

MATH412/COMPSCI434/MATH713
Fall 2025

Topological Data Analysis

Topic 5: Stability

Instructor: Ling Zhou

Different choices of intervals

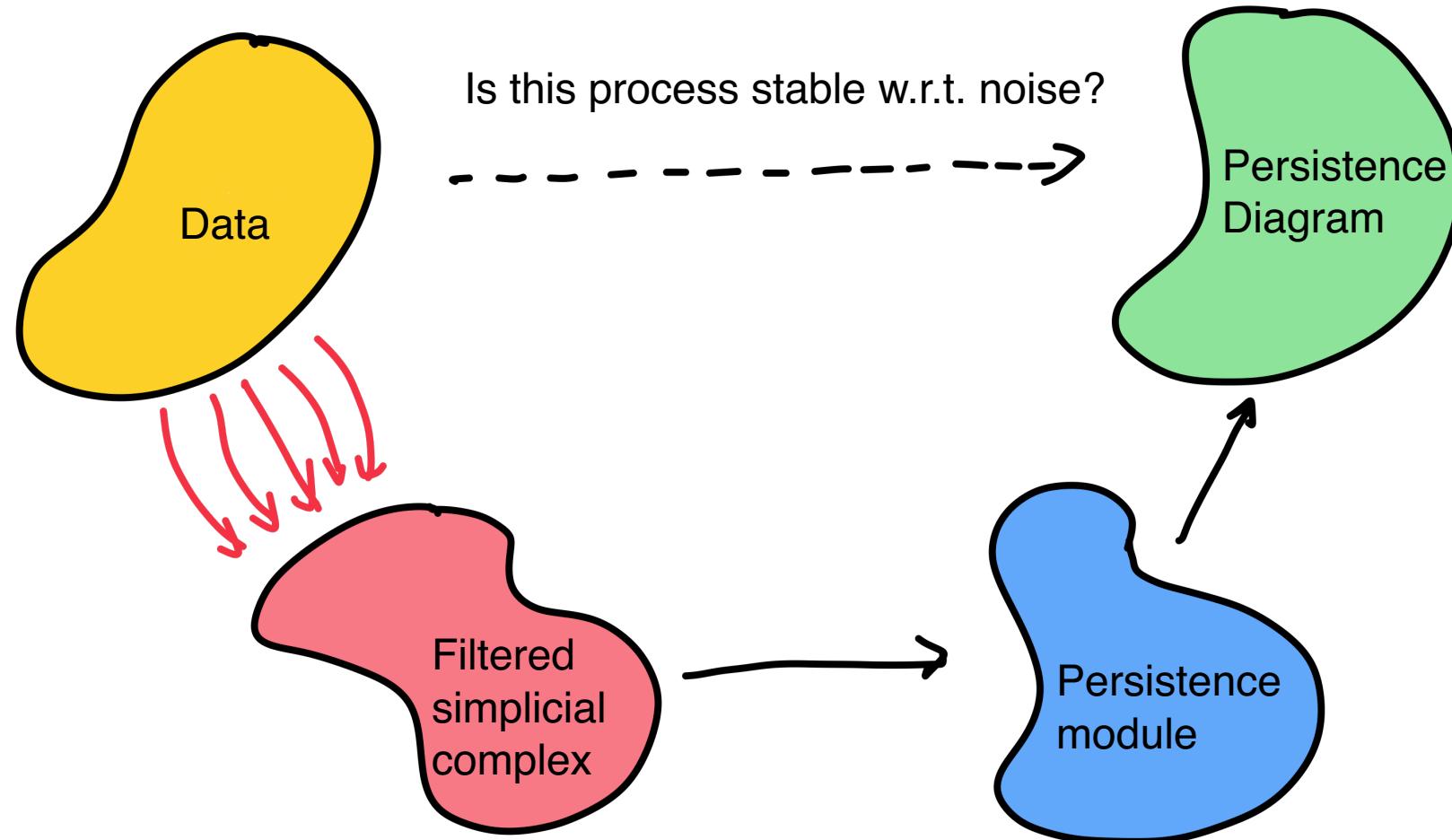
Definition 3.20 (Interval module). Given an index set $A \subseteq \mathbb{R}$ and a pair of indices $b, d \in A$, $b \leq d$, four types of *interval modules* denoted $\mathbb{I}[b, d)$, $\mathbb{I}(b, d]$, $\mathbb{I}[b, d]$, $\mathbb{I}(b, d)$ respectively are special persistence modules defined as:

