

Dataset comparison using persistent homology morphisms

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Outline and scope of the talk:

- **Review:** Vietoris-Rips filtration and Persistent Homology
- **Example:** Morphisms between Vietoris-Rips filtrations.
- Motivation for induced partial matchings/Block Functions.
- **Review:** of Bauer-Lesnick matching
- Quick introduction to the induced block function.
- Explore **examples** of point-clouds embedded in \mathbb{R}^2 .

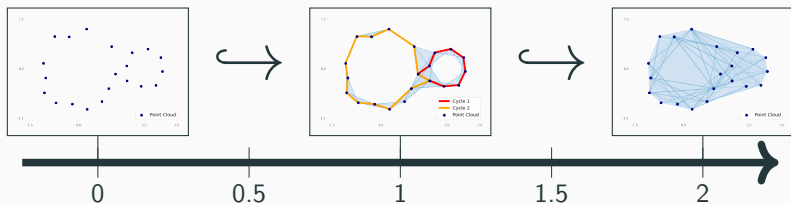
Filtered Complexes: Vietoris-Rips filtration

- Consider a point sample $\mathbb{X} \subseteq \mathbb{R}^n$.
- Let $r \geq 0$, $\text{VR}_r(\mathbb{X})$ is the maximal **simplicial complex** with edges

$$[x, y] \in \text{VR}_r(\mathbb{X}) \iff \|x - y\|_n \leq 2r .$$

- Given a sequence $a_0 < a_1 < \dots < a_n$ from \mathbb{R} , there are inclusions

$$\text{VR}_{a_0}(\mathbb{X}) \hookrightarrow \text{VR}_{a_1}(\mathbb{X}) \hookrightarrow \dots \hookrightarrow \text{VR}_{a_n}(\mathbb{X})$$

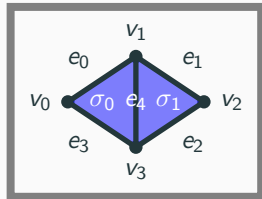
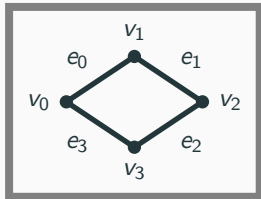
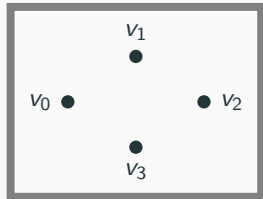


- **Category \mathbf{R}** : objects $a \in \mathbb{R}$, arrows $a \rightarrow b$ iff $a \leq b$
- **Filtered Complex** : $\text{VR}(\mathbb{X}) : \mathbf{R} \rightarrow \text{SpCpx}$

Computation of Persistence Barcode

- Pick up a maximum radius $R > 0$
- Given $\sigma \in \text{VR}_R(\mathbb{X})$, define $\text{filt}(\sigma) = \max \{ \|x - y\|_n / 2 \mid x, y \in \sigma \}$.
- Given $D \in \mathbb{Z}_{\geq 0}$, Consider $\text{VR}_R^D(\mathbb{X})$, the D -skeleton given by simplices $\sigma \in \text{VR}_R(\mathbb{X})$ such that $\dim(\sigma) \leq D$.
- Sort simplices from $\text{VR}_R^D(\mathbb{X})$ by increasing filtration values and dimension, i.e. $\sigma_1 \leq \sigma_2 \Rightarrow \text{filt}(\sigma_1) \leq \text{filt}(\sigma_2)$ and $\dim(\sigma_1) \leq \dim(\sigma_2)$.
- Choose a field k ; e.g. \mathbb{Z}_{11}
- Perform a Gaussian elimination on the boundary matrix of $\text{VR}_R^D(\mathbb{X})$.
- We obtain the **persistence barcode** and **representatives**.

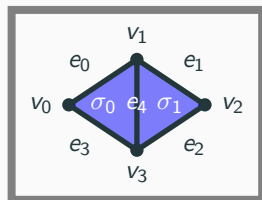
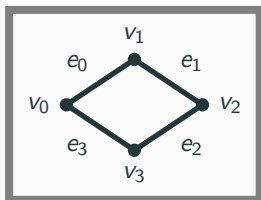
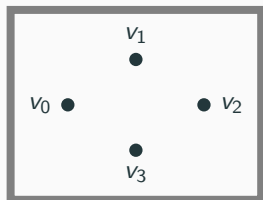
Example of Computation of Persistent Homology



Boundary Matrix:

$$\begin{pmatrix} & e_0 & e_1 & e_2 & e_3 & e_4 & \sigma_0 & \sigma_1 \\ v_0 & -1 & 0 & 0 & -1 & 0 & 0 & 0 \\ v_1 & 1 & -1 & 0 & 0 & -1 & 0 & 0 \\ v_2 & 0 & 1 & -1 & 0 & 0 & 0 & 0 \\ v_3 & 0 & 0 & 1 & 1 & 1 & 0 & 0 \\ e_0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ e_1 & 0 & 0 & 0 & 0 & 0 & 0 & -1 \\ e_2 & 0 & 0 & 0 & 0 & 0 & 0 & -1 \\ e_3 & 0 & 0 & 0 & 0 & 0 & -1 & 0 \\ e_4 & 0 & 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix} \rightarrow \left\{ \begin{array}{c} \text{Persistence Pairs} \\ ? \end{array} \right\}$$

Example of Computation of Persistent Homology



Reduced Boundary Matrix: obtain persistence pairs and representatives:

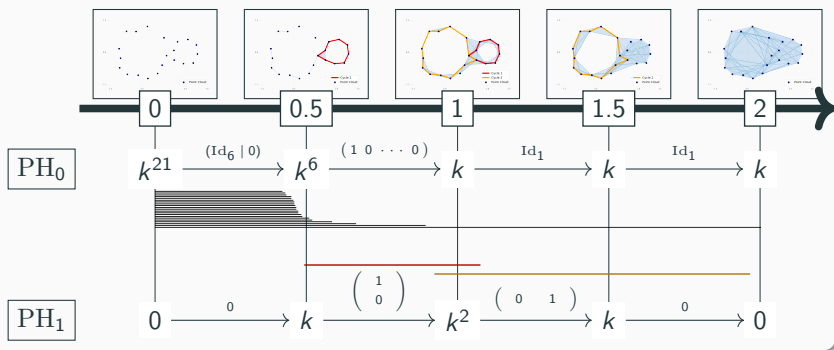
	e_0	e_1	e_2	e_3	e_4	σ_0	σ_1
v_0	-1	0	0	0	0	0	0
v_1	1	-1	0	0	0	0	0
v_2	0	1	-1	0	0	0	0
v_3	0	0	1	0	0	0	0
e_0	0	0	0	0	0	1	-1
e_1	0	0	0	0	0	0	-1
e_2	0	0	0	0	0	0	-1
e_3	0	0	0	0	0	-1	1
e_4	0	0	0	0	0	1	0

→ { Persistence Pairs }
 (v_1, e_0)
 (v_2, e_1)
 (v_3, e_2)
 (e_4, σ_0)
 (e_3, σ_1)
Convention:
 (positive, negative)

Example: Interval decomposition

- For each pair (τ, σ) , we obtain an interval $I = [\text{filt}(\tau), \text{filt}(\sigma))$.
- I is nontrivial iff $\text{filt}(\tau) < \text{filt}(\sigma)$
- $\text{filt}(\tau)$ is the **birth value** and $\text{filt}(\sigma)$ is the **death value** of I .

