

Topology Tools for Explainable and Green Artificial Intelligence

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**Sixth EACA International School on Computer Algebra and its
Applications**

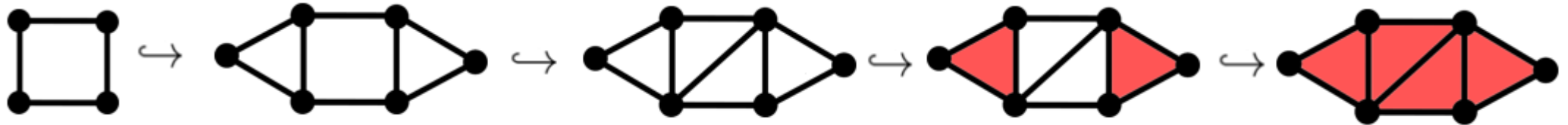
SANTIAGO DE COMPOSTELA, JULY 18-21, 2023



- Context: Green and Explainable artificial intelligence (REXASI-PRO)
- Computational topology tools: Persistent homology, barcodes, distance bottleneck, simplicial maps, Persistence modules, morphisms between persistence modules
- Partial matchings between barcodes
- Simplicial maps neural networks

Concepts from computational topology

- (abstract) Simplicial complexes
- Homology
- Vertex map, simplicial map
- Simplicial Approximation Theorem
- Category SpCpx : simplicial complexes, simplicial maps
- Filtration
- Persistent homology
- **Persistence module**



$$H_1(K_1) \longrightarrow H_1(K_2) \longrightarrow H_1(K_3) \longrightarrow H_1(K_4) \longrightarrow H_1(K_5)$$

$$V = PH_1(K)$$

$$0 \longrightarrow V_1 \xrightarrow{\rho_1^2} V_2 \xrightarrow{\rho_2^3} V_3 \xrightarrow{\rho_3^4} V_4 \xrightarrow{\rho_4^5} V_5 \longrightarrow 0$$

persistence module

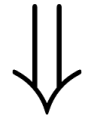
Concepts from computational topology

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- **Persistence module**

A persistence module can be seen as a functor from $\{1, \dots, n\}$ to the category of vector spaces

This definition can be extended to any other totally ordered set

$$\textcircled{1} \leq \textcircled{2} \leq \dots \leq \textcircled{n}$$



$$0 \xrightarrow{\rho_0^1} V_1 \xrightarrow{\rho_1^2} V_2 \xrightarrow{\rho_2^3} \dots \xrightarrow{\rho_{n-1}^n} V_n \xrightarrow{\rho_n^{n+1}} 0$$

$$(V_t, \rho_p^q)$$

- $\rho_q^l \rho_p^q = \rho_p^l$ if $0 \leq p \leq q \leq l \leq n+1$
- ρ_p^p is the identity map

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- **Persistence module**

The category of persistence modules satisfies that

- The **direct sum** of persistence modules is a persistence module
- The **intersection** of persistence modules is a persistence module
- The **quotient** of persistence modules is a persistence module
- The notion of **submodule** is well defined

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- **Persistence module**

Example: $B = \{([1,4], 1), ([2,3], 2), ([3,4], 1)\}$

$\text{Rep } B = \langle [1,4], [2,3], [2,3], [3,4] \rangle$

$\text{SS} = \{[1,4], [2,3], [3,4]\}$

$$V \simeq \bigoplus_{I \in S} \left(\bigoplus_{m_I} k_I \right) \quad \begin{array}{l} B = \{(I, m_I)\} \\ S = \{I\} \end{array}$$

Interval module $k_{It} = \begin{cases} 1 & \text{if } t \in I \\ 0 & \text{otherwise} \end{cases}$

$$I = [a, b] \quad t \in I$$

$$\text{Im}_{at}^+(V) := \text{im } \rho_a^t, \quad \text{Ker}_{bt}^+(V) := \text{ker } \rho_t^{b+1}$$

$$\text{Im}_{at}^-(V) := \text{im } \rho_{a-1}^t, \quad \text{Ker}_{bt}^-(V) := \text{ker } \rho_t^b$$

$$k_I \simeq \frac{V_{It}^+}{V_{It}^-}$$

$$m_I = \dim \frac{V_{It}^+}{V_{It}^-}$$

$$V_{It}^+ := \text{Im}_{at}^+(V) \cap \text{Ker}_{bt}^+(V)$$

$$V_{It}^- := \text{Im}_{at}^-(V) \cap \text{Ker}_{bt}^+(V) + \text{Im}_{at}^+(V) \cap \text{Ker}_{bt}^-(V)$$

Concepts from computational topology

- (abstract) Simplicial complexes
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- **Morphism between persistence modules**

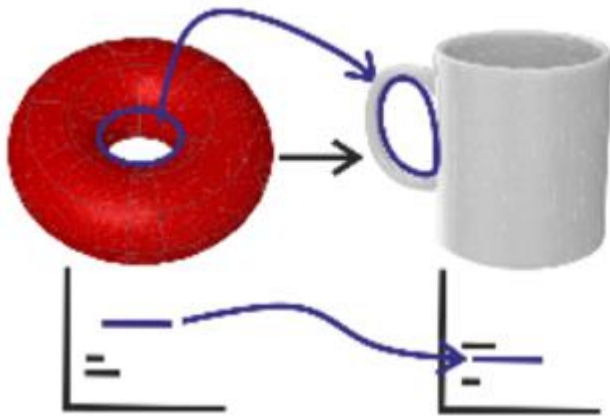
$$\begin{array}{ccccccc} U & & U_1 & \longrightarrow & U_2 & \longrightarrow & \dots & \longrightarrow & U_n \\ f \uparrow & & f_1 \uparrow & & f_2 \uparrow & & & & f_n \uparrow \\ V & & V_1 & \longrightarrow & V_2 & \longrightarrow & \dots & \longrightarrow & V_n \end{array}$$

Example:

- Let $\mathbb{X}$ and $\mathbb{Y}$ be two finite subsets from $\mathbb{R}^n$ such that $\mathbb{X} \subseteq \mathbb{Y}$.
- This induces an embedding $\text{VR}(\mathbb{X}) \hookrightarrow \text{VR}(\mathbb{Y})$.
- In turn, this induces a persistence morphism $f : V \rightarrow U$, where $V = \text{PH}_n(\text{VR}(\mathbb{X}))$ and $U = \text{PH}_n(\text{VR}(\mathbb{Y}))$

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Partial matchings between barcodes

Joint work with Manuel Soriano-Trigueros and Álvaro Torras-Casas



Manuel Soriano-
Trigueros

PhD student
msoriano4@us.es



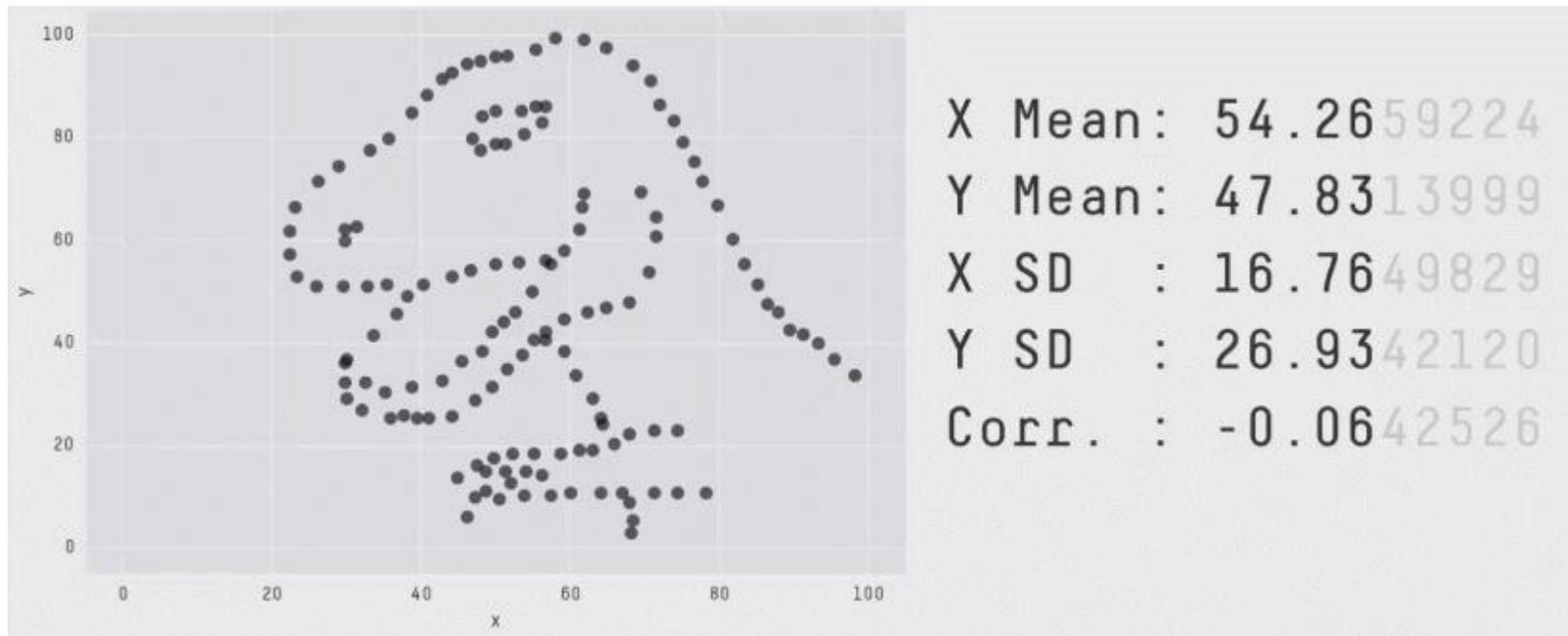
Álvaro Torras
Casas

Research Associate
(Cardiff University)
atorras@us.es

Partial matchings between barcodes

A motivation

Stats is not enough:



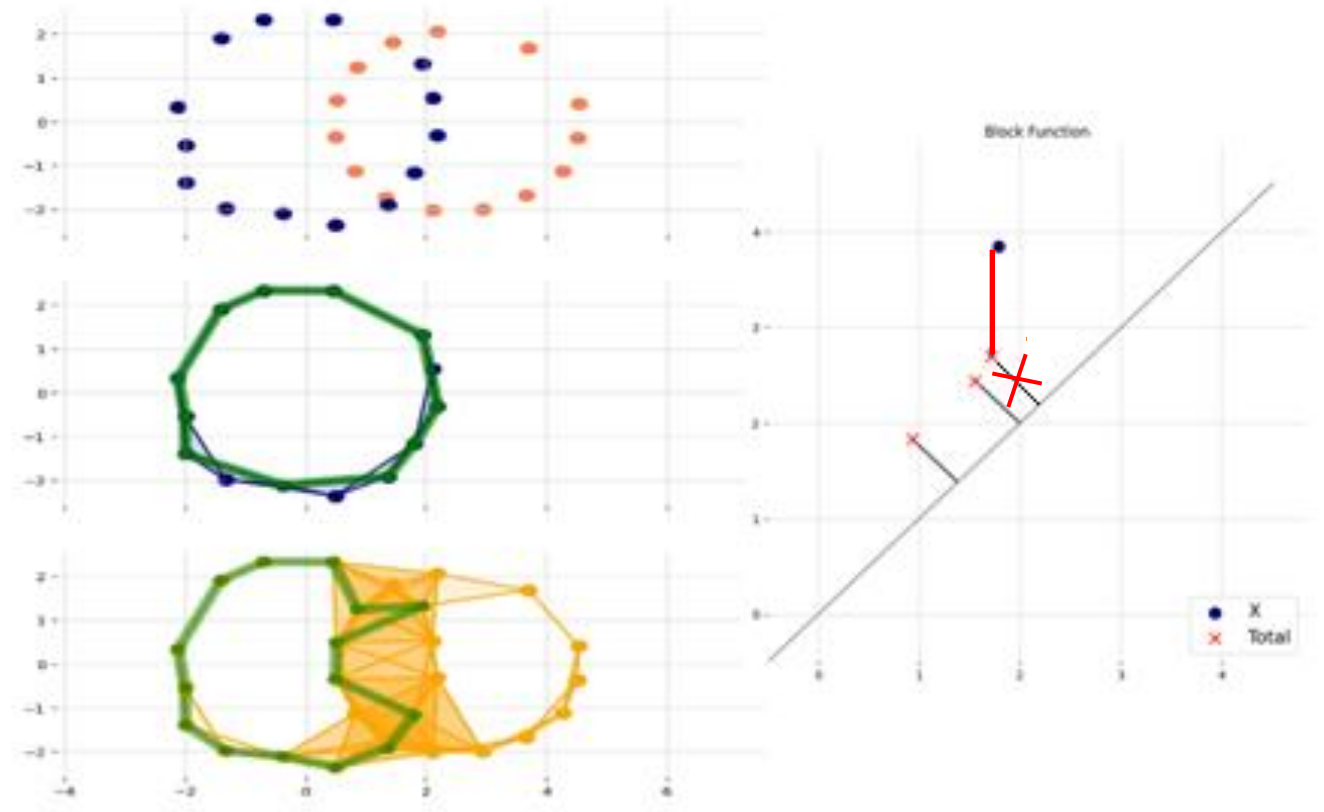
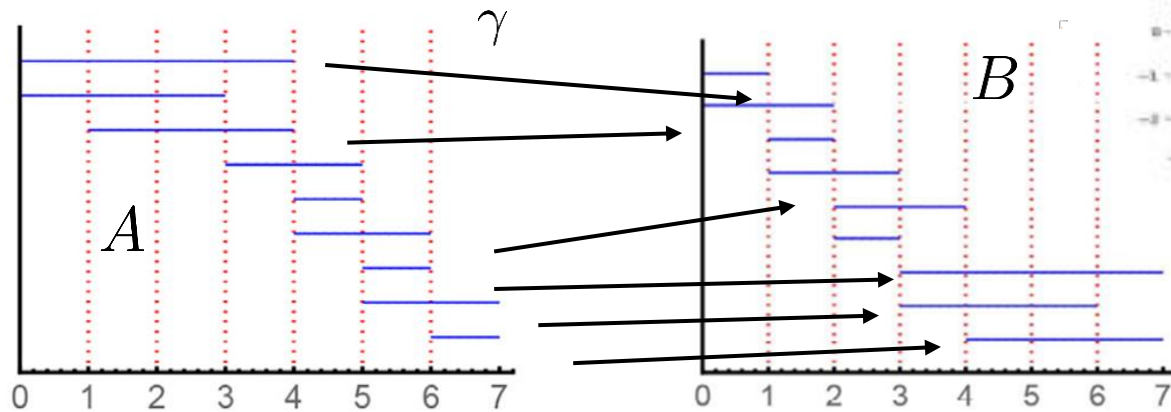
**Datasaurus
dozen dataset**