- (closed-open): $\mathbb{I}[b, d) : \{V_a \xrightarrow{\nu_{a,a'}} V_{a'}\}_{a,a' \in A}$ where (i) $V_a = \mathbb{Z}_2$ for all $a \in [b, d)$ and $V_a = 0$ otherwise, (ii) $\nu_{a,a'}$ is identity map for $b \leq a \leq a' < d$ and zero map otherwise.
- (open-closed): $\mathbb{I}(b, d] : \{V_a \xrightarrow{\nu_{a,a'}} V_{a'}\}_{a,a' \in A}$ where (i) $V_a = \mathbb{Z}_2$ for all $a \in (b, d]$ and $V_a = 0$ otherwise, (ii) $\nu_{a,a'}$ is identity map for $b < a \leq a' \leq d$ and zero map otherwise.
- (closed-closed): $\mathbb{I}[b, d] : \{V_a \xrightarrow{\nu_{a,a'}} V_{a'}\}_{a,a' \in A}$ where (i) $V_a = \mathbb{Z}_2$ for all $a \in [b, d]$ and $V_a = 0$ otherwise, (ii) $\nu_{a,a'}$ is identity map for $b \leq a \leq a' \leq d$ and zero map otherwise.
- (open-open): $\mathbb{I}(b, d) : \{V_a \xrightarrow{\nu_{a,a'}} V_{a'}\}_{a,a' \in A}$ where (i) $V_a = \mathbb{Z}_2$ for all $a \in (b, d)$ and $V_a = 0$ otherwise, (ii) $\nu_{a,a'}$ is identity map for $b < a \leq a' < d$ and zero map otherwise.