Example



- **Homology:** H_0 “connected components”, H_1 “holes”, etc.
- **Persistent Homology :** $\text{PH}_n(\mathbb{X}) := H_n(\text{VR}(\mathbb{X}); k) : \mathbf{R} \rightarrow \mathbf{Vect}_k$

Persistence Modules and Morphisms

- **Persistence Module:** a functor $V : \mathbf{R} \rightarrow \mathbf{Vect}_k$. Sometimes written as a pair (V, ρ) where ρ are the **structure maps** $\rho_{st} : V_s \rightarrow V_t$ for all $s < t$.
- **Morphism between Persistence Modules:** Given persistence modules (V, ρ) and (U, τ) , then $f : V \rightarrow U$ is a set of linear maps $f_t : V_t \rightarrow U_t$ for all $t \in \mathbf{R}$ s.t. $\tau_{st} f_s = f_t \rho_{st}$ for all $s < t$.
- **Alternative names:** “Persistence Morphism” or “Ladder Module”.
- **Interval Module:** $k_{[a,b)} : \mathbf{R} \rightarrow \mathbf{Vect}_k$, with

$$k_{[a,b)}(r) = \begin{cases} k, & \text{for } r \in [a, b) \\ 0, & \text{otherwise.} \end{cases}$$

A little more about Barcode Decompositions

- Let (V, ρ) be a persistence module.
- If V satisfies the **descending chain condition** for images and kernels then

$$V \simeq \bigoplus_{I \in S_V} (\bigoplus_{m_I} k_I),$$

as proved in¹.

- The **barcode** of V , $\mathbf{B}(V)$, is a **multiset** (S_V, m) where S_V is a set of intervals and $m : S_V \rightarrow \mathbb{Z}_{\geq 0} \cup \{\infty\}$ is the multiplicity of bars.
- The *representation of a multiset* (S, m) is the set

$$\text{Rep}(S, m) = \{(I, i) \in S \times \mathbb{N} : i \leq m_I\}.$$

¹W. Crawley-Boevey *Decomposition of pointwise finite-dimensional persistence modules*, Journal of Algebra and Its Applications, 14 (5) (2015)

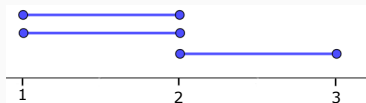
Example of (Representation of) Barcodes

Example

Consider $U : \mathbf{R} \rightarrow \mathbf{Vect}_k$ such that

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

Then its barcode is $\mathbf{B}(U) = \{([1, 2], 2), ([2, 3], 1)\}$ and the representation of its barcode is $\text{Rep } \mathbf{B}(U) = \{[1, 2]_1, [1, 2]_2, [2, 3]_1\}$, which can be displayed as:



Persistence Morphisms

- Let a morphism between persistence modules $f : V \rightarrow U$.
Problem: $f : V \rightarrow U$ has indecomposables of wild type²; i.e. there is no “barcode” for f .
Idea: Use the barcode decompositions $\mathbf{B}(V)$ and $\mathbf{B}(U)$.
- A **barcode basis** for V is a choice $V \simeq \bigoplus_{i \in \Gamma} k_{[a_i, b_i]}$
- Given a choice of bases for V and U , we might understand f by means of an associated matrix F .

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 $U = \text{PH}_n(\text{VR}(\mathbb{Y}))$ for some $n \in \mathbb{Z}_{\geq 0}$.

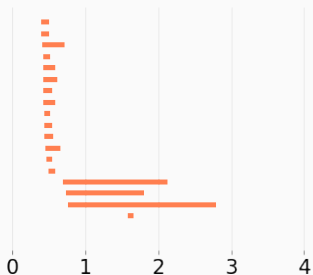
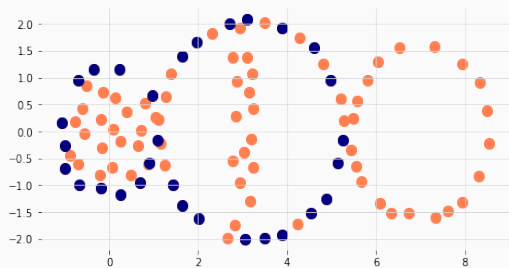
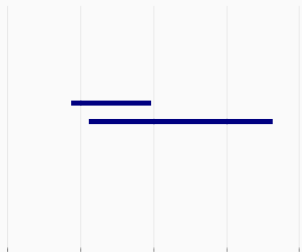
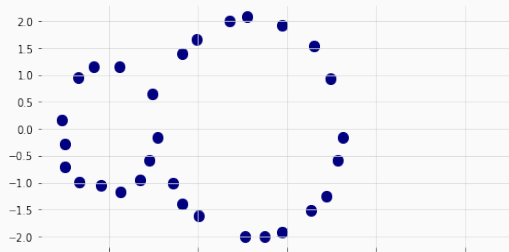
²E. Escolar, Y. Hiraoka. *Persistence modules on commutative ladders of finite type*, Discrete and Computational Geometry, 55 (2014), pp. 100-157

Associated Matrix Computation (Skip?)

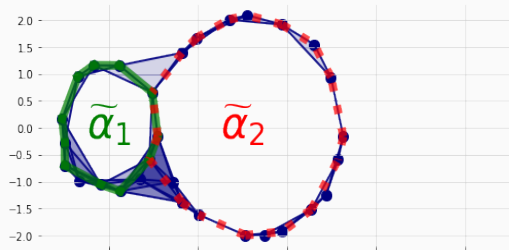
Example

- Consider the reduced matrices R_V and R_U that result from computing $V = \text{PH}_*(\text{VR}(\mathbb{X}))$ and $U = \text{PH}_*(\text{VR}(\mathbb{Y}))$ resp.
- Consider the **cycle representatives** of V , i.e. the submatrix \widetilde{R}_V from R_V that results from keeping the columns labelled by **negative simplices** from nontrivial intervals.
- Note that the rows from \widetilde{R}_V correspond to simplices from $\text{VR}_R^D(\mathbb{X})$.
- Using $\iota : C_*(\text{VR}_R^D(\mathbb{X}); k) \hookrightarrow C_*(\text{VR}_R^D(\mathbb{Y}); k)$, obtain the matrix product $E_V = M_\iota \widetilde{R}_V$, where M_ι is the matrix associated to ι .
- Consider the matrix $(R_U \mid E_V)$ and reduce it; all columns from E_V should vanish.
- Tracking the additions, one gets the associated matrix of $V \rightarrow U$.
- **Caveat:** One might need to do a little more work for “infinite bars”.

