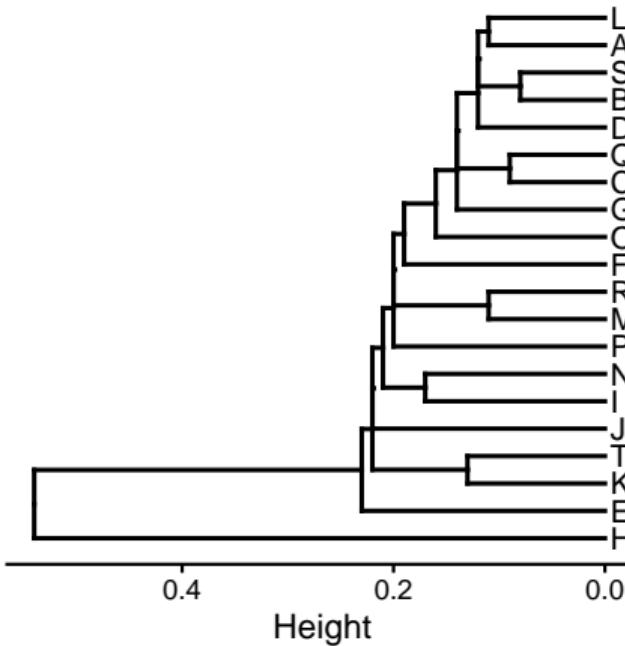
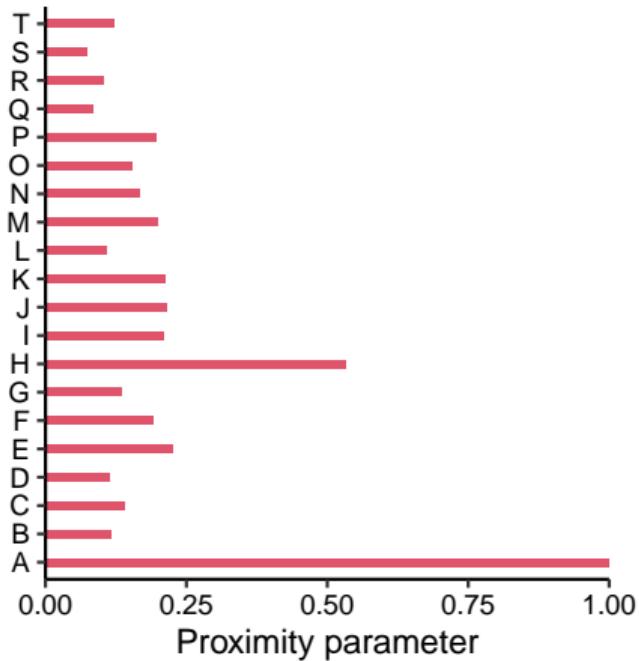
Two dendograms with single-linkage



Tanglegram with the entanglement of 0.01



Barcode and Enriched Barcode



Turkish Cities and Mantel Statistics



- 24 Türkiye cities
- Single-linkage
- Different metrics
- Dendograms

Metrics	Bray-Curtis	Cosine	Manhattan	Euclidean	Minkowski	Homological
Bray-Curtis	1.00	0.64	0.96	0.90	0.90	0.90
Cosine		1.00	0.61	0.52	0.69	0.59
Manhattan			1.00	0.96	0.87	0.97
Euclidean				1.00	0.75	0.98
Minkowski					1.00	0.78
Homological						1.00

Turkish Cities and Mantel Statistics



- 24 Türkiye cities
- Single-linkage
- Different metrics
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Metrics	Bray-Curtis	Cosine	Manhattan	Euclidean	Minkowski	Homological
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Euclidean				1.00	0.75	0.98
Minkowski					1.00	0.78
Homological						1.00