Partial matchings between barcodes

A motivation

Bottleneck distance is not enough:

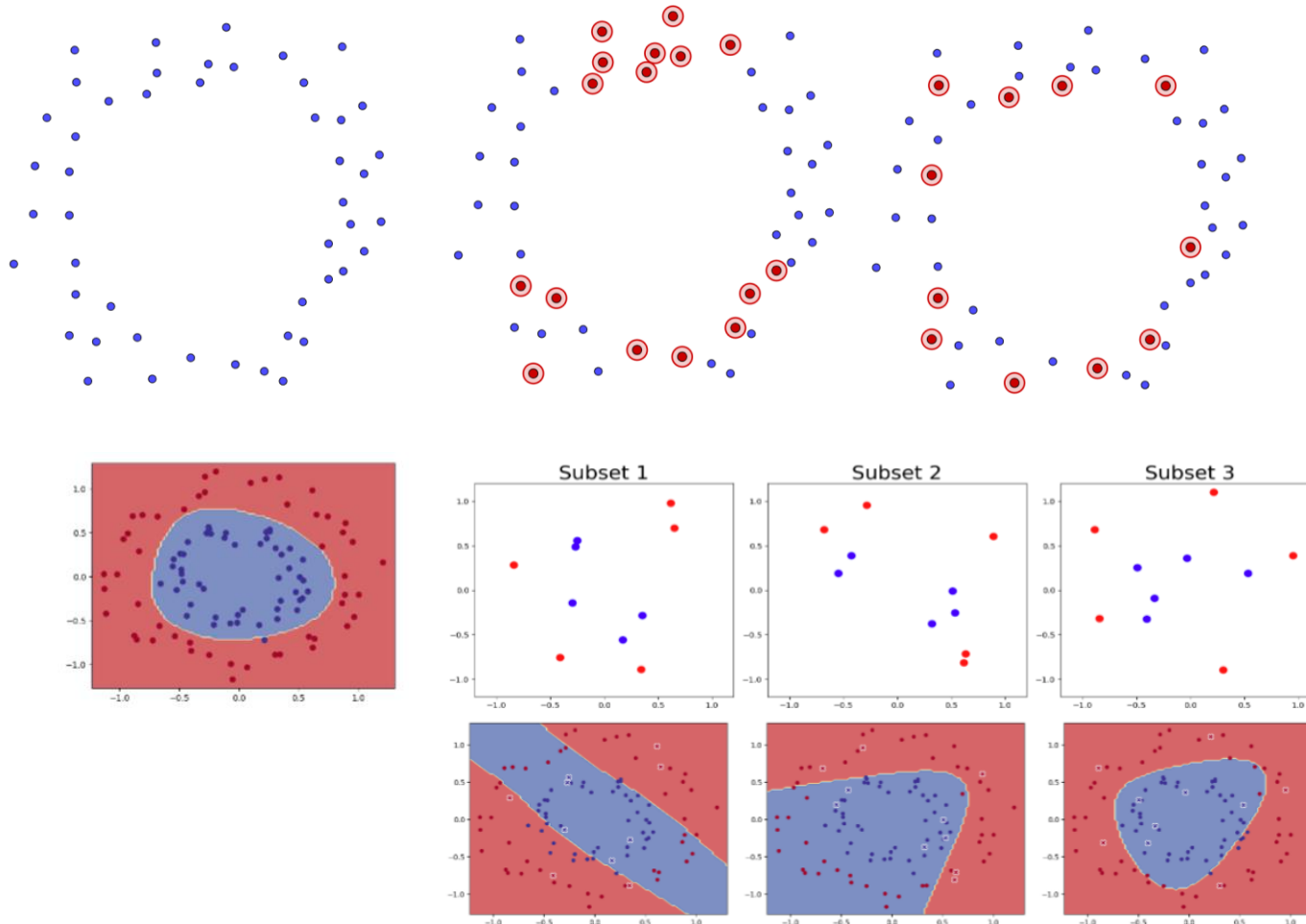
RECALL:



Partial matchings between barcodes

A motivation

When can we say that a subset Y of a given a dataset X “samples” the same continuous space than X ?

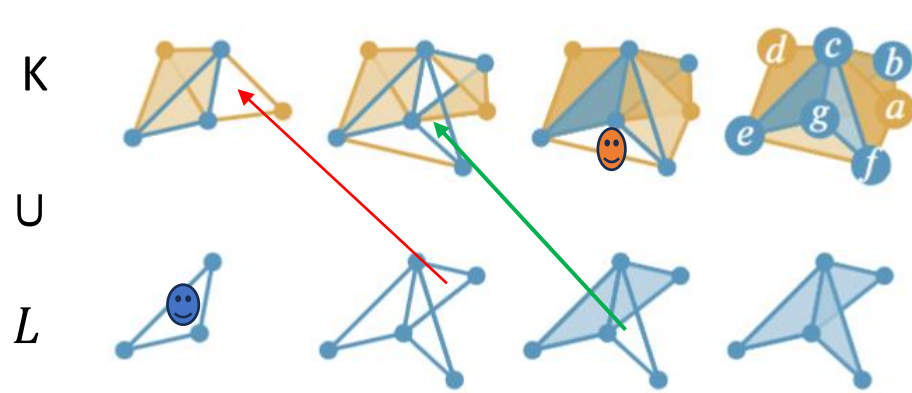


Why is it important?

- Data hungry
- Energy saving
- Storage saving
- Data is expensive

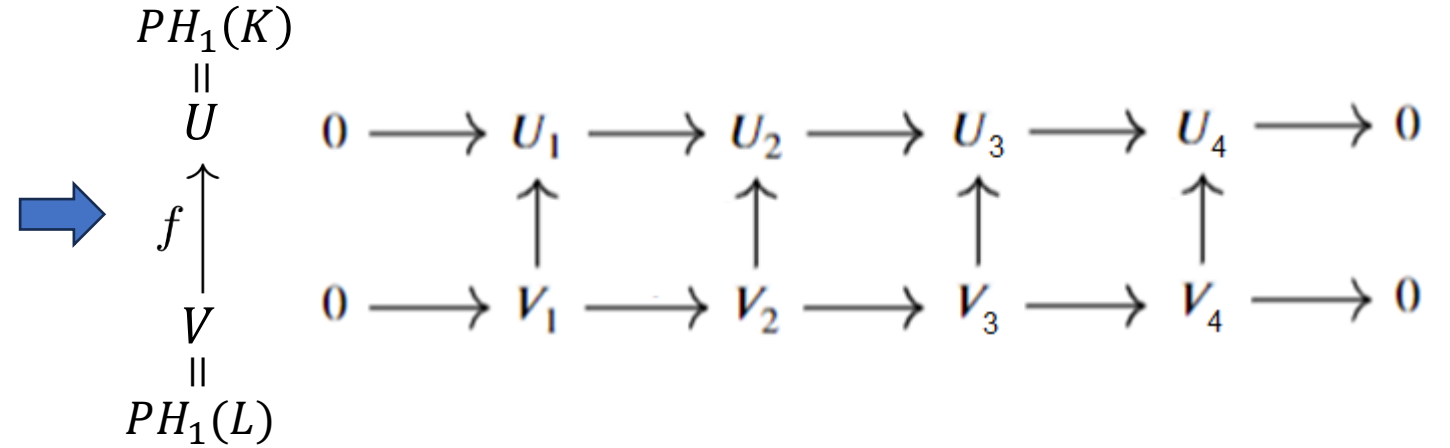
Partial matchings between barcodes

Example:



$$PH_1(K) \simeq k_{[1,2]} \oplus k_{[2,2]} \oplus k_{[2,3]}^{\text{orange smiley}}$$

$$PH_1(L) \simeq k_{[1,2]} \oplus k_{[2,2]} \oplus k_{[2,3]}^{\text{blue smiley}}$$



$$d_{\infty}(B(U), B(V)) = 0$$

Partial matchings between barcodes

Given $B(V) = \{(I, m_I)\}$ and $B(U) = \{(J, m_J)\}$,
we define a ~~partial matching~~ satisfying

block function

$$S_V = \{I\}$$

$$S_U = \{J\}$$

$$\mathcal{M} : S_V \times S_U \longrightarrow \mathbb{Z}_{\geq 0}$$

$$\sum_{J \in S_U} \mathcal{M}(I, J) \leq m_I$$

The number of arrows leaving from an interval is smaller than its multiplicity

~~$$\sum_{I \in S_V} \mathcal{M}(I, J) \leq m_J$$~~

~~The number of arrows arriving to an interval is smaller than its multiplicity~~

Partial matchings between barcodes

Block function: $\mathcal{M} : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0} \quad \sum_{J \in S_U} \mathcal{M}(I, J) \leq m_I$

Example:

$B(V) = \{([2, 4], 1), ([1, 5], 2)\}$ and $B(U) = \{([2, 3], 1), ([1, 4], 2)\}$

Consider $\mathcal{M}$ is zero except for

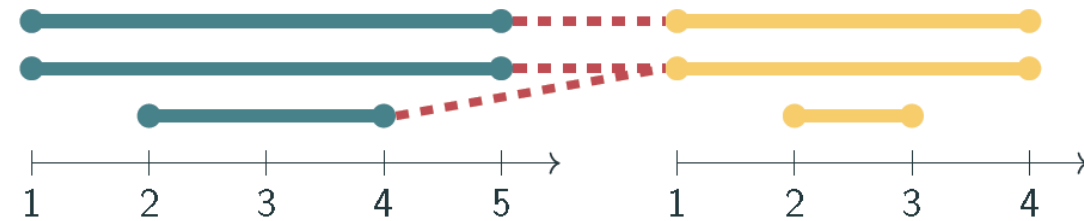
$$\mathcal{M}([2, 4], [1, 4]) = 1 \text{ and } \mathcal{M}([1, 5], [1, 4]) = 2.$$