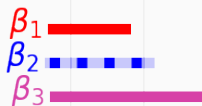
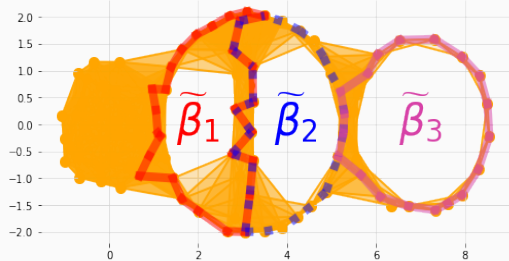
Example: Subset of a bigger Point Cloud



Example: Subset of a bigger Point Cloud



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



0 1 2 3 4

- Let $f : V \rightarrow U$ be a persistence morphism with associated matrix F .
- Sort the intervals from $\mathbf{B}(V)$ following the **standard order**:

$$[a, b) \leq [c, d) \text{ iff } a < c, \text{ or if } a = c \text{ and } d \leq b .$$

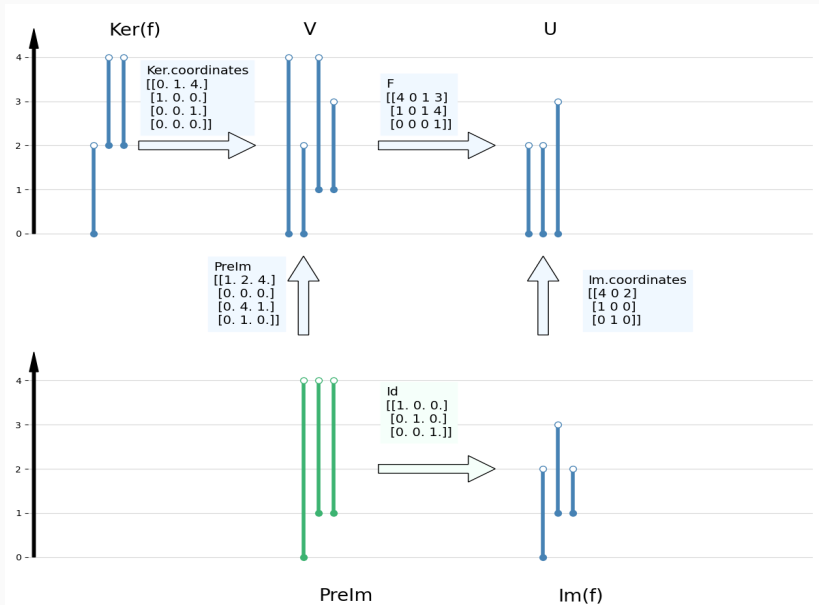
- Sort the intervals from $\mathbf{B}(U)$ following the **endpoint order**:

$$[a, b) \leq [c, d) \text{ iff } b < d, \text{ or if } b = d \text{ and } a \leq c .$$

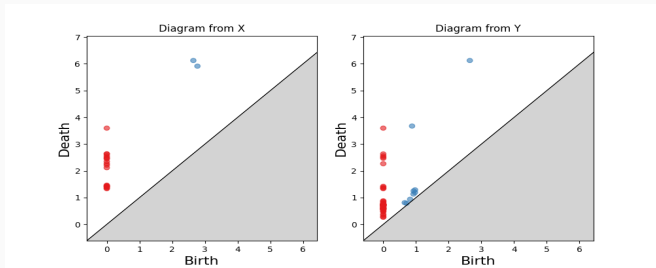
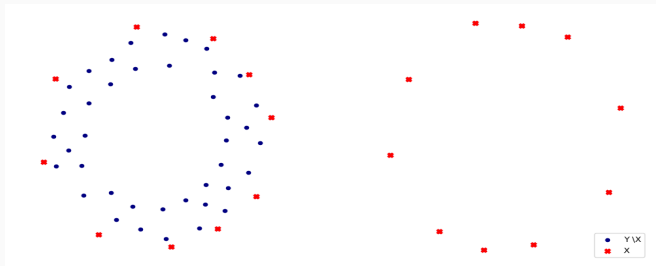
- Consider F with rows and columns reordered.
- Let R be the Gaussian column reduction of F .
- The columns from R generate $\text{Im}(f) \subseteq U$
- A pivot in a column associated to $[a, b)$ and a row associated to $[c, d)$ leads to a bar $[a, d)$ for $\mathbf{B}(\text{Im}(f))$.
- Similarly one can compute kernels and quotients³

³Ch.4 Á Torras-Casas, *Persistence Spectral Sequences*, (2022) Cardiff University.

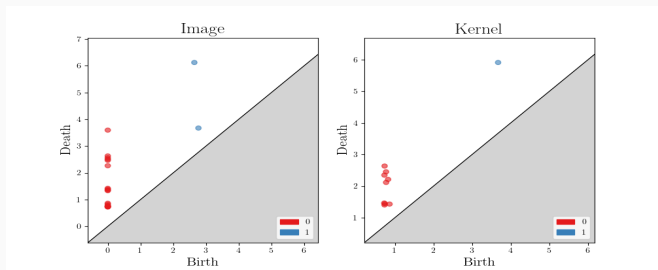
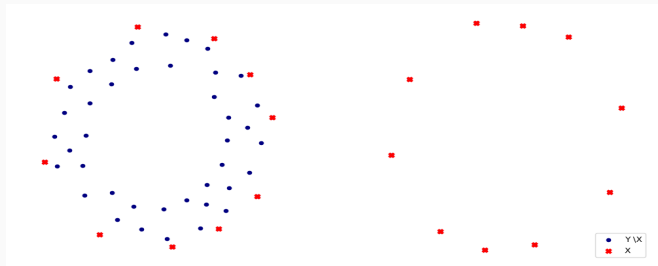
Images and Kernels Illustration



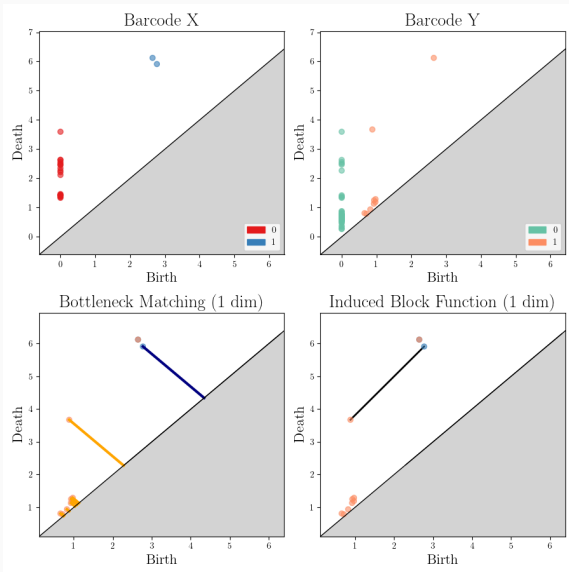
Downside of Images and Kernels (example)



Downside of Images and Kernels (example)



Motivation for the Induced Block Function (example)



Block Functions and Partial Matchings

- A **block function** between $\mathbf{B}_1 = (S_1, m)$ and $\mathbf{B}_2 = (S_2, n)$ is a function $\mathcal{M} : S_1 \times S_2 \rightarrow \mathbb{Z}_{\geq 0} \cup \{\infty\}$ such that:

$$\sum_{J \in S_2} \mathcal{M}(I, J) \leq m_I .$$

- **Assignment:** $\mathcal{M}_f : R_1 \rightarrow R_2$ between subsets $R_1 \subseteq \text{Rep } \mathbf{B}_1$ and $R_2 \subseteq \text{Rep } \mathbf{B}_2$. For ease, we write $\mathcal{M}_f : \text{Rep } \mathbf{B}_1 \rightarrow \text{Rep } \mathbf{B}_2$.
- A **partial matching** is a bijection $\sigma : R_1 \rightarrow R_2$.
- If a block function satisfies

$$\sum_{I \in S_1} \mathcal{M}(I, J) \leq n_J ,$$

it induces a partial matching $\text{Rep } \mathbf{B}_1 \rightarrow \text{Rep } \mathbf{B}_2$.