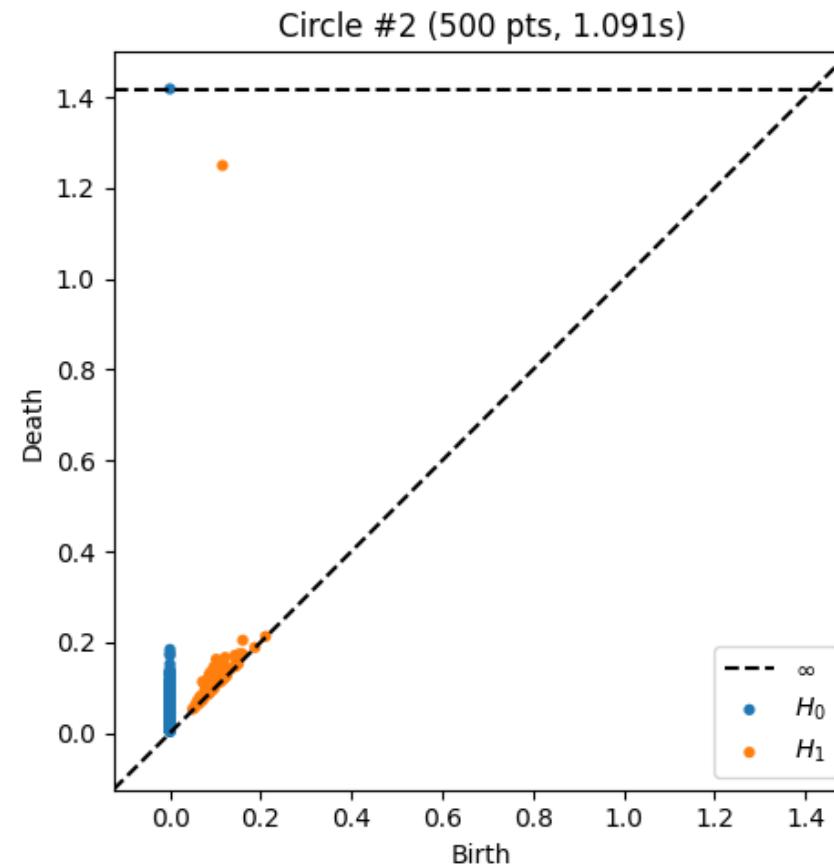
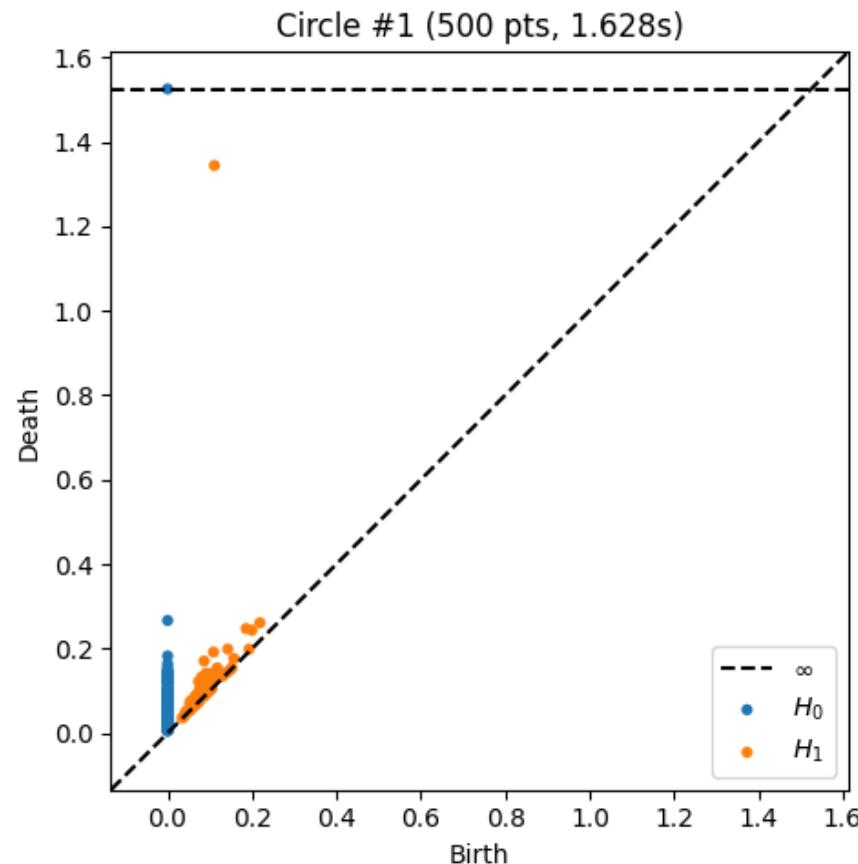
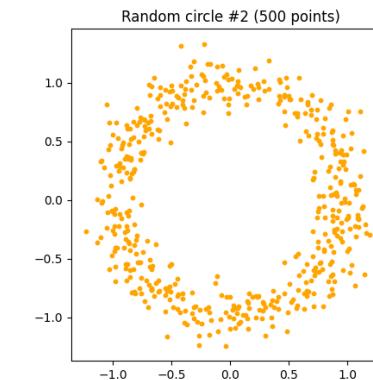
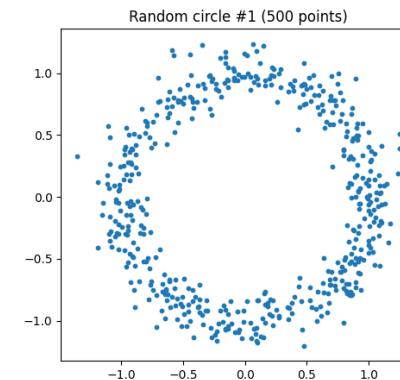
Different choices of intervals

- ▶ **(closed-open)** $\beta_p^{i,j}$ = number of intervals **[k, l)** that contains $[i, j]$
- ▶ **(open-closed)** $\beta_p^{i,j}$ = number of intervals **(k, l]** that contains $[i, j]$
- ▶ **(closed-closed)** $\beta_p^{i,j}$ = number of intervals **[k, l]** that contains $[i, j]$
- ▶ **(open-open)** $\beta_p^{i,j}$ = number of intervals **(k, l)** that contains $[i, j]$
- ▶ $\mu_p^{i,j} = (\beta_p^{i,j-1} - \beta_p^{i,j}) - (\beta_p^{i-1,j-1} - \beta_p^{i-1,j})$