$\mathcal{M}$ is a block function, since

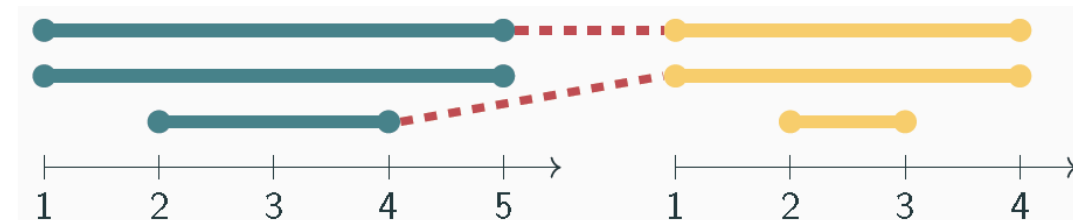
$$\mathcal{M}([2, 4], [1, 4]) = 1 \leq m_{[2, 4]} \text{ and } \mathcal{M}([1, 5], [1, 4]) = 2 \leq m_{[1, 5]}$$

$\mathcal{M}$ is not a partial matching

$$\mathcal{M}([2, 4], [1, 4]) + \mathcal{M}([1, 5], [1, 4]) = 3 \not\leq n_{[1, 4]} = 2.$$



We can always obtain a partial matching γ from a block function



$$\sum_{J \in S_2} \gamma(I, J) \leq m_I \quad \sum_{I \in S_1} \gamma(I, J) \leq n_J$$

Partial matchings between barcodes

The induced block function: $\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0}$ $\sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$

$$\begin{array}{ccc}
 V & \xrightarrow{\quad f \quad} & U \\
 (I, m_I) & \xrightarrow{\quad \mathcal{M}_f(I, J) \quad} & (J, m_J)
 \end{array}$$

Partial matchings between barcodes

The induced block function: $\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0}$ $\sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$

Partial Matchings Induced by Morphisms
between Persistence Modules. Computational
Geometry, 112, 101985 (2023)

$$\begin{array}{ccc}
 V & \xrightarrow{f} & U \\
 \downarrow (I, m_I) & \xrightarrow{\mathcal{M}_f(I, J)} & (J, m_J) \downarrow \\
 \textcircled{V_{I_t}^+ / V_{I_t}^-} & & \textcircled{U_{J_t}^+ / U_{J_t}^-} \\
 \searrow & & \swarrow \\
 X_{IJt} = \frac{fV_{I_t}^+ \cap U_{J_t}^+}{fV_{I_t}^- \cap U_{J_t}^+ + V_{I_t}^+ \cap U_{J_t}^-}
 \end{array}$$

$$\mathcal{M}_f(I, J) = \dim X_{IJt}$$

$$t \in I \cap J$$

Partial matchings between barcodes

The induced block function: $\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0}$ $\sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$

Let $I = [a, b]$ and $J = [c, d]$.

$$\begin{aligned} \mathcal{M}_f(I, J) &= \dim X_{IJd} && \text{if } J \leq I \\ \mathcal{M}_f(I, J) &= 0 && \text{otherwise} \end{aligned}$$

Partial Matchings Induced by Morphisms
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We say that $J \leq I$ if $c \leq a \leq d \leq b$.

$$X_{IJt} = \frac{fV_{It}^+ \cap U_{Jt}^+}{fV_{It}^- \cap U_{Jt}^+ + V_{It}^+ \cap U_{Jt}^-}$$

Partial matchings between barcodes

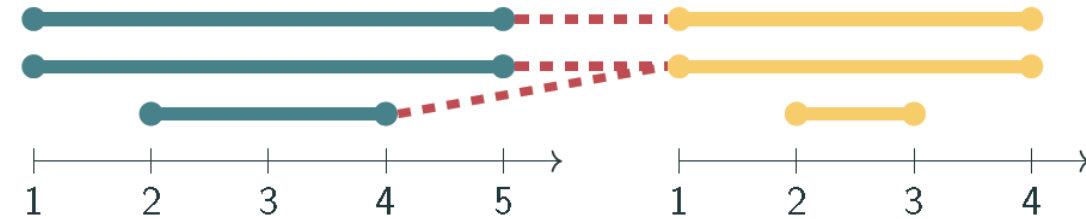
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Partial Matchings Induced by Morphisms
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Nested intervals:

- $[a, b]$ and $[c, d]$ are nested if $a < c < d < b$



Prop.: If for any set of intervals $S \subseteq S_V$ we have that

$$\sum_{I \in S} \mathcal{M}_f(I, J) > n_J,$$

then there exists a pair of nested intervals in S_V .

Corollary: If there are no two nested intervals in S_V then $\mathcal{M}_f$ is a partial matching.

Partial matchings between barcodes

Linearity of the induced block function:

Given a direct sum of morphisms of persistence modules

$$f^1 \oplus f^2 : V^1 \oplus V^2 \rightarrow U^1 \oplus U^2$$

we have that

$$\mathcal{M}_{f^1 \oplus f^2}(I, J) = \mathcal{M}_{f^1}(I, J) + \mathcal{M}_{f^2}(I, J)$$

Example:

$$\begin{array}{ccccccc} U & & k & \xrightarrow{\text{Id}} & k & \longrightarrow & 0 \\ f \uparrow & \simeq & \uparrow & & \begin{pmatrix} 0 \\ 1 \end{pmatrix} \uparrow & & \uparrow \\ V & & 0 & \longrightarrow & k^2 & \xrightarrow{\begin{pmatrix} 1 \\ 0 \end{pmatrix}} & k \end{array}$$

$\mathcal{M}_f ?$

Partial Matchings Induced by Morphisms
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$$\begin{array}{ccccccc} U & & 0 & \longrightarrow & 0 & \longrightarrow & 0 \\ f \uparrow & \simeq & \uparrow & & \uparrow & & \uparrow \\ V & & 0 & \longrightarrow & k & \xrightarrow{\text{Id}} & k \end{array} \oplus \begin{array}{ccccccc} k & \xrightarrow{\text{Id}} & k & \longrightarrow & 0 \\ \uparrow & & \uparrow_{\text{Id}} & & \uparrow \\ 0 & \longrightarrow & k & \longrightarrow & 0 \end{array}$$

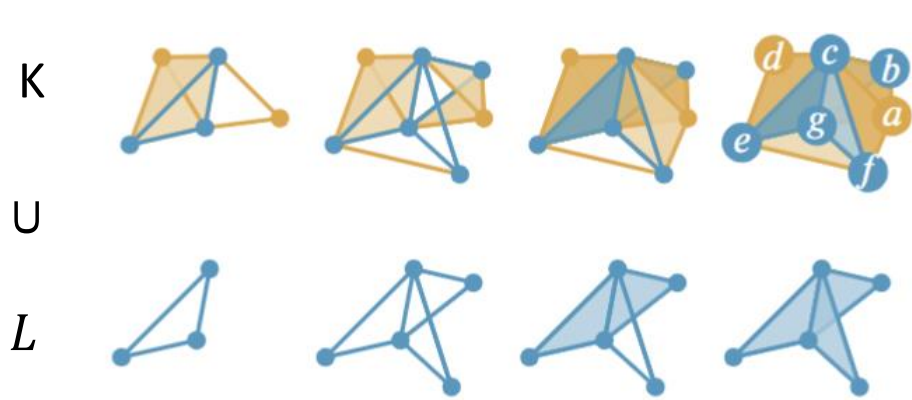
$\{([1, 2], 1)\}$

$\uparrow$

$\{([2, 3], 1), ([2, 2], 1)\}$

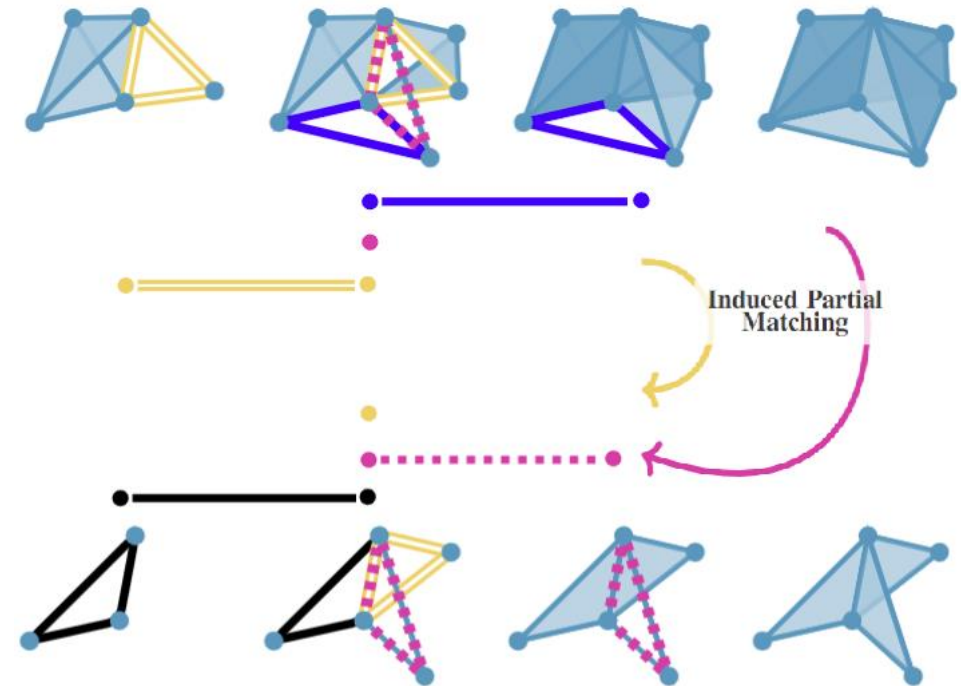
Partial matchings between barcodes

Example:

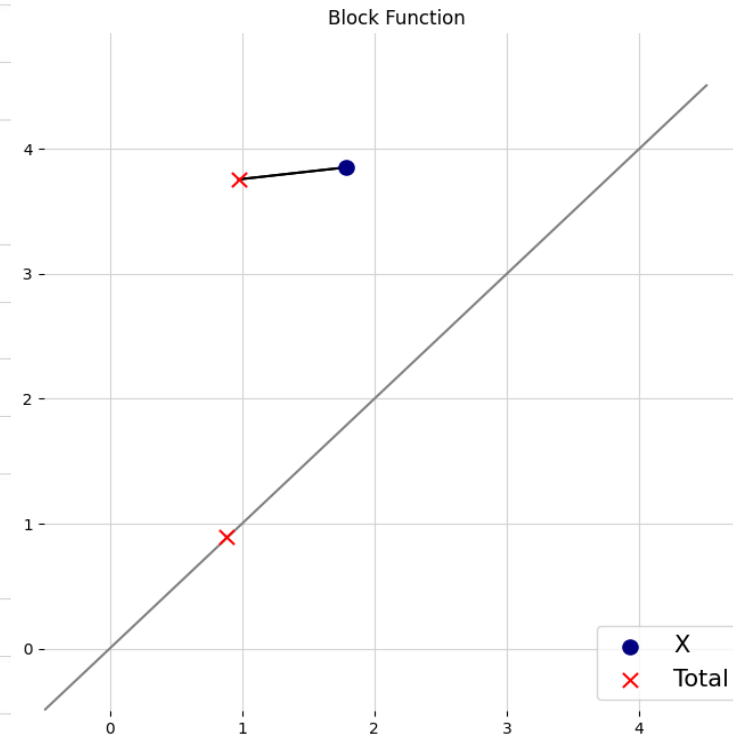
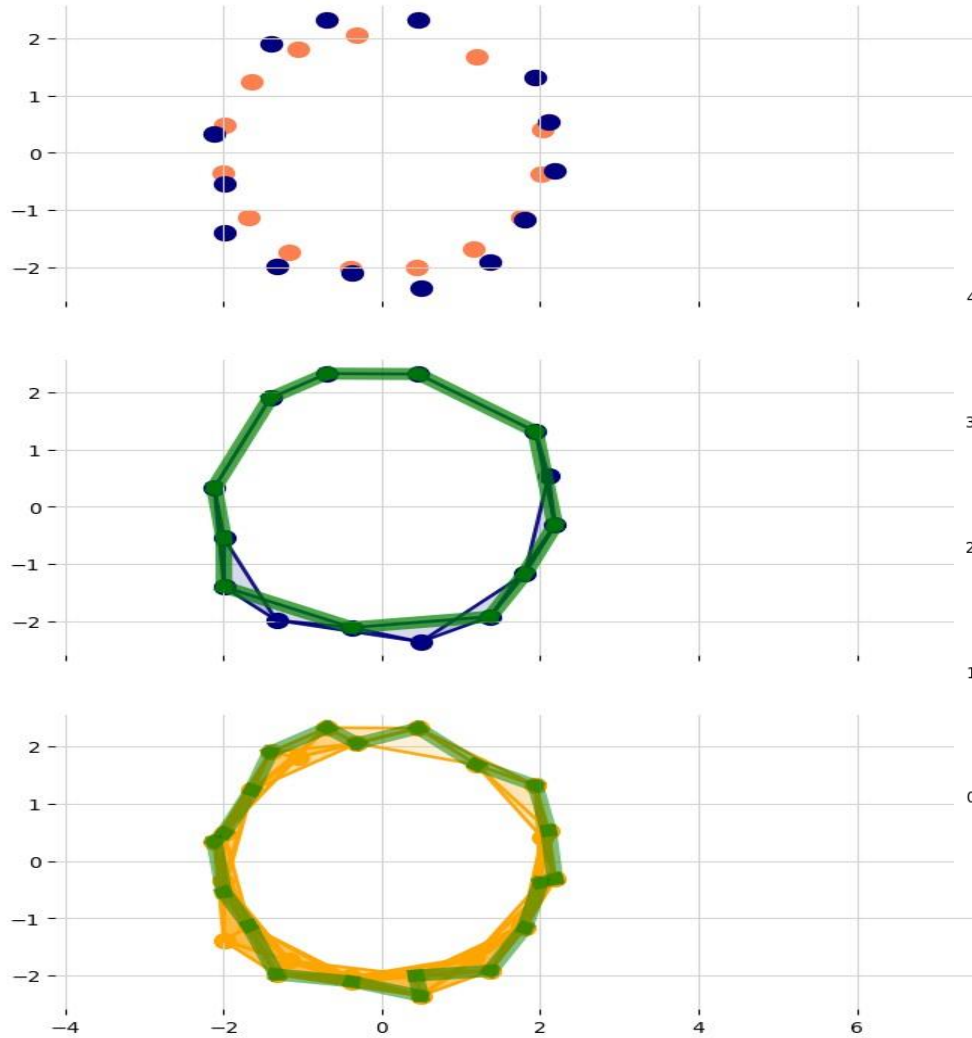