Example: A block function NOT inducing a partial matching

Example

$B_1 = (S_1, m) = \{([2, 4], 1), ([1, 5], 2)\}$ and

$B_2 = (S_2, n) = \{([2, 3], 1), ([1, 4], 2)\}$

Consider $\mathcal{M} : S_1 \times S_2 \rightarrow \mathbb{Z}_{\geq 0} \cup \{\infty\}$ which is zero except for

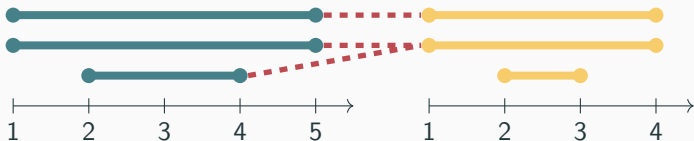
$$\mathcal{M}([2, 4], [1, 4]) = 1 \text{ and } \mathcal{M}([1, 5], [1, 4]) = 2.$$

\mathcal{M} is a block function, since

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

however \mathcal{M} does not induce a partial matching since

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



Example: A block function inducing a Partial Matching

Example

$\mathbf{B}_1 = (S_1, m) = \{([2, 4], 1), ([1, 5], 2)\}$ and

$\mathbf{B}_2 = (S_2, n) = \{([2, 3], 1), ([1, 4], 2)\}$

Consider $\mathcal{M} : S_1 \times S_2 \rightarrow \mathbb{Z}_{\geq 0} \cup \{\infty\}$ which is zero except for

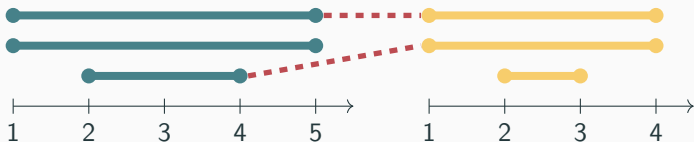
$$\mathcal{M}([2, 4], [1, 4]) = \mathcal{M}([1, 5], [1, 4]) = 1.$$

\mathcal{M} is a block function inducing a partial matching

$\sigma_{\mathcal{M}} : \text{Rep } \mathbf{B}_1 \rightarrow \text{Rep } \mathbf{B}_2$ given by:

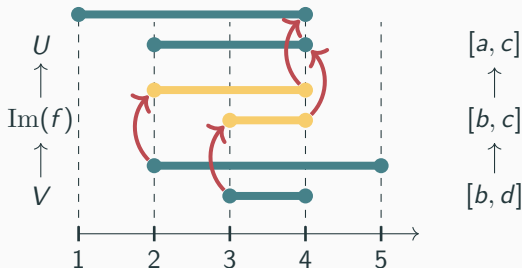
$$[2, 4]_1 \mapsto [1, 4]_1 \text{ and } [1, 5]_1 \mapsto [1, 4]_2$$

while $[2, 3]_1 \in \text{Rep } \mathbf{B}_2$ remains unmatched.



The Bauer-Lesnisk induced partial matching

- Let $f : V \rightarrow U$ be a persistence morphism.
- In 2015 Bauer and Lesnick introduced⁴ an induced partial matching $\chi_f : \text{Rep } \mathbf{B}(V) \rightarrow \text{Rep } \mathbf{B}(U)$.
- χ_f is defined by using $\mathbf{B}(V)$, $\mathbf{B}(U)$ and $\mathbf{B}(\text{Im}(f))$:



⁴U. Bauer, M. Lesnick. *Induced matchings and the algebraic stability of persistence barcodes*, Journal of Computational Geometry, 6 (2) (2015), pp. 162-191

Downside to the Bauer-Lesnicks Partial Matching

- χ_f might be “blind” to f .

Example

Consider the persistence morphism $f : V \rightarrow U$ given by:

$$f = (k_{[2,3]} \rightarrow 0) \oplus (\text{Id} : k_{[2,2]} \rightarrow k_{[1,2]})$$

i.e. $f : k_{[2,3]} \oplus k_{[2,2]} \rightarrow k_{[1,2]}$ with associated matrix:

$$F = \begin{pmatrix} 0 & 1 \end{pmatrix}.$$

One would expect: $[2, 3]_1 \mapsto \emptyset$, $[2, 2]_1 \mapsto [1, 2]_1$.

However, $\text{Im}(f) \simeq k_{[2,2]}$ and χ_f produces:

$$[2, 3]_1 \xrightarrow{\chi_f} [1, 2]_1, \quad [2, 2]_1 \xrightarrow{\chi_f} \emptyset$$

- Additionally, when computing χ_f we might need to check equality between double type variables (!).

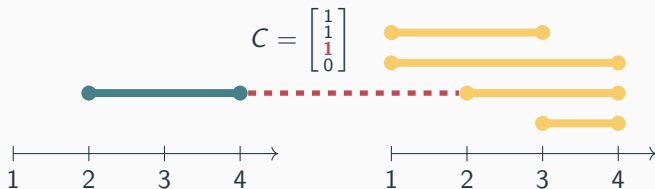
Quick Introduction to the Induced Block Function \mathcal{M}_f

- Let $f: V \rightarrow U$ be a persistence morphism.
- There is an induced block funct.⁵ \mathcal{M}_f from $\mathbf{B}(V)$ to $\mathbf{B}(U)$ s. t.:
 - **Additivity:** Given a direct sum of morphisms:

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

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

- **Pivots:** Let $f: k_I \rightarrow U$ with associated matrix C :



then $\mathcal{M}_f(I, J) \neq 0$ where J the “pivot” that results from the order: $[a, b] \leq [c, d]$ iff $b < d$ or, if $b = d$ then $a \leq c$.

⁵R. González-Díaz, M. Soriano-Trigueros, Á. Torras-Casas, *Partial Matchings induced by Morphisms between Persistence Morphisms*, Comput. Geom., Vol. 112, 2023.

Revisiting the Example and additional property of \mathcal{M}_f

Example

Consider the persistence morphism $f : V \rightarrow U$ given by:

$$f = (k_{[2,3]} \rightarrow 0) \oplus (\text{Id} : k_{[2,2]} \rightarrow k_{[1,2]})$$

then,

- by additivity $\mathcal{M}_f = \mathcal{M}_g$, where $g = \text{Id} : k_{[2,2]} \rightarrow k_{[1,2]}$
- by the pivot property, $\mathcal{M}_f([2, 2], [1, 2]) = 1$.