Persistence-based Framework

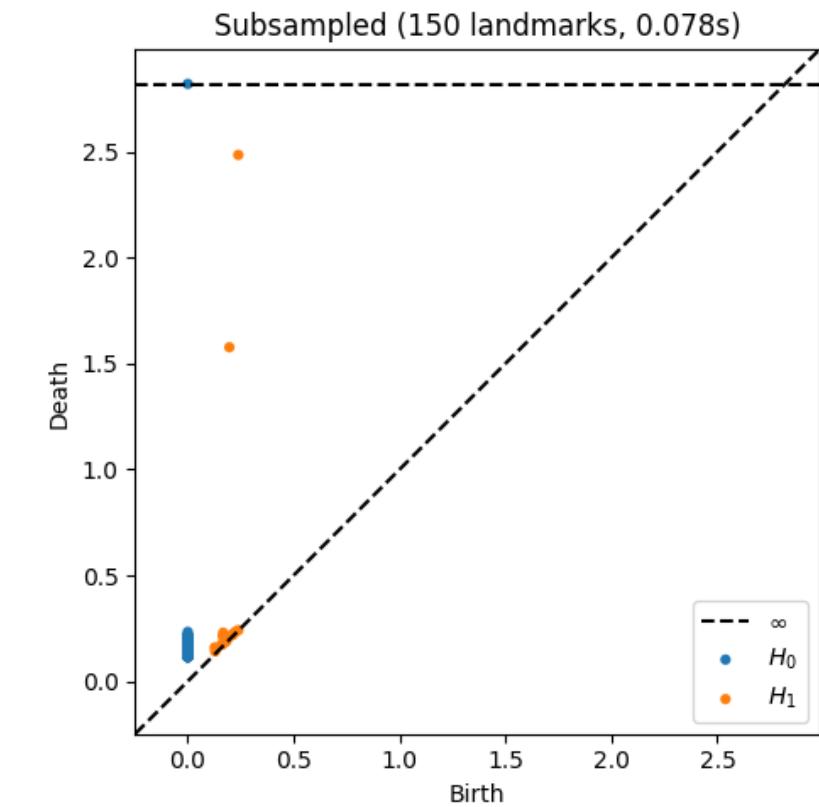
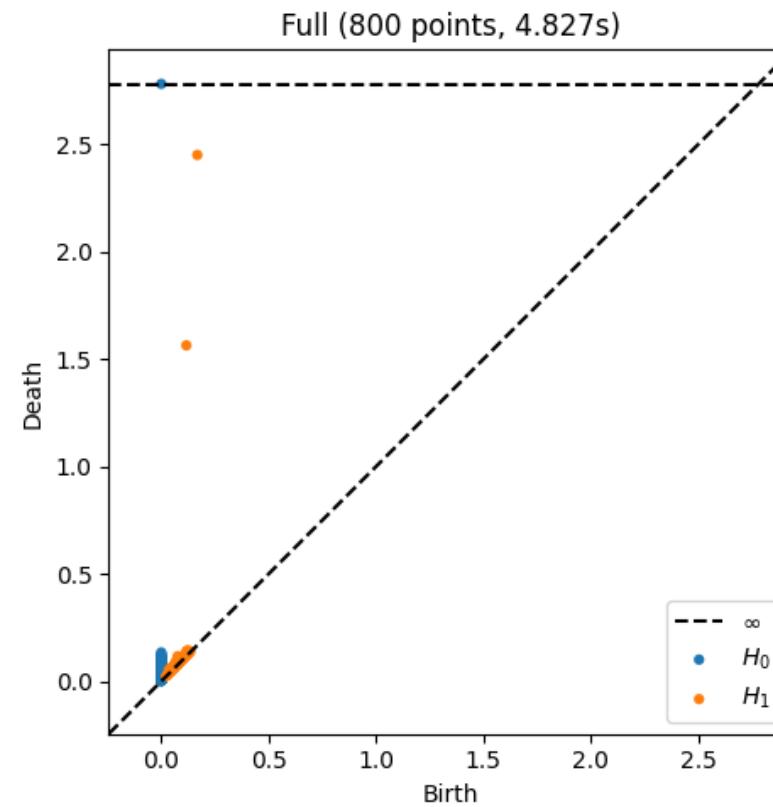