$$PH_1(K) \simeq k_{[1,2]} \oplus k_{[2,2]} \oplus k_{[2,3]}$$

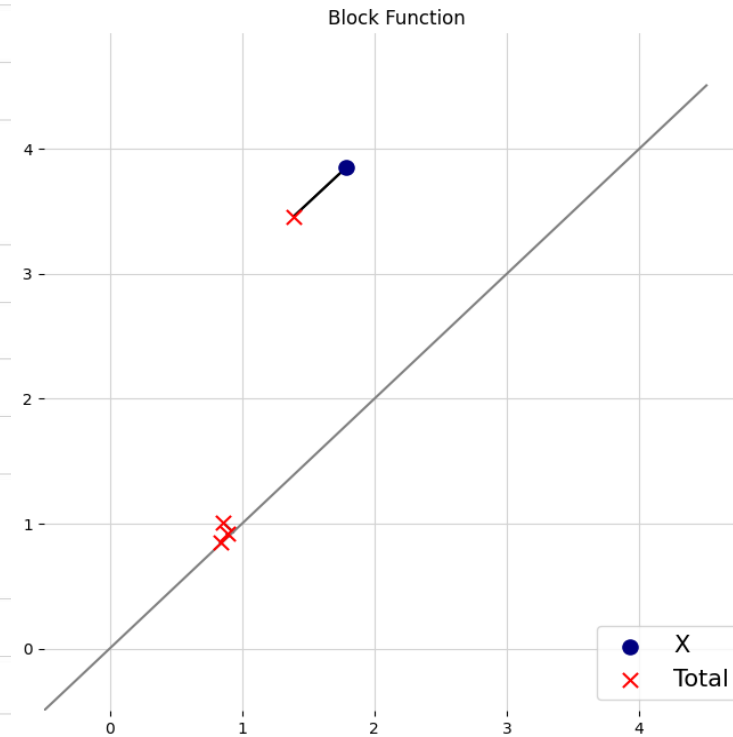
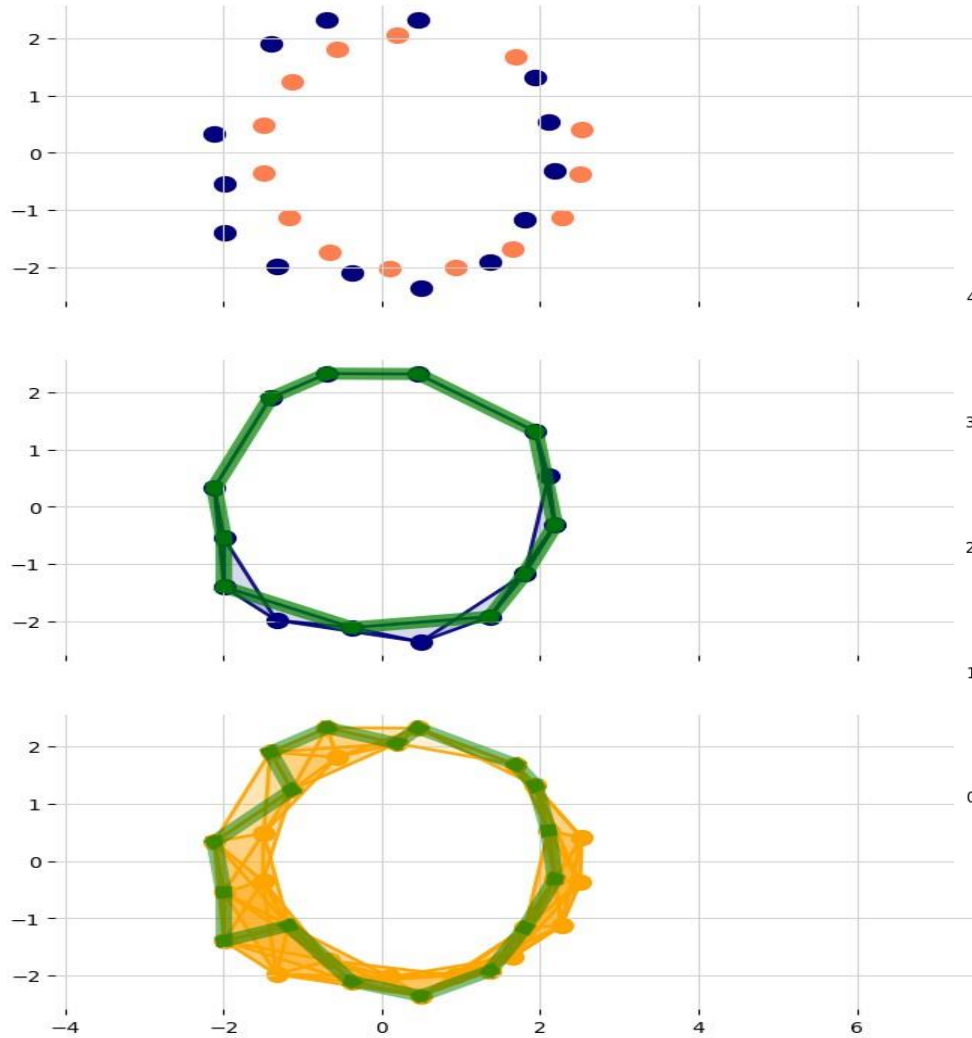
$$PH_1(L) \simeq k_{[1,2]} \oplus k_{[2,2]} \oplus k_{[2,3]}$$



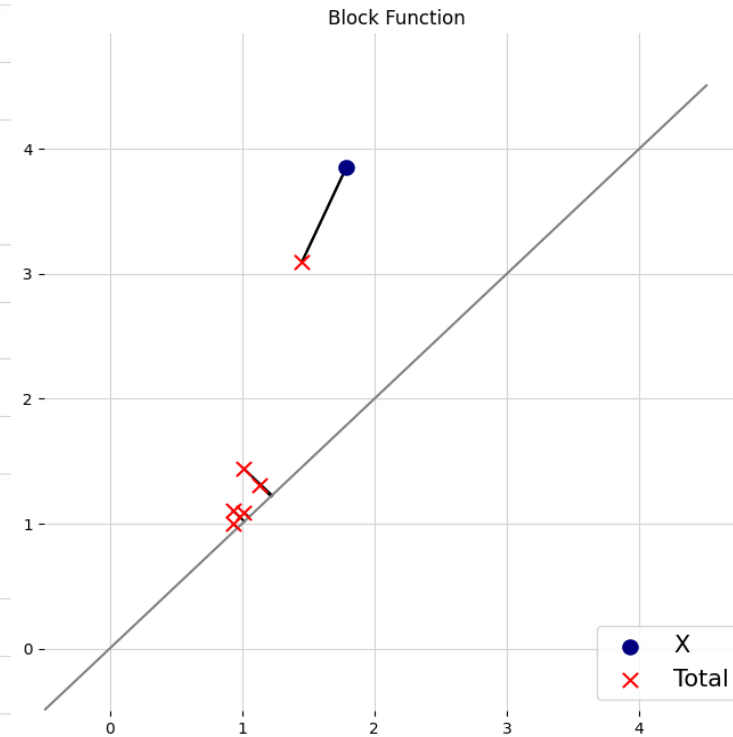
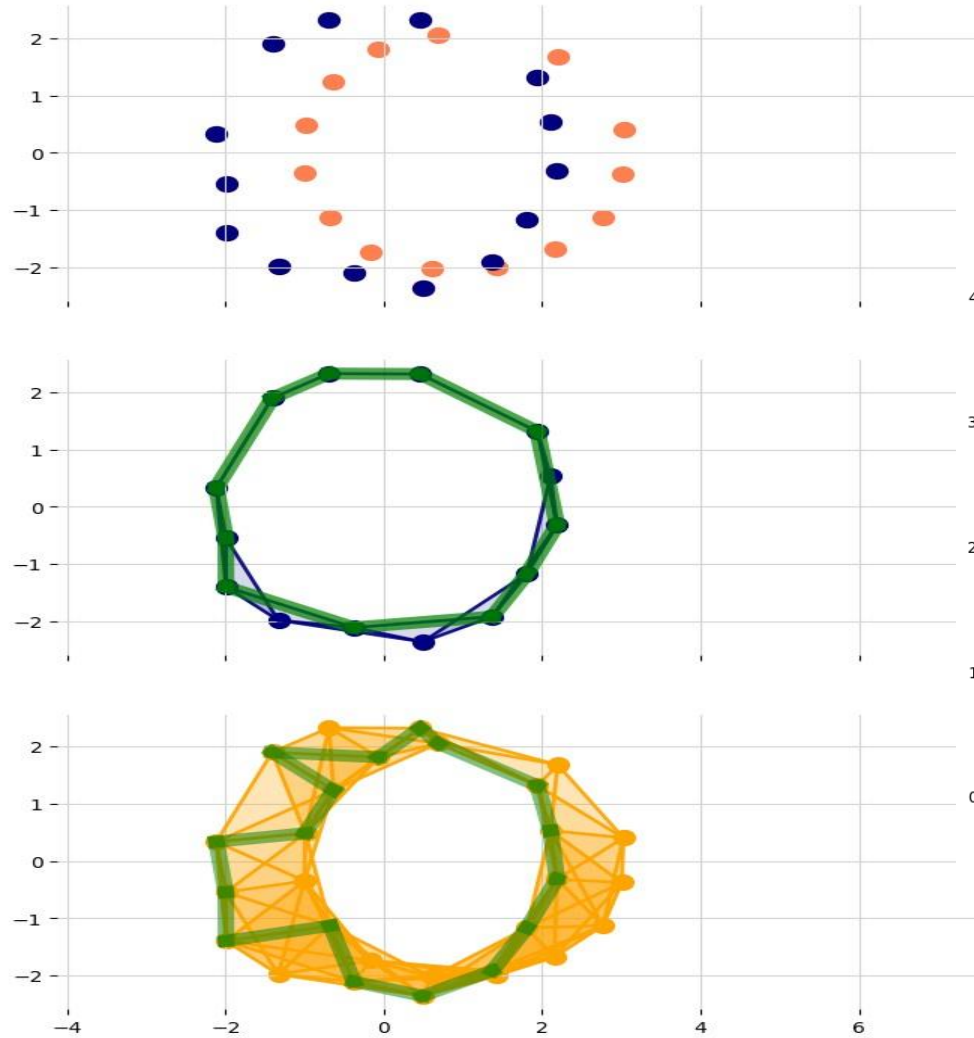
Partial matchings between barcodes