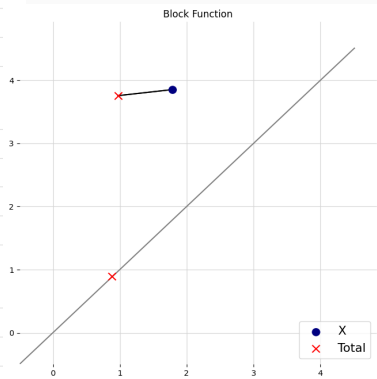
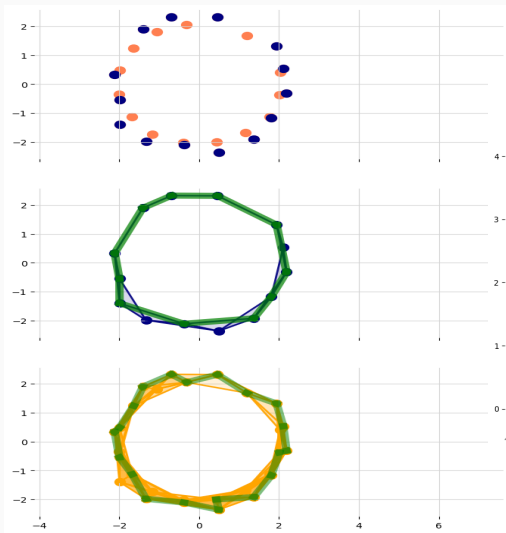
Altogether \mathcal{M}_f is zero everywhere except $\mathcal{M}_f([2, 2], [1, 2]) = 1$. Thus, \mathcal{M}_f induces the expected matching:

$$[2, 3]_1 \mapsto \emptyset, \quad [2, 2]_1 \mapsto [1, 2]_1.$$

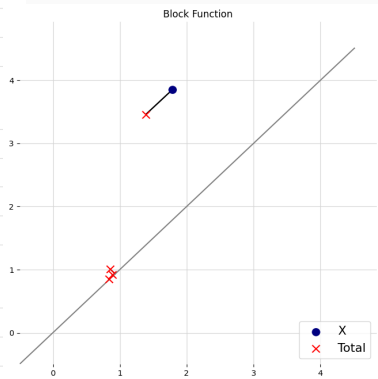
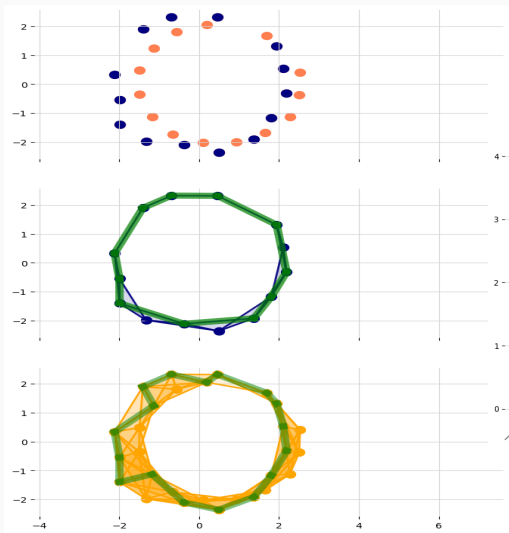
- **Interval Order condition:** given $I = [a, b]$ and $J = [c, d]$, if $\mathcal{M}_f(I, J) \neq 0$, then

$$c \leq a \leq d \leq b.$$

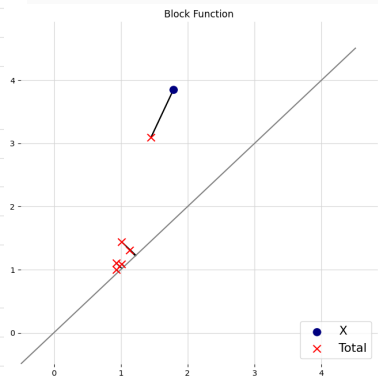
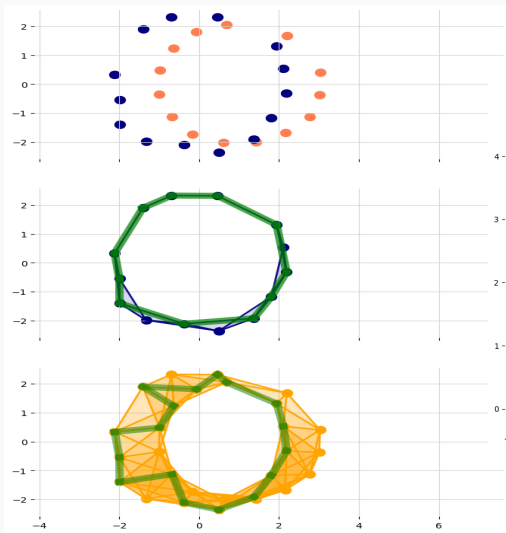
Matching circles in the plane



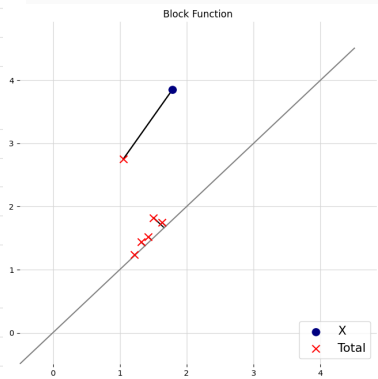
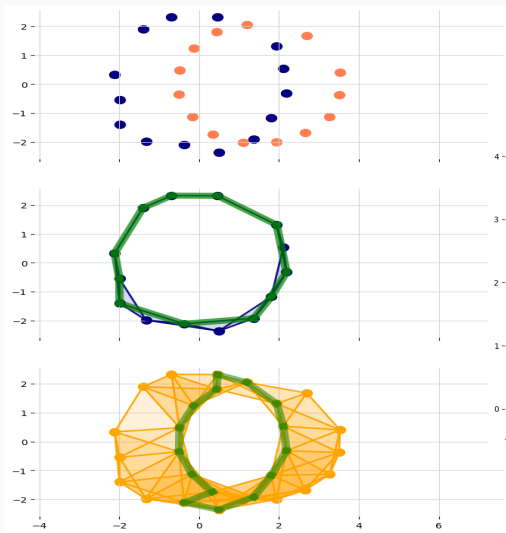
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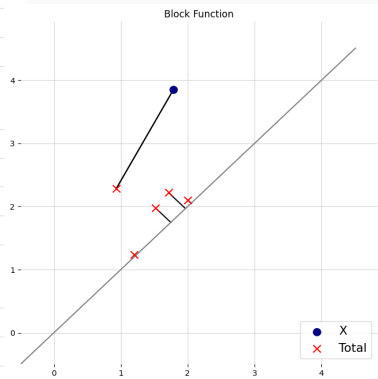
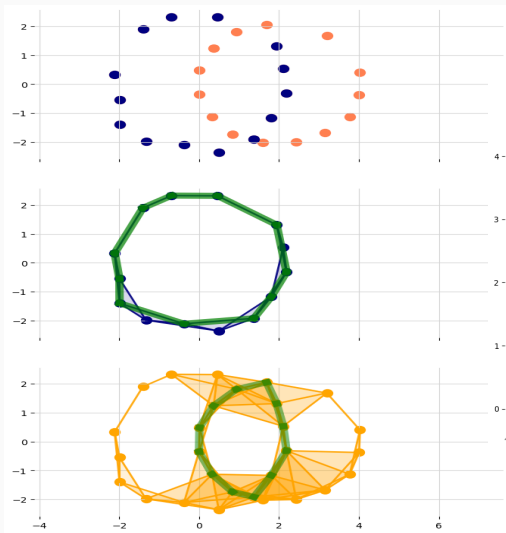
Matching circles in the plane