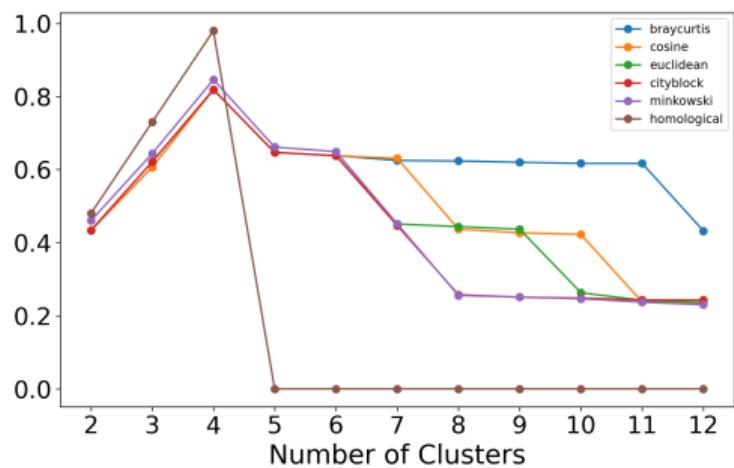
Datasets

Table: Datasets used and their properties.

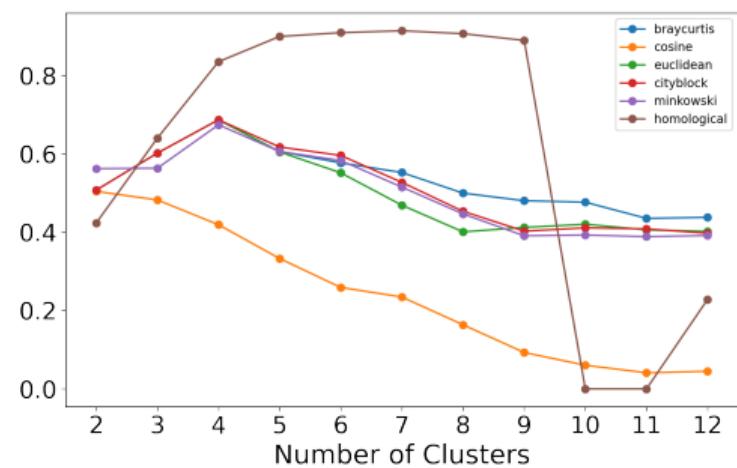
Dataset	#Instances	#Attributes	Supervised	#Classes
Turkish Cities	82	2	No	-
Iris	150	4	Yes	3
Cancer Coimbra	116	10	Yes	2
Synthetic (total separation)	100	100	Yes	4
Synthetic (with mixture)	100	2	Yes	4

Silhouette Scores

Synthetic Perfect

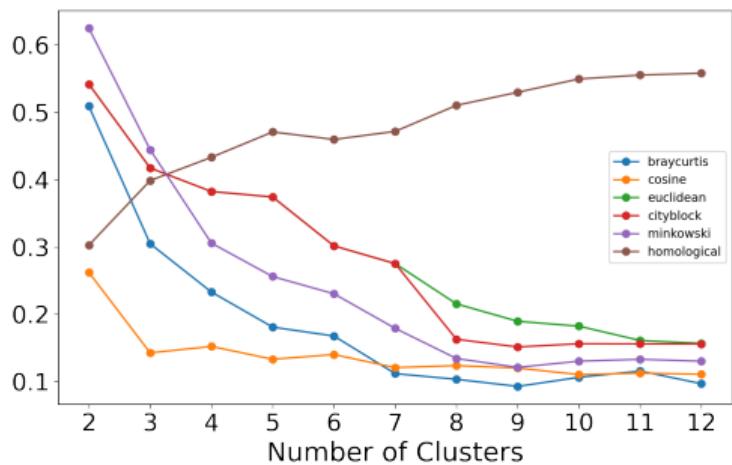


Synthetic Mixed

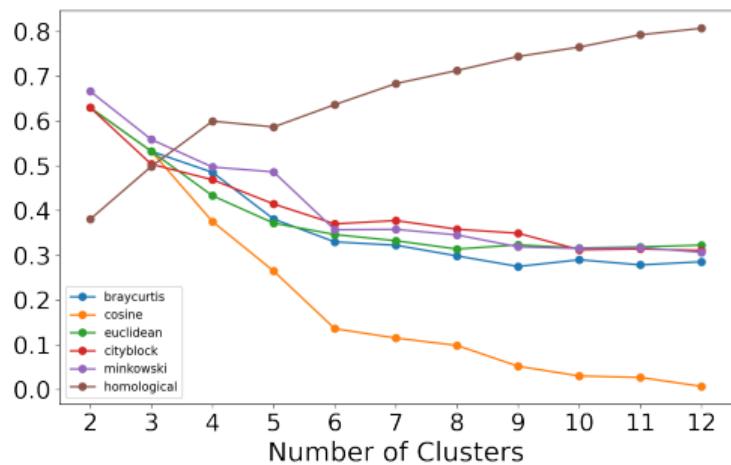


Silhouette Scores

Cancer Coimbra



Iris



A comparison of metrics on Mixed Synthetic datasets.