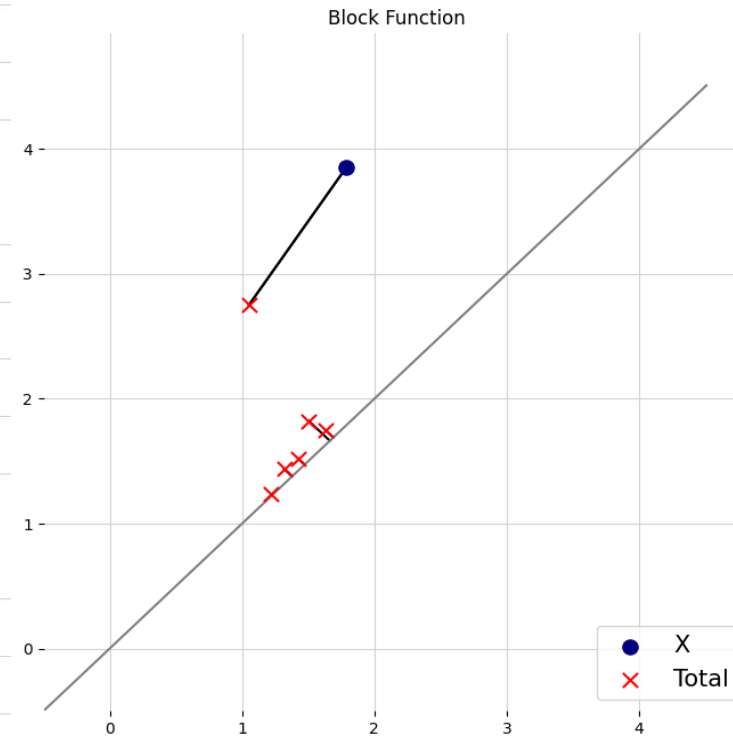
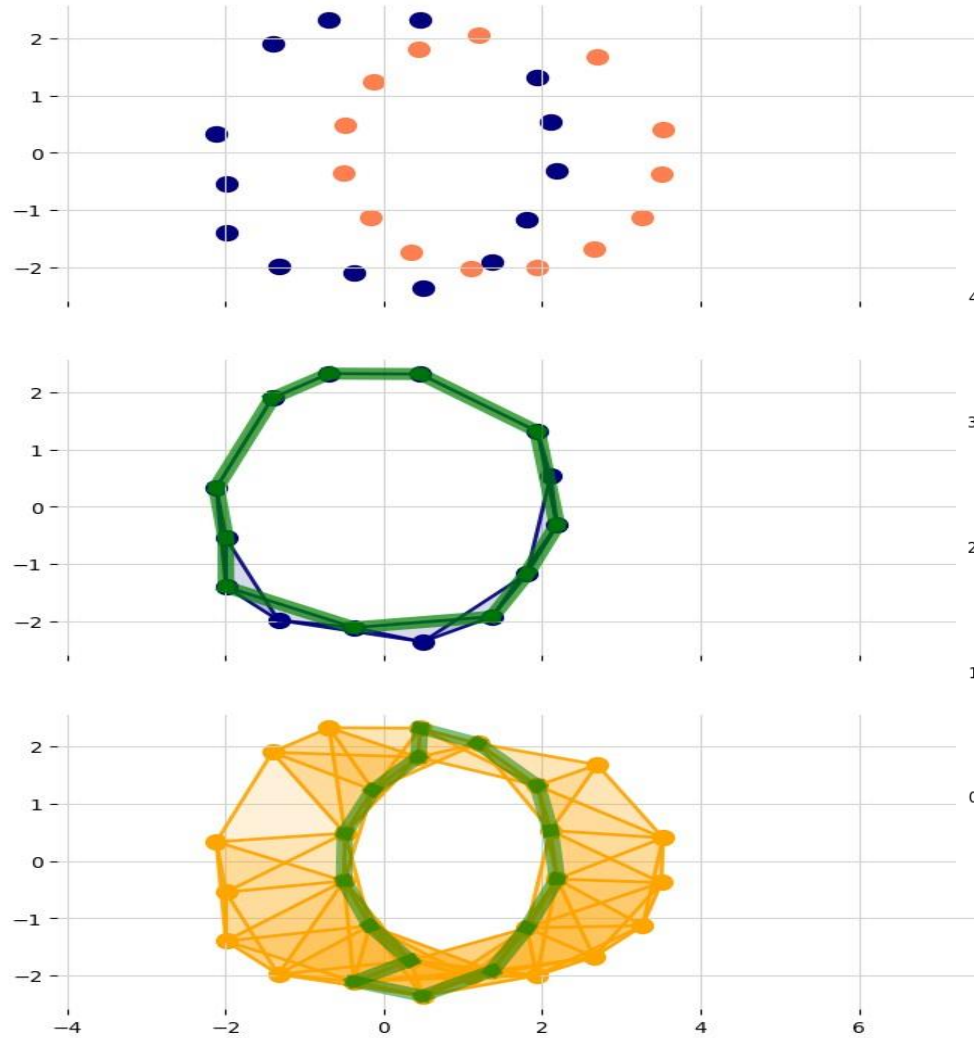
Partial matchings between barcodes



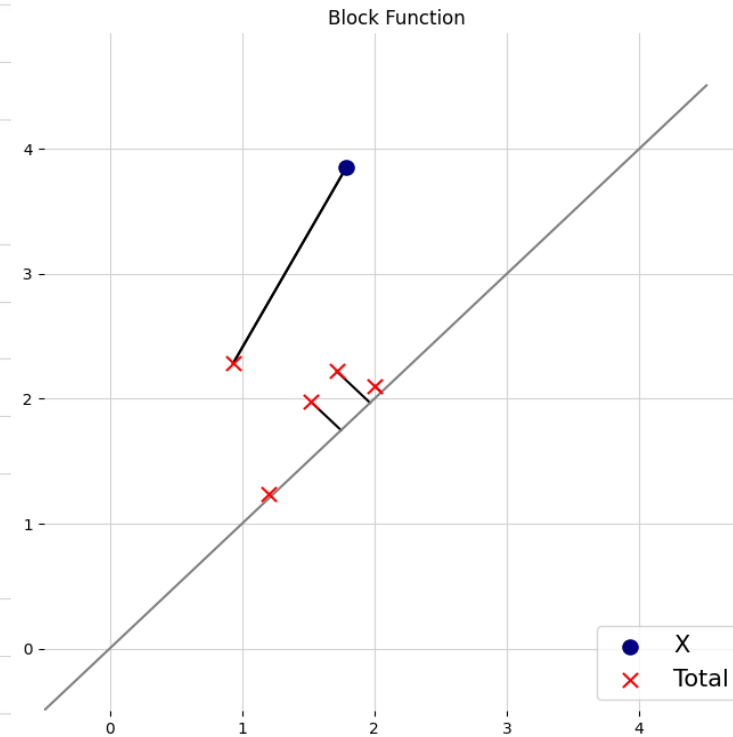
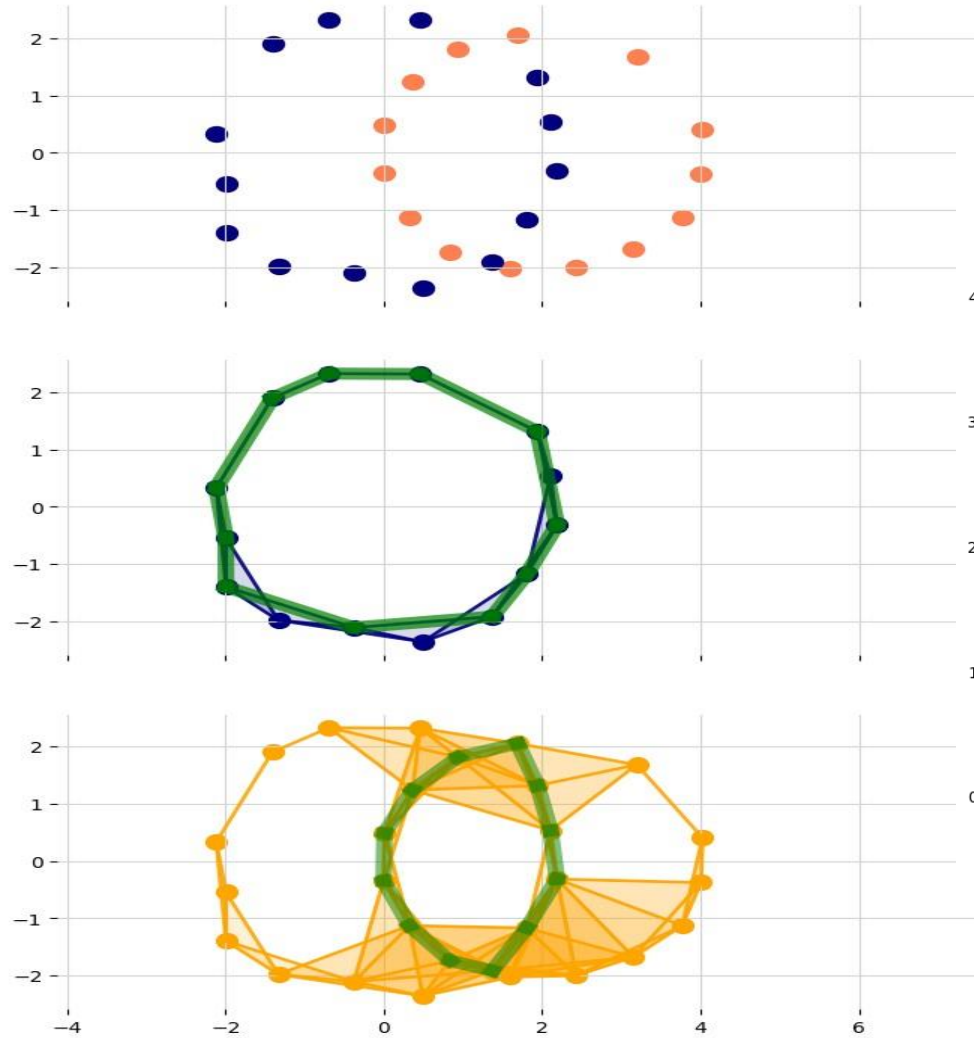
Partial matchings between barcodes



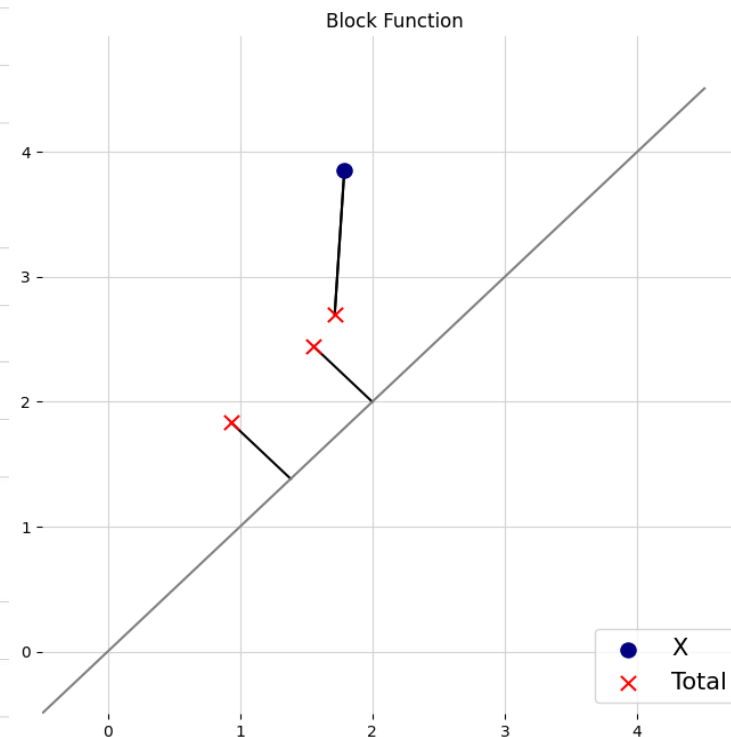
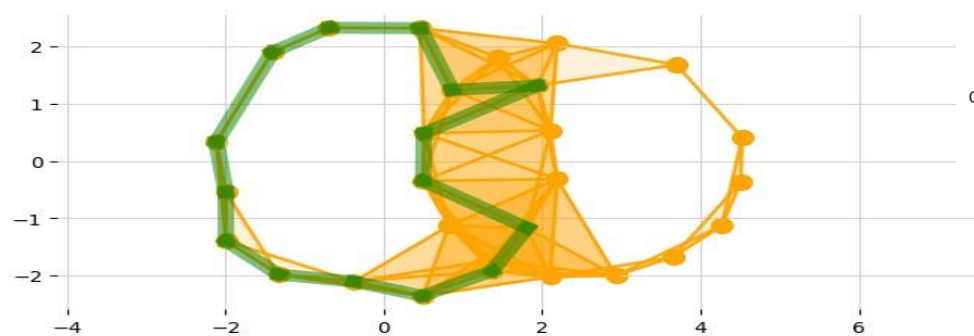
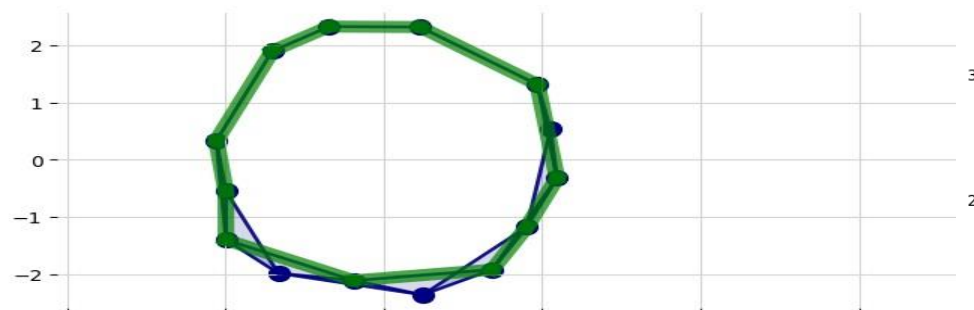
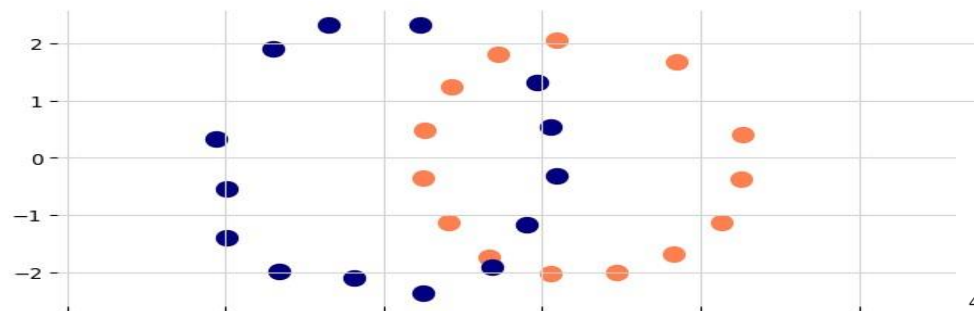
Partial matchings between barcodes