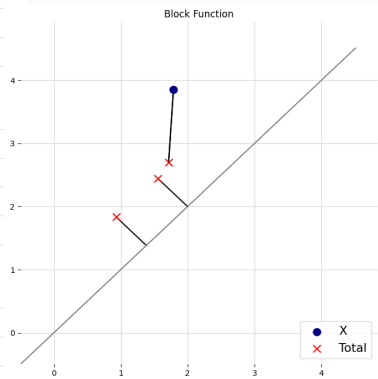
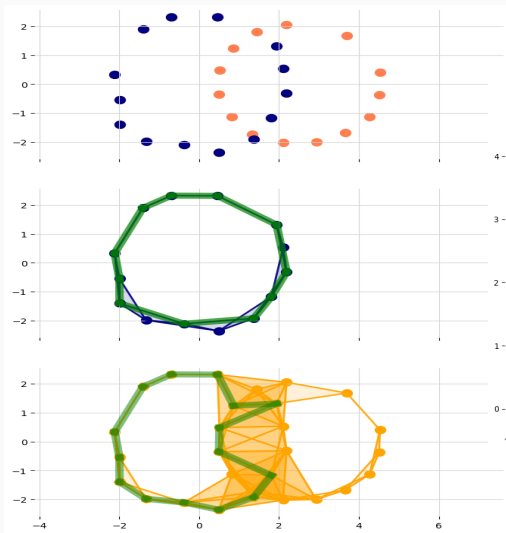
Matching circles in the plane



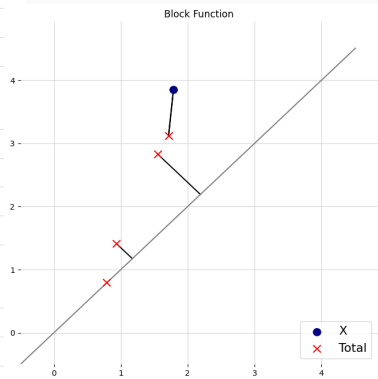
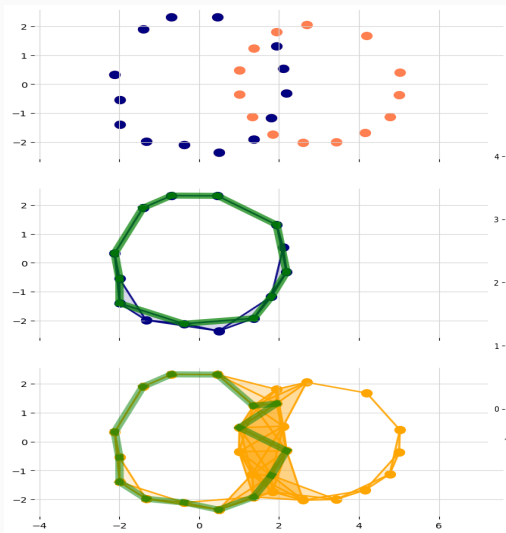
Matching circles in the plane



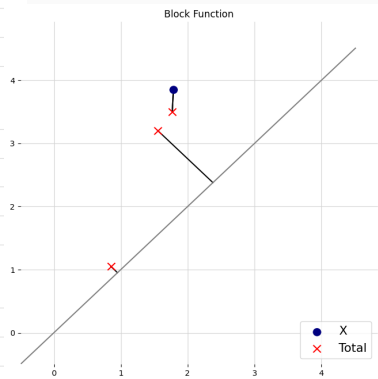
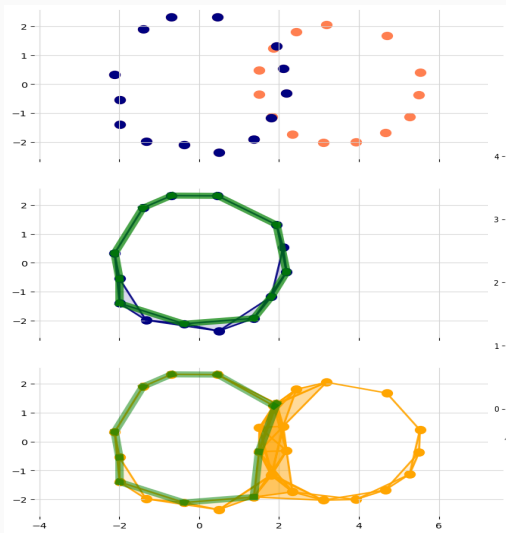
Matching circles in the plane



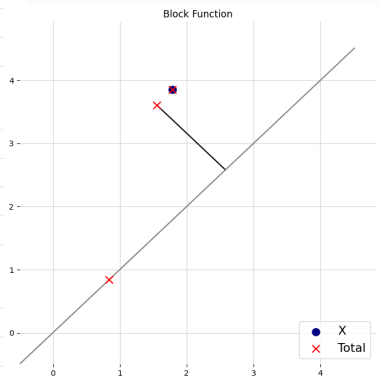
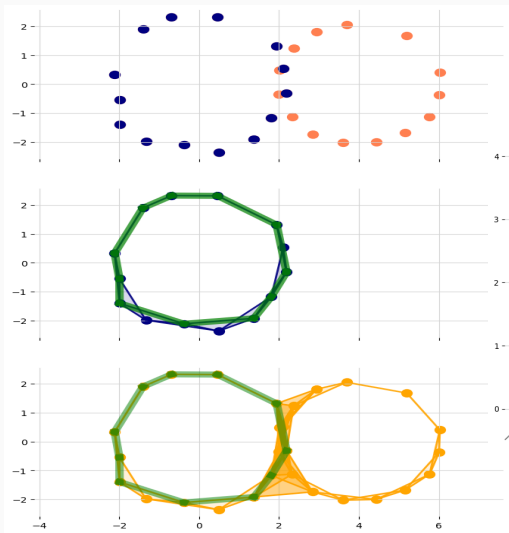
Matching circles in the plane