Synthetic Dataset

Metric	F1	Acc.	Hom.	Comp.	M.Info	Rand
Bray-Curtis	1.00 A	1.00 A	1.00 A	1.00 A	1.38 A	1.00 A
Cosine	0.83 A	0.91 A	0.72 C	0.77 S	1.38 C	0.87 A
Manhattan	1.00 S	1.00 S	1.00 S	1.00 S	0.99 S	1.00 S
Euclidean	1.00 A	1.00 A	1.00 A	1.00 A	1.38 A	1.00 A
Minkowski	1.00 C	1.00 C	1.00 C	1.00 C	1.38 C	1.00 C
Homological	0.98 A	0.99 A	0.95 A	1.00 S	1.31 A	0.98 A

A comparison of metrics on Cancer datasets.

Real Dataset

Metric	F1	Acc.	Hom.	Comp.	M.Info	Rand
Bray-Curtis	0.56 S	0.56 S	0.02 A	0.14 S	0.02 A	0.50 S
Cosine	0.55 C	0.55 C	0.01 S	0.12 S	0.01 S	0.50 C
Manhattan	0.53 S	0.53 S	0.02 A	0.13 A	0.02 A	0.50 S
Euclidean	0.54 W	0.54 W	0.02 A	0.13 A	0.02 A	0.50 W
Minkowski	0.53 S	0.53 S	0.02 A	0.13 A	0.02 A	0.50 S
Homological	0.61 W	0.61 W	0.03 W	1.00 S	0.02 W	0.52 W

BAP, TÜBİTAK, Michigan State University, Other Works

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Classification of Stochastic Processes with Topological Data Analysis

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Research Article

Attitudes and Behaviors of Turkish Consumers Regarding the Olive Oil Consumption

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Ismail Guzel (İTÜ)

Ph.D. Thesis

SIAM Data Mining: TDA, ML

A Case Study on Identifying Bifurcation and Chaos with CROCKER Plots

İsmail Güzel * Elizabeth Munch † Firas Khasawneh ‡

Abstract

The CROCKER plot is a coarsened but easy to visualize representation of the data in a one-parameter varying family of persistence barcodes. In this paper, we use the CROCKER plot to view changes in the persistence under a varying bifurcation parameter. We perform experiments to support our methods using the Rössler and Lorenz system and show the relationship with common methods for bifurcation analysis such as the Lyapunov exponent.



Article

Phenolic Constituents, Antioxidant and Antimicrobial Activity and Clustering Analysis of Propolis Samples Based on PCA from Different Regions of Anatolia

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March 13, 2022

31 / 33

BAP, TÜBİTAK, Michigan State University, Other Works

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SIAM Data Mining: TDA, ML

Classification
Topology

AIP Chaos: An Interdisciplinary Journal of Nonlinear Science

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Featured

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Ismail Güzel, Elizabeth Munch and Firas A. Khasawneh

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C. Kanchana, J. A. Vélez, L. M. Pérez, et al.

Templex: A bridge between homologies and templates for chaotic attractors
Gisela D. Charó, Christophe Letellier and Denisse Sclamarella

Generalized multistability and its control in a laser
Riccardo Meucci, Jean Marc Ginoux, Mahtab Mehrabbeik, et al.

Chimeras on annuli
Carlo R. Laing

Editor's picks

¹Istanbul Technical University, Faculty of Electrical and Electronics Engineering, Department of Mathematics, Istanbul, Turkey
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³Istanbul Technical University, Faculty of Electrical and Electronics Engineering, Department of Mathematics, Istanbul, Turkey
⁴Bayburt University, Faculty of Engineering, Department of Mathematics, Bayburt, Turkey
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Ismail Guzel (ITU)

Ph.D. Thesis

March 13, 2022

31 / 33

Future Work

On this thesis

- Apply real-world dataset for the first degree of homology.
- Visualization tools for the first degree of homology.
- Apply to the categorical dataset.
- Deal with problems about computational power and memory.