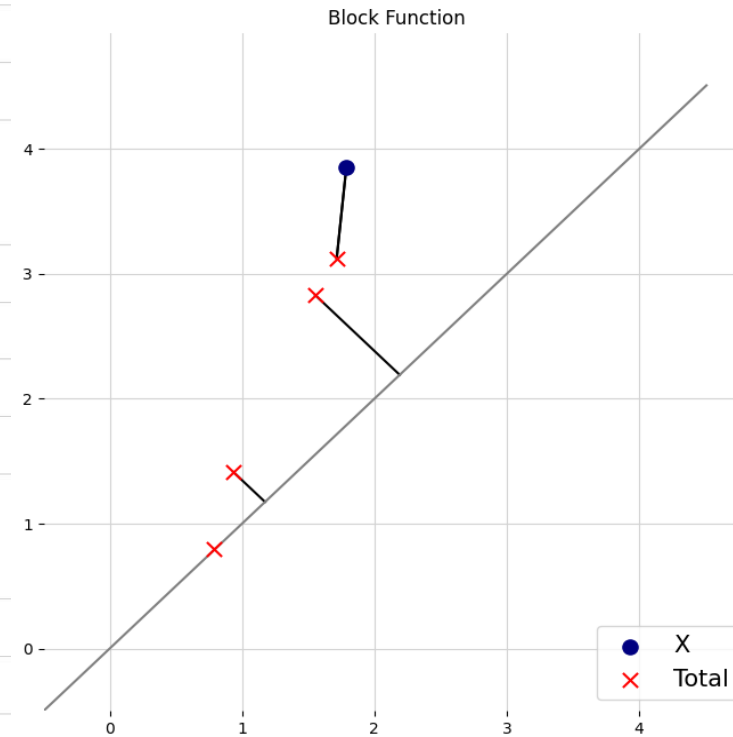
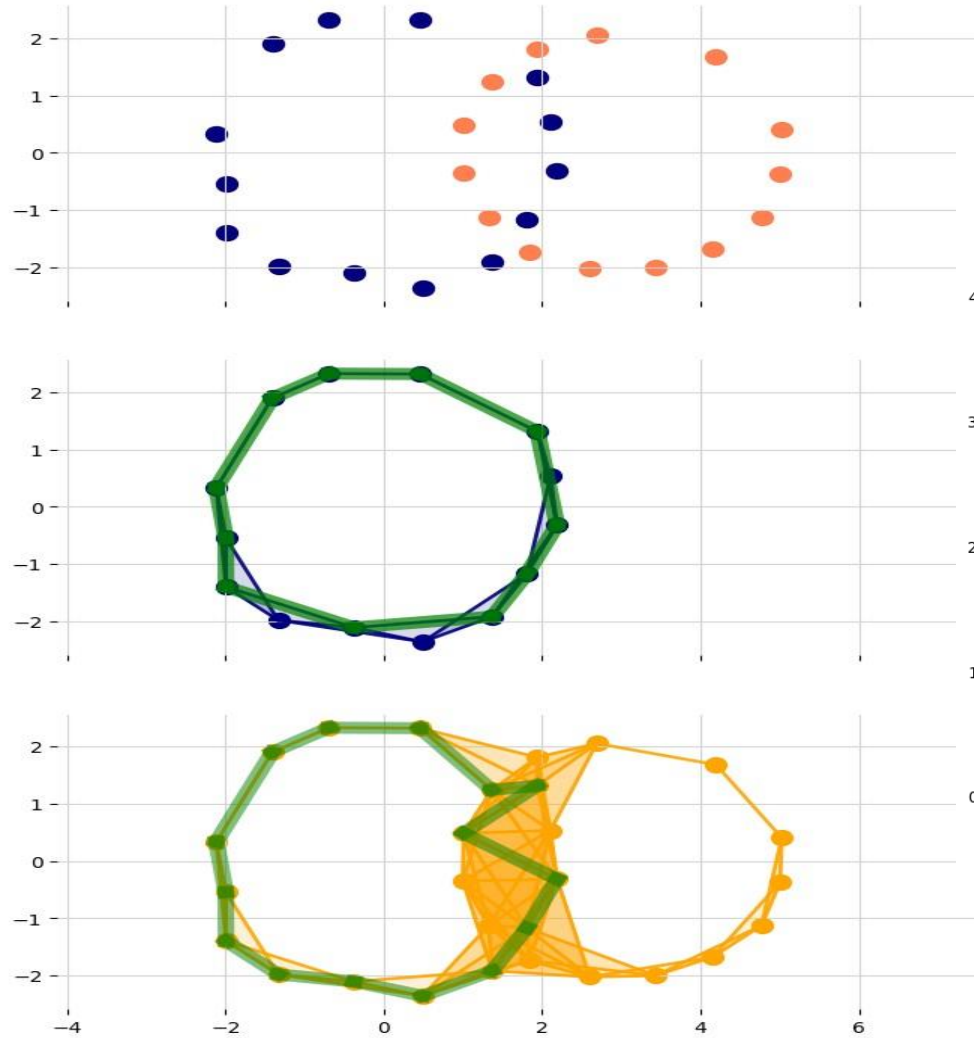
Partial matchings between barcodes



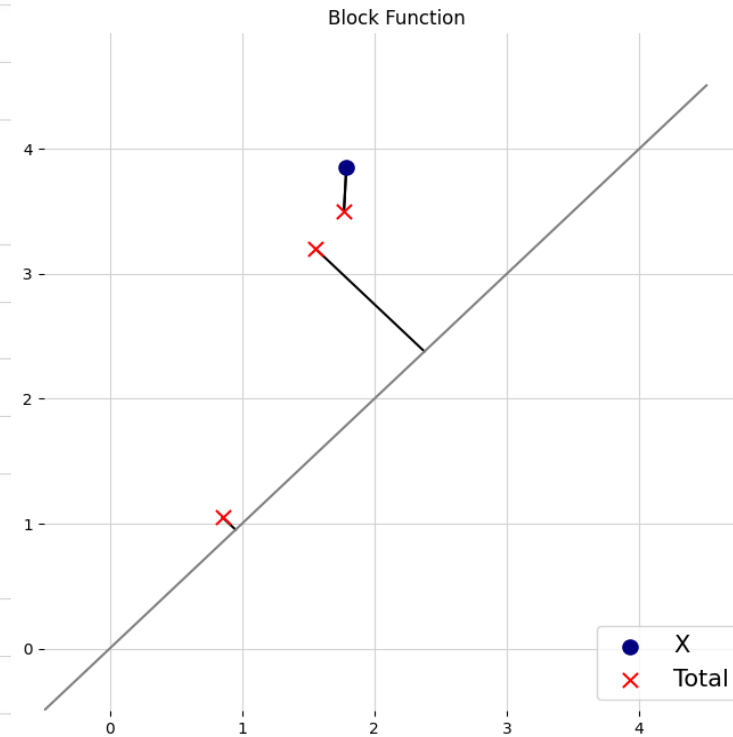
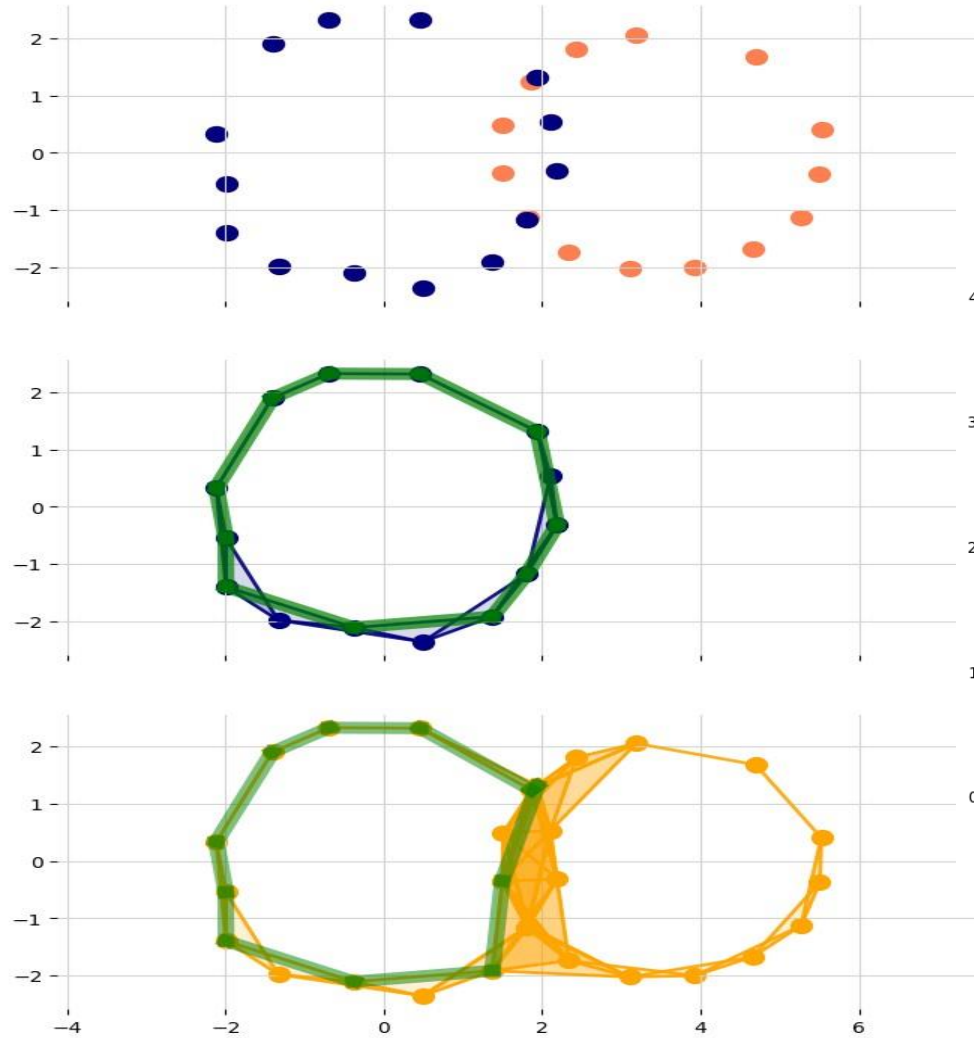
Partial matchings between barcodes