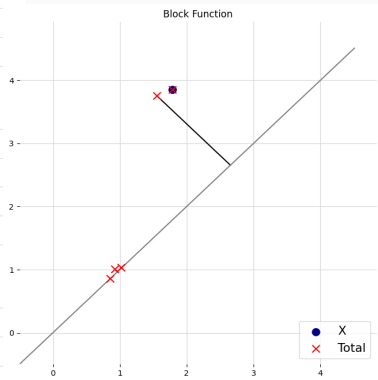
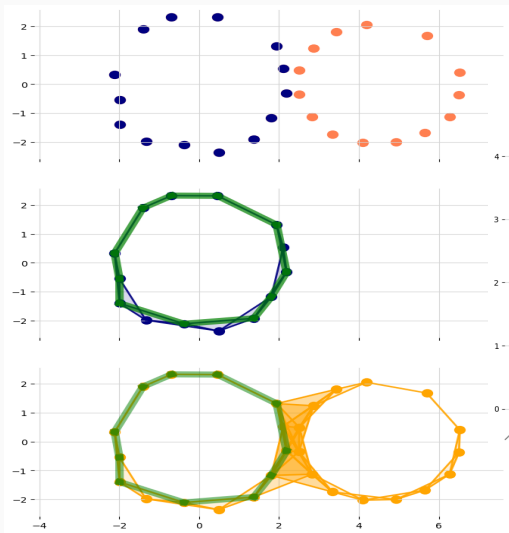
Matching circles in the plane



Matching circles in the plane

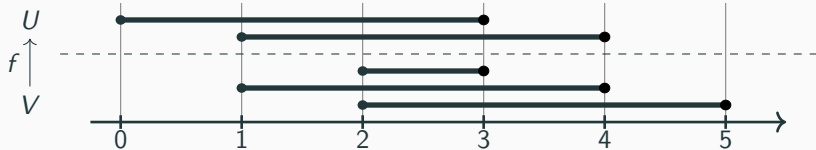


Matching circles in the plane



Example: 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 $\mathbf{B}(V)$ and $\mathbf{B}(U)$ following the endpoint order.



- Suppose that f is associated to the following matrix:

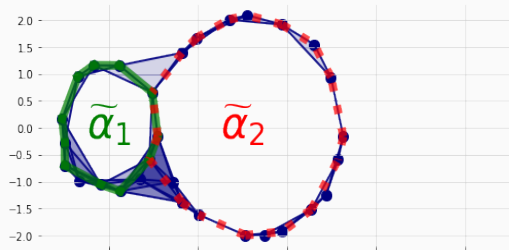
$$F = \left[\begin{array}{c|ccc} & [2, 3] & [1, 4] & [2, 5] \\ \hline [0, 3] & 1 & 1 & 0 \\ [1, 4] & 0 & 1 & 1 \end{array} \right]$$

- Let $I = [a, b]$. Consider F_I , the reduced minor of F restricted to columns associated to $[c, d]$ with $c \leq a$ and $d \leq b$:

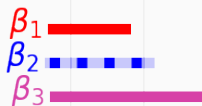
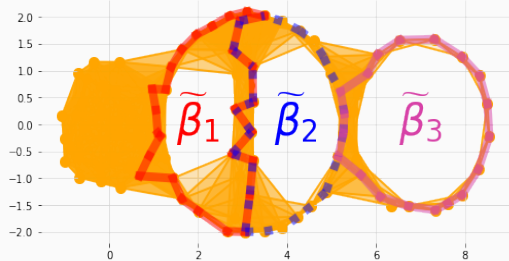
$$F_{[2,3]} = \begin{bmatrix} \mathbf{1} \\ 0 \end{bmatrix}, F_{[1,4]} = \begin{bmatrix} 1 \\ \mathbf{1} \end{bmatrix}, \text{ and } F_{[2,5]} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & \mathbf{0} \end{bmatrix}$$

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

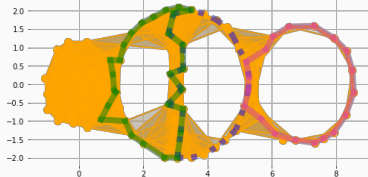
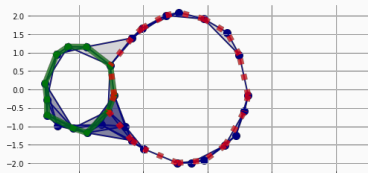
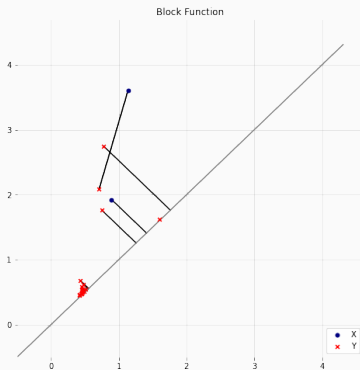
Example: Subset of a bigger Point Cloud



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



Example: Subset of a bigger Point Cloud



Nested intervals and \mathcal{M}_f

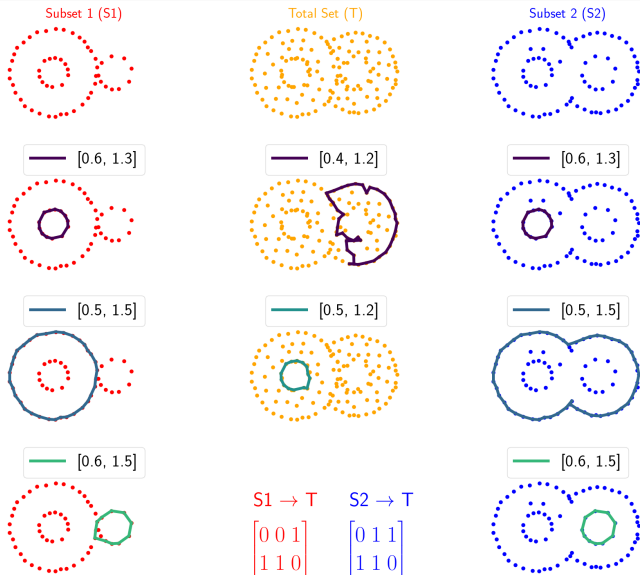
- **Nested Intervals:** $[a, b]$ and $[c, d]$ are nested if $a < c < d < b$
- If for any set of intervals $S \subseteq S_V$ we have that

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

then there exists a pair of nested intervals in S .

- **Corollary** If there are no two nested intervals in S_V then \mathcal{M}_f induces a partial matching.

Example: Two subsets with the same intervals and image



Example: Image computation for S_1

- $f_1 : S_1 \hookrightarrow T$ with
 $\mathbf{B}(\text{PH}_1(\text{VR}(S_1))) = \{[0.6, 1.3], [0.5, 1.5], [0.6, 1.5]\}$ and
 $\mathbf{B}(\text{PH}_1(\text{VR}(T))) = \{[0.4, 1.2], [0.5, 1.2]\}$.
- Order domain by standard order and codomain by endpoint order:

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

- Obtain the reduction:

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

- Image barcodes: $\mathbf{B}(\text{Im}(f_1)) = \{[0.5, 1.2], [0.6, 1.2]\}$.