Other tasks

- Relation between Lyapunov exponent and persistent homology
- Two dimension bifurcation and CROCKER
- Classification Alpha-stable processes via TDA

Thank You!



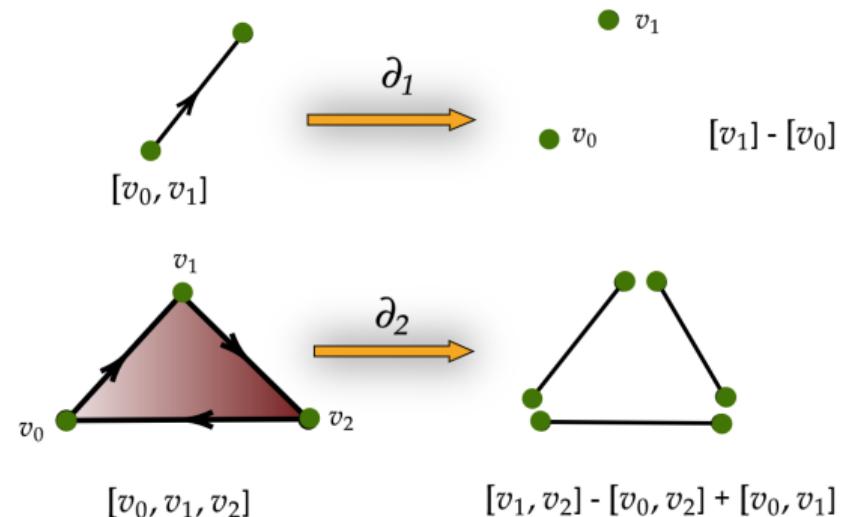
Separation of complex

Chain Complex

A chain complex of a simplicial complex \mathcal{K} is a sequence of abelian groups or modules \mathcal{C}_k connected by homomorphisms $\partial_k : \mathcal{C}_k \rightarrow \mathcal{C}_{k-1}$ such that $\partial_{k-1} \circ \partial_k = 0$ for $k \in \mathbb{Z}$.

$$\dots \xrightarrow{\partial_{k+2}} \mathcal{C}_{k+1} \xrightarrow{\partial_{k+1}} \mathcal{C}_k \xrightarrow{\partial_k} \dots$$

$$\partial_k \sigma = \sum_{i=0}^k (-1)^i [v_0, v_1, \dots, \hat{v}_i, \dots, v_k]$$



Homology and Betti Number

The k^{th} homology group of a simplicial complex K is defined by

$$H_k(K) := \ker(\partial_k) / \text{im}(\partial_{k+1}).$$

The dimension of the k^{th} homology group of K is called the k^{th} *Betti number* $\beta_k(K)$.

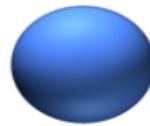
Point



Circle



Sphere



Torus



$$\beta_0 = 1$$

$$\beta_1 = 0$$

$$\beta_2 = 0$$

$$\beta_0 = 1$$

$$\beta_1 = 1$$

$$\beta_2 = 0$$

$$\beta_0 = 1$$

$$\beta_1 = 0$$

$$\beta_2 = 1$$

$$\beta_0 = 1$$

$$\beta_1 = 2$$

$$\beta_2 = 1$$

Well-defined Persistence barcode

- Persistence modules

$$V_1 \rightarrow V_2 \rightarrow \cdots \rightarrow V_r \rightarrow \cdots \rightarrow V_s \rightarrow \cdots \rightarrow V_n$$

- Decompose persistence module into interval modules,

$$0 \rightarrow 0 \rightarrow \cdots \rightarrow 0 \rightarrow \mathbb{Z}_2 \rightarrow \cdots \rightarrow \mathbb{Z}_2 \rightarrow 0 \rightarrow \cdots 0$$

- A persistence module V indexed by $T \subset \mathbb{R}$ is q -tame if for any $v < s$ in T , the rank of the linear map $v_r^s : V_r \rightarrow V_s$ is finite.
- If V is a q -tame persistence module, then it has a well-defined persistence barcode.