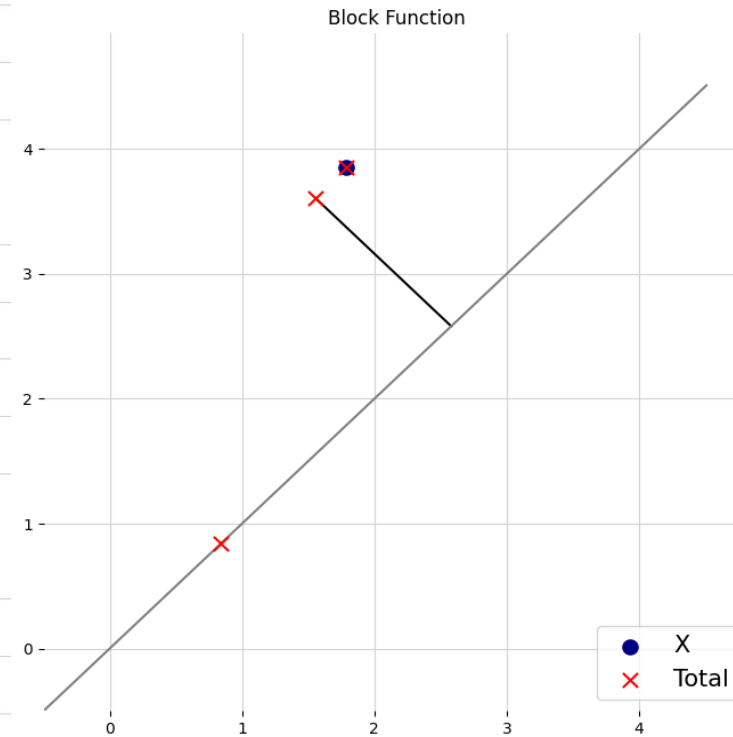
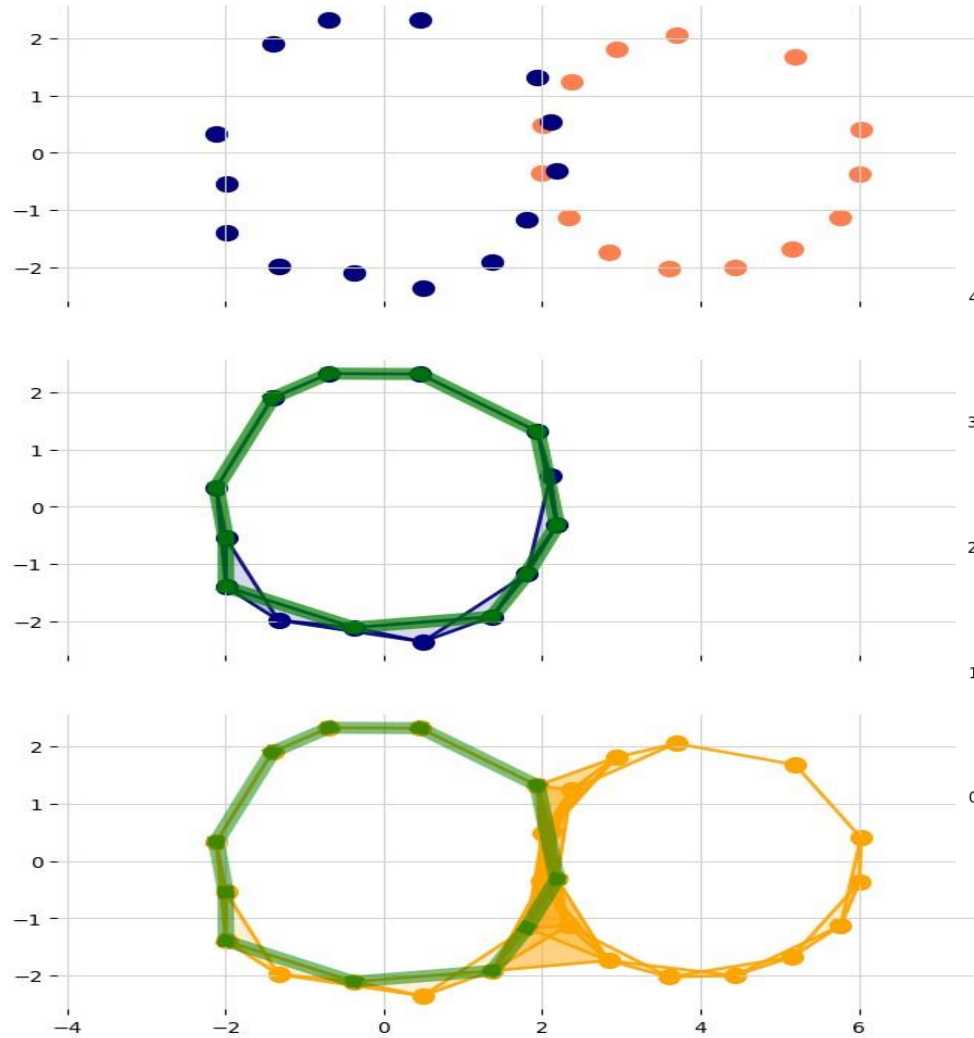
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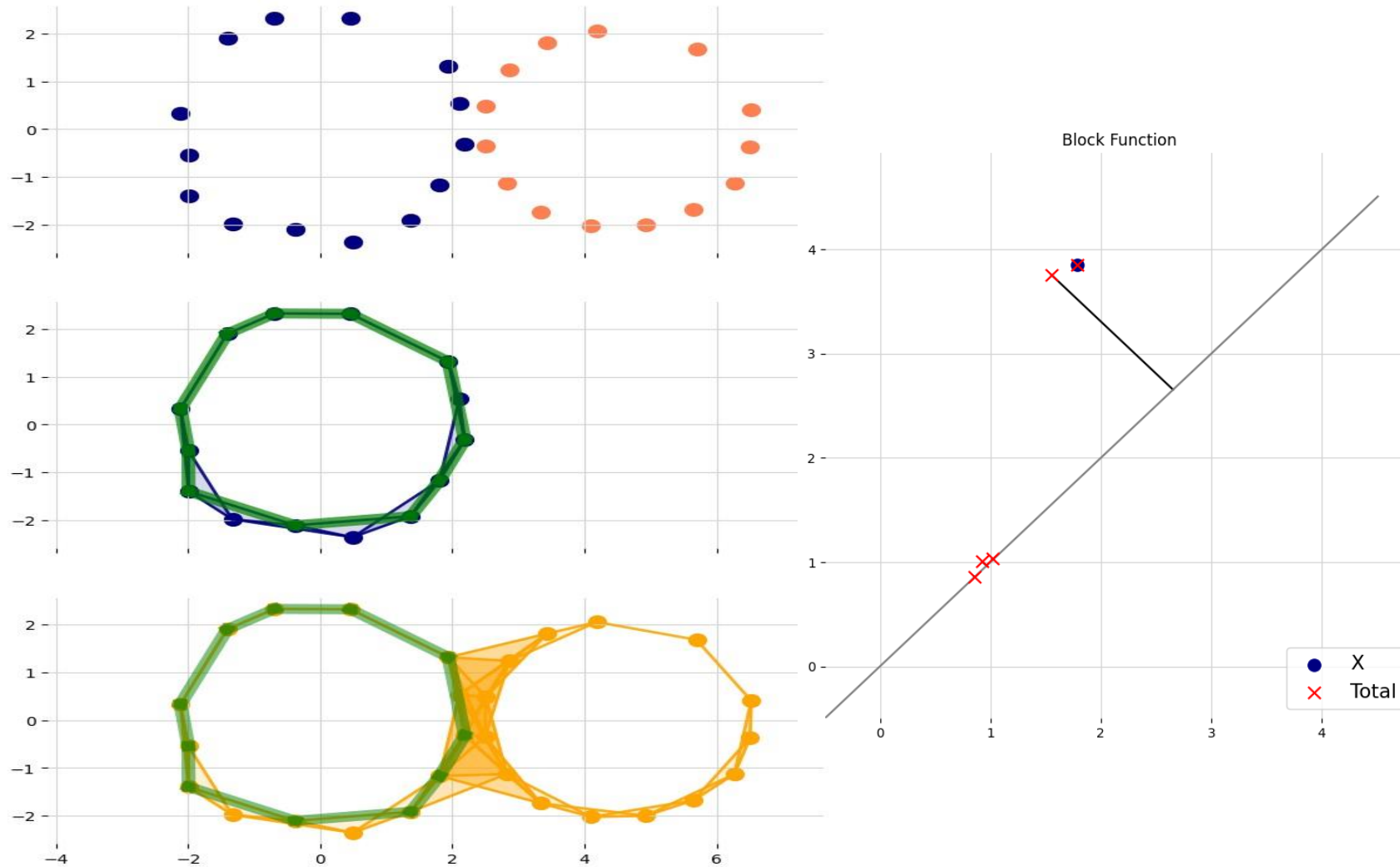
Partial matchings between barcodes



Partial matchings between barcodes



Partial matchings between barcodes



Partial matchings between barcodes

Matrix
computation:

$$U \simeq k_{[1,2]} \oplus k_{[1,2]} \oplus k_{[2,3]}$$

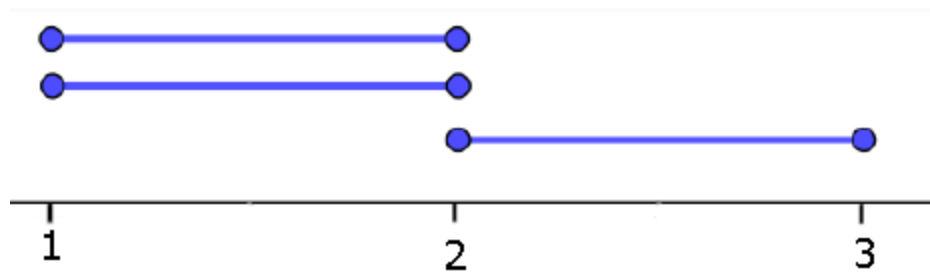


$$\mathbf{B}(U) = \{([1, 2], 2), ([2, 3], 1)\}$$

$$S_U = \{[1,2], [2,3]\}$$

Endpoint order: $[a_1, b_1] \leq [a_2, b_2]$ iff $b_1 < b_2$ or $b_1 = b_2$ and $a_1 \leq a_2$

$$\text{Rep } B(U) = \langle [1,2], [1,2], [2,3] \rangle$$

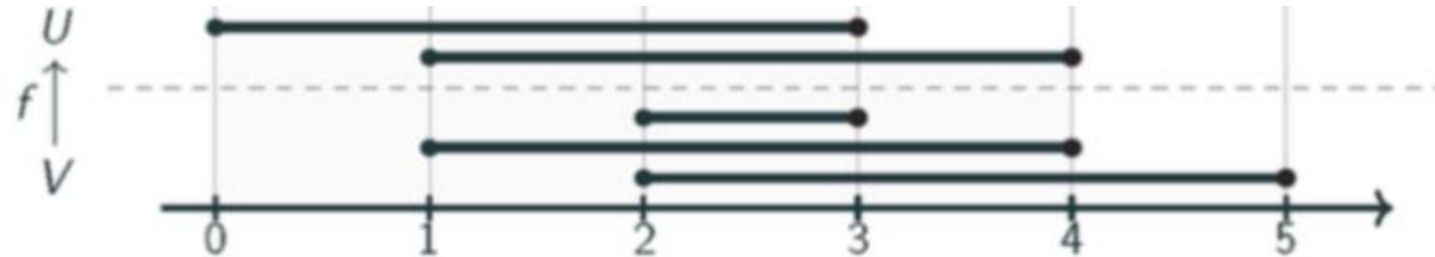


Partial matchings between barcodes

Endpoint order: $[a_1, b_1] \leq [a_2, b_2]$ iff $b_1 < b_2$ or $b_1 = b_2$ and $a_1 \leq a_2$

Matrix
computation:

- Consider $V \simeq k_{[2,3]} \oplus k_{[1,4]} \oplus k_{[2,5]}$ and $U \simeq k_{[0,3]} \oplus k_{[1,4]}$.
- Order the intervals in $\text{Rep } B(V)$ and $\text{Rep } B(U)$ following the endpoint order:



- Suppose f associated to the following matrix:

$$F = \left[\begin{array}{c|ccc} & [2, 3] & [1, 4] & [2, 5] \\ \hline [0, 3] & 1 & 1 & 0 \\ [1, 4] & 0 & 1 & 1 \end{array} \right]$$

- Let $I = [a, b]$. Consider F_I , the reduced minor of F restricted to $[c, d] \in \text{Rep } B(U)$ with $c \leq a$ and $d \leq b$ and $[a_1, b_1] \in \text{Rep } B(V)$ with $b_1 < b$ or $b_1 = b$ and $a_1 \leq a$.

$$F_{[2,3]} = \begin{bmatrix} 1 \\ \end{bmatrix}, F_{[1,4]} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \text{ and } F_{[2,5]} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

- $\mathcal{M}_f$ is given by $[2, 3] \mapsto [0, 3]$ and $[1, 4] \mapsto [1, 4]$ and $[2, 5] \mapsto \emptyset$.