Example: Image computation for S_2

- $f_2 : S_2 \hookrightarrow T$ with
 $\mathbf{B}(\text{PH}_1(\text{VR}(S_2))) = \{[0.6, 1.3], [0.5, 1.5], [0.6, 1.5]\}$ and
 $\mathbf{B}(\text{PH}_1(\text{VR}(T))) = \{[0.4, 1.2], [0.5, 1.2]\}$.
- Order domain by standard order and codomain by endpoint order:

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

- Obtain the reduction:

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

- Image barcodes: $\mathbf{B}(\text{Im}(f_2)) = \{[0.5, 1.2], [0.6, 1.2]\}$.
- i.e. $\text{Im}(f_1) \simeq \text{Im}(f_2) \simeq k_{[0.5, 1.2]} \oplus k_{[0.6, 1.2]}$

Example: Computation of \mathcal{M}_{f_1}

- Now, sort both $\mathbf{B}(S_1)$ and $\mathbf{B}(T)$ by endpoint order.
- We have a matrix

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

- Obtain the matrices:

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

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

Example: Computation of \mathcal{M}_{f_2}

- Now, sort both $\mathbf{B}(S_2)$ and $\mathbf{B}(T)$ by endpoint order.
- We have a matrix

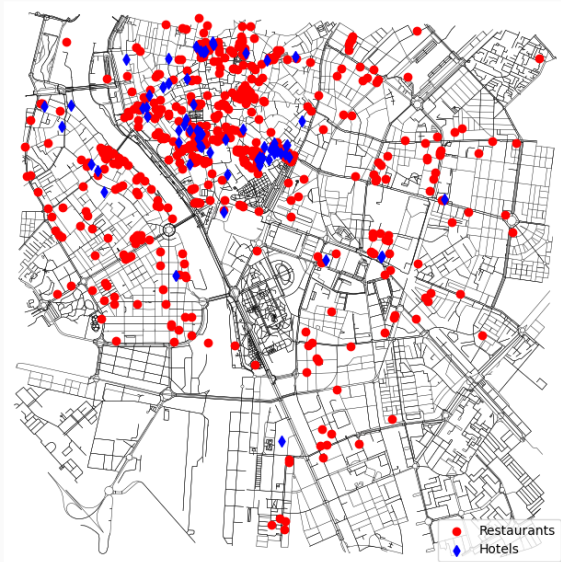
$$F = \left[\begin{array}{c|ccc} & [0.6, 1.3] & [0.5, 1.5] & [0.6, 1.5] \\ \hline [0.4, 1.2] & 0 & 1 & 1 \\ [0.5, 1.2] & 1 & 1 & 0 \end{array} \right]$$

- Obtain the matrices:

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

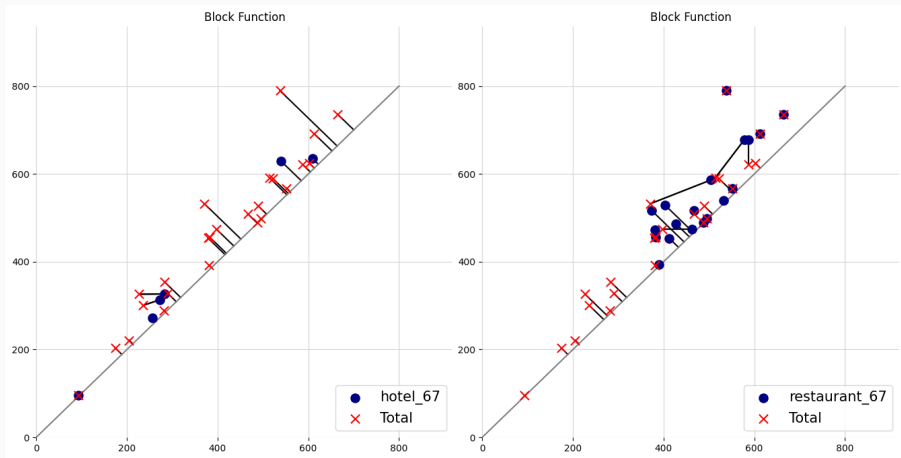
- Assignment: $[0.6, 1.3] \mapsto [0.5, 1.2]$ and $[0.5, 1.5] \mapsto [0.5, 1.2]$.
- We might distinguish f_1 and f_2 based on \mathcal{M}_{f_1} and \mathcal{M}_{f_2}

OSM Data Example: Hotels and Restaurants in Seville

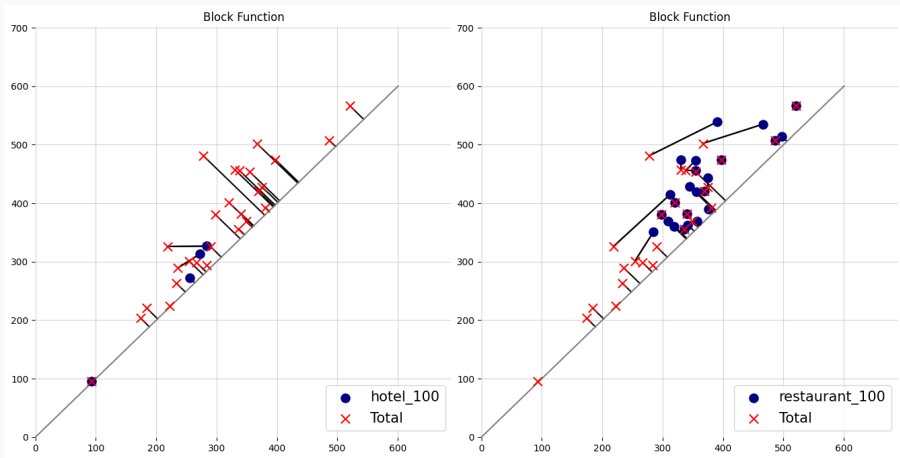


- There are 67 Hotels and 499 restaurants.

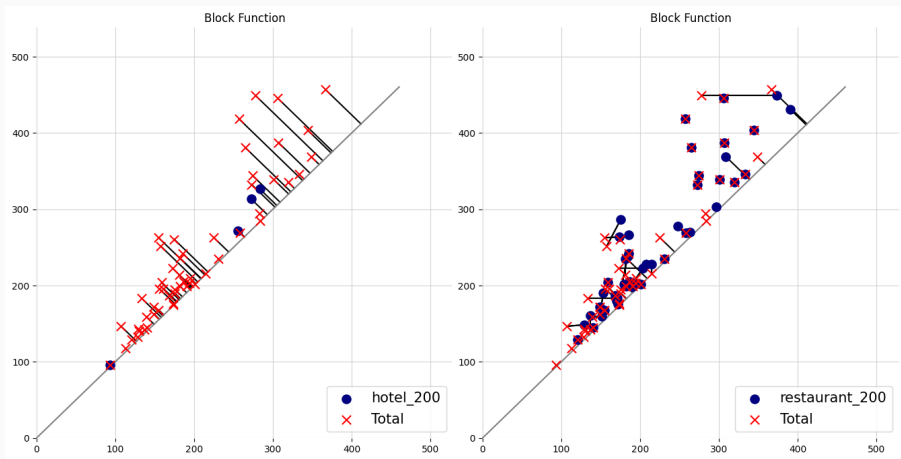
Sample of 67 restaurants



Sample of 100 restaurants








Sample of 200 restaurants



- Can we obtain an alternative definition for an induced block function $\widetilde{\mathcal{M}}_f$ which always induces a partial matching? **yes, work in progress.**
- Optimal implementations for computing the associated matrix.
- Work with other filtrations; e.g. Block functions between alpha complexes.
- Find stability conditions for \mathcal{M}_f
- Find use-cases for this block function.

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Thank You!

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