Go Back

Combinatoric Possibilities

- Check whether $\psi_{\varepsilon_1, \varepsilon_2}^k(\alpha)$ and $\psi_{\varepsilon_1, \varepsilon_2}^k(\beta)$ are linearly independent by evaluating the rank of the differential matrix \mathcal{D} at ε_2 with appending α' , β' and α', β' together.
-

$$r_\alpha := \text{rank}(\mathcal{D}_\alpha) - \text{rank}(\mathcal{D})$$

$$r_\beta := \text{rank}(\mathcal{D}_\beta) - \text{rank}(\mathcal{D})$$

$$r_{\alpha, \beta} := \text{rank}(\mathcal{D}_{\alpha, \beta}) - \text{rank}(\mathcal{D})$$

with the following cases:

$$\begin{cases} \alpha \text{ and } \beta \text{ both die} & \text{if } r_{\alpha, \beta} = 0, \\ \alpha \text{ and } \beta \text{ both live} & \text{if } r_{\alpha, \beta} = 2, \\ \alpha \text{ dies and } \beta \text{ lives} & \text{if } r_{\alpha, \beta} = 1 \text{ and } r_\alpha = 0, \\ \alpha \text{ lives and } \beta \text{ dies} & \text{if } r_{\alpha, \beta} = 1 \text{ and } r_\beta = 0, \\ \alpha \text{ and } \beta \text{ merge} & \text{if } r_{\alpha, \beta} = 1, r_\alpha = 1 \text{ and } r_\beta = 1. \end{cases}$$

Matroids

Definition

A partially ordered set is defined as an ordered pair $P = (X, \leq)$.

Here, X is called the ground set of P and \leq is the partial order of P

Definition

A matroid $M = (S, \mathbb{I})$ is a finite ground set S together with a collection of sets $\mathbb{I} \subset 2^S$ satisfying

- Downward closed: $A \in \mathbb{I}$ and $B \subseteq A \Rightarrow B \in \mathbb{I}$
- Exchange property: $A, B \in \mathbb{I}$ and $|B| < |A| \Rightarrow \exists x \in A \setminus B \text{ such that } \{x\} \cup B \in \mathbb{I}$.

Matroids

Terminology

- Independent set: $I \in \mathbb{I}$
- Circuit: Minimal dependent set of M
- Basis: Maximal independent set of M
- Span: Basis B and $B \subseteq \mathbb{I} \Rightarrow \mathbb{I}$ is a spanning set.

- Ground set \mathbb{V} : set of vectors spanning \mathbb{R}^d
- Independent set \mathbb{I} : bases of \mathbb{R}^d in \mathbb{V}
- Matroid: (\mathbb{V}, \mathbb{I})

The Rank Function of a Matroid

Definition

Let M be a matroid on a finite ground set E . The rank $r(X)$ of a subset $X \subseteq E$ is the cardinality of the largest independent set contained in X . In other words

$$r(X) = \max\{|A| \in N \mid A \subseteq X \text{ and } A \in \mathcal{I}\}$$

Cobordisms

For two linearly independent pair of homology classes α and β in $H_1(R_\varepsilon)$, one can see $\psi_{\varepsilon,\eta}^1(\alpha)$ and $\psi_{\varepsilon,\eta}^1(\beta)$ are linearly dependent in $H_1(R_\eta)$. We, then, visualize that two classes α and β merged at time η .

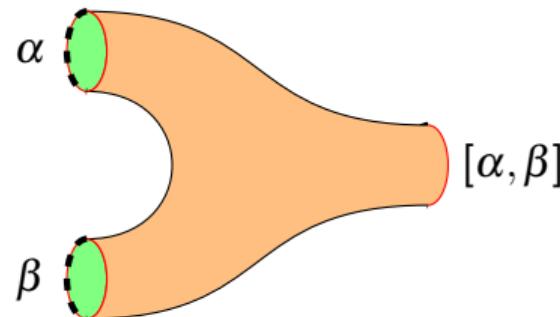


Figure: Cobordism in the merging case from $S^1 \sqcup S^1$ to S^1 representing two cycles α and β evolve in $[\alpha, \beta]$ from ε to η .