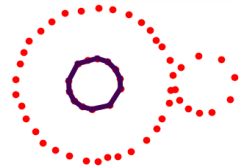
Partial matchings between barcodes

Example:

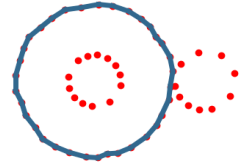
Subset 1 (S_1)



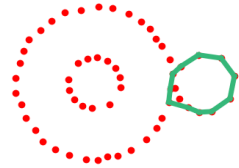
— [0.6, 1.3]



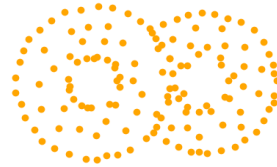
— [0.5, 1.5]



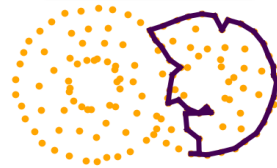
— [0.6, 1.5]



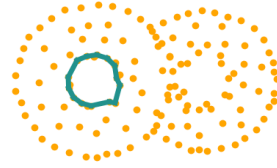
Total set (T)



— [0.4, 1.2]



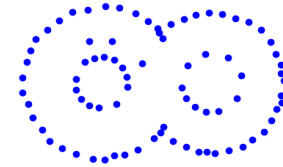
— [0.5, 1.2]



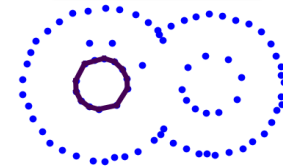
$$S_1 \rightarrow T \quad S_2 \rightarrow T$$

$$\begin{bmatrix} 0 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 1 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

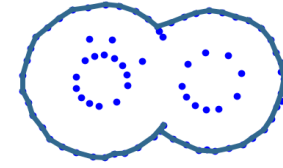
Subset 2 (S_2)



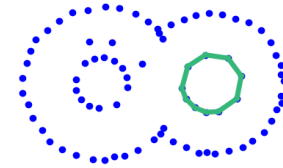
— [0.6, 1.3]



— [0.5, 1.5]



— [0.6, 1.5]



Partial matchings between barcodes

Endpoint order: $[a_1, b_1] \leq [a_2, b_2]$ iff $b_1 < b_2$ or $b_1 = b_2$ and $a_1 \leq a_2$

Example:

- Sort both $Rep B(S_1)$ and $Rep B(T)$ by the endpoint order
- We have the matrix

$$F = \begin{array}{c|ccc} & [0.6, 1.3] & [0.5, 1.5] & [0.6, 1.5] \\ \hline [0.4, 1.2] & 0 & 0 & 1 \\ [0.5, 1.2] & 1 & 1 & 0 \end{array}$$

- We obtain the matrices

$$F_{[0.6, 1.3]} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, F_{[0.5, 1.5]} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, F_{[0.6, 1.5]} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix},$$

- Assignment: $[0.6, 1.3] \mapsto [0.5, 1.2]$, $[0.5, 1.5] \mapsto [0.5, 1.2]$ and $[0.6, 1.5] \mapsto [0.4, 1.2]$.

Partial matchings between barcodes

Endpoint order: $[a_1, b_1] \leq [a_2, b_2]$ iff $b_1 < b_2$ or $b_1 = b_2$ and $a_1 \leq a_2$

Example:

- Sort both $Rep B(S_2)$ and $Rep B(T)$ by the endpoint order
- We have the matrix

$$F = \left[\begin{array}{c|ccc} & [0.6, 1.3] & [0.5, 1.5] & [0.6, 1.5] \\ \hline [0.4, 1.2] & 0 & 1 & 1 \\ [0.5, 1.2] & 1 & 1 & 0 \end{array} \right]$$

- We obtain the matrices

$$F_{[0.6, 1.3]} = \begin{bmatrix} 0 \\ \mathbf{1} \end{bmatrix}, F_{[0.5, 1.5]} = \begin{bmatrix} 1 \\ \mathbf{1} \end{bmatrix}, F_{[0.6, 1.5]} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

- Assignment: $[0.6, 1.3] \mapsto [0.5, 1.2]$ and $[0.5, 1.5] \mapsto [0.5, 1.2]$.

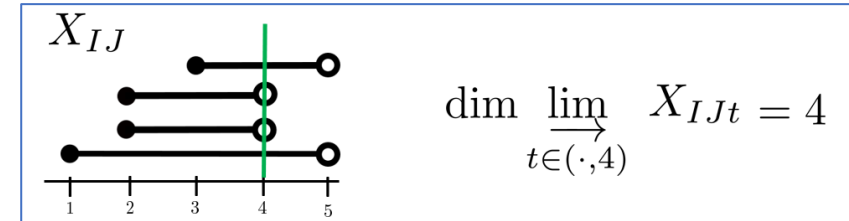
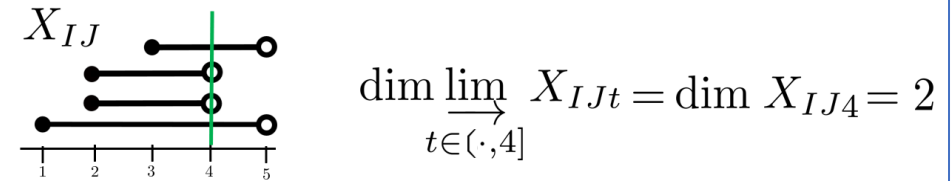
Partial matchings between barcodes

$$\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0}$$

$$\sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$$

$$\mathcal{M}_f(I, J) = \dim \varinjlim_{t \in I \cap J} X_{IJt}$$

Proof: Decorated points and persistence modules:



We will represent the intervals using pairs, (a, b) , of elements of

$$\mathbf{E} = (\mathbb{R} \times \{-, +\}) \cup \{-\infty, +\infty\}$$

$a \backslash b$	s^-	s^+	∞
$-\infty$	$(-\infty, s)$	$(-\infty, s]$	$(-\infty, \infty)$
r^-	$[r, s)$	$[r, s]$	$[r, \infty)$
r^+	(r, s)	$(r, s]$	(r, ∞)

Partial matchings between barcodes

$$\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0} \qquad \sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$$

$$\mathcal{M}_f(I, J) = \dim \varinjlim_{t \in I \cap J} X_{IJt}$$

Proof:

A **section** of a vector space, $\mathcal{V}$, is a pair of vector spaces, (F^-, F^+) , such that

$$F^- \hookrightarrow F^+ \hookrightarrow \mathcal{V}$$

We say that a set of sections, $\{(F_\lambda^-, F_\lambda^+) : \lambda \in \Lambda\}$, of $\mathcal{V}$ is **disjoint** if, for all $\lambda \neq \mu$,

$$F_\mu^+ \hookrightarrow F_\lambda^- \quad \text{or} \quad F_\lambda^+ \hookrightarrow F_\mu^-$$

Lemma. If $\{(F_\lambda^-, F_\lambda^+) : \lambda \in \Lambda\}$ is a set of disjoint sections of $\mathcal{V}$, we have that

$$\bigoplus_{\lambda \in \Lambda} (F_\lambda^+ / F_\lambda^-) \hookrightarrow \mathcal{V}$$

Partial matchings between barcodes

$$\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0}$$

$$\sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$$

$$\mathcal{M}_f(I, J) = \dim \varinjlim_{t \in I \cap J} X_{IJt}$$

Proof:

$$\dim V_{(s,r)t}^+$$

nº of intervals (a, b) such that $t \in (a, b)$,
 $a \leq s$ and $b \leq r$

$$\dim V_{(s,r)t}^-$$

nº of intervals (a, b) such that $t \in (a, b)$,
 $(a < s$ and $b \leq r)$ or $(a \leq s$ and $b < r)$

$V_{(s,r)t}^-$ and $V_{(s,r)t}^+$ are persistence submodules of V and , for each t ,
 $\left\{ \left(V_{(s,r)t}^-, V_{(s,r)t}^+ \right) \right\}_{s < r}$ are disjoint sections of V_t

$$\bigoplus_{s < r} \left(V_{(s,r)t}^+ / V_{(s,r)t}^- \right) \hookrightarrow V_t$$

Partial matchings between barcodes

$$\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0}$$

$$\sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$$

$$\mathcal{M}_f(I, J) = \dim \varinjlim_{t \in I \cap J} X_{IJt}$$

Proof:

fixed ↘
 $I = (a, b) \in S_V$
↙ variable
 $J = (c, d) \in S_U$

$$A_{ct}^d := \frac{fV_{It}^- \cap U_{(c,d)t}^+ + fV_{It}^+ \cap U_{(c,d)t}^-}{fV_{It}^- \cap U_{(c,d)t}^+} \quad \text{and} \quad B_{ct}^d := \frac{fV_{It}^+ \cap U_{(c,d)t}^+}{fV_{It}^- \cap U_{(c,d)t}^+}$$

$$\frac{B_{ct}^d}{A_{ct}^d} \simeq X_{IJt}$$

Partial matchings between barcodes

$$\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0} \qquad \sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$$

$$\mathcal{M}_f(I, J) = \dim \varinjlim_{t \in I \cap J} X_{IJt}$$

Proof:

$$\tilde{A}_c^d := \varinjlim_{t \in I \cap J} A_{ct}^d$$

$$\tilde{B}_c^d := \varinjlim_{t \in I \cap J} B_{ct}^d$$

- For fixed d and variable c , they are persistence modules indexed by $\mathbf{E}$
- They satisfies that

$$\frac{\tilde{B}_c^d}{\tilde{A}_c^d} \simeq \varinjlim_{t \in I \cap J} X_{IJt}$$

Partial matchings between barcodes

$$\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0} \qquad \sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$$

$$\mathcal{M}_f(I, J) = \dim \varinjlim_{t \in I \cap J} X_{IJt}$$

Proof:

For each d ,

$\{(\tilde{A}_c^d, \tilde{B}_c^d) : c \in \mathbf{E}\}$ is a disjoint set of sections of $\varinjlim_{c < d} \tilde{B}_c^d$.

$$\left(\underbrace{\bigoplus_{c < d} \varinjlim_{t \in I \cap J} X_{I(c,d)t}}_{\frac{\tilde{B}_c^d}{\tilde{A}_c^d}} \right) \hookrightarrow \varinjlim_{c < d} \tilde{B}_c^d \hookrightarrow \varinjlim_{t \in (-\infty, d)} (fV_{It}^+ / fV_{It}^-)$$

Partial matchings between barcodes

$$\mathcal{M}_f : S_V \times S_U \rightarrow \mathbb{Z}_{\geq 0}$$

$$\sum_{J \in S_U} \mathcal{M}_f(I, J) \leq m_I$$

$$\mathcal{M}_f(I, J) = \dim \varinjlim_{t \in I \cap J} X_{IJt}$$

Proof:

$$\begin{aligned} & \sum_{J \in S_U} \mathcal{M}_f(I, J) \\ & \quad \parallel \\ & \sum_{d \leq b} \dim \left(\bigoplus_{c < d} \varinjlim_{t \in (-\infty, d)} X_{I(c, d)t} \right) \\ & \quad \left(\bigoplus_{c < d} \varinjlim_{t \in I \cap J} X_{I(c, d)t} \right) \hookrightarrow \varinjlim_{c < d} \tilde{B}_c^d \\ & \quad \varinjlim_{c < d} \tilde{B}_c^d \hookrightarrow \varinjlim_{t \in (-\infty, d)} (fV_{It}^+ / fV_{It}^-) \\ & \quad \leq \sum_{d \leq b} \dim \varinjlim_{c < d} \tilde{B}_c^d \\ & \quad \leq \# \text{ intervals in } fV_{It}^+ / fV_{It}^- \\ & \quad \leq \# \text{ intervals in } V_{It}^+ / V_{It}^- = m_I \end{aligned}$$

□

Links

<https://doi.org/10.1016/j.comgeo.2023.101985>

<https://arxiv.org/abs/2306.02411>

Decomposition of pointwise finite-dimensional persistence modules. W. Crawley-Boevey.

Topology Tools for Explainable and Green Artificial Intelligence

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- Context: Green and Explainable artificial intelligence (REXASI-PRO)
- Computational topology tools: Persistent homology, barcodes, distance bottleneck, simplicial maps, Persistence modules, morphisms between persistence modules
- Partial matchings between barcodes
- Simplicial maps neural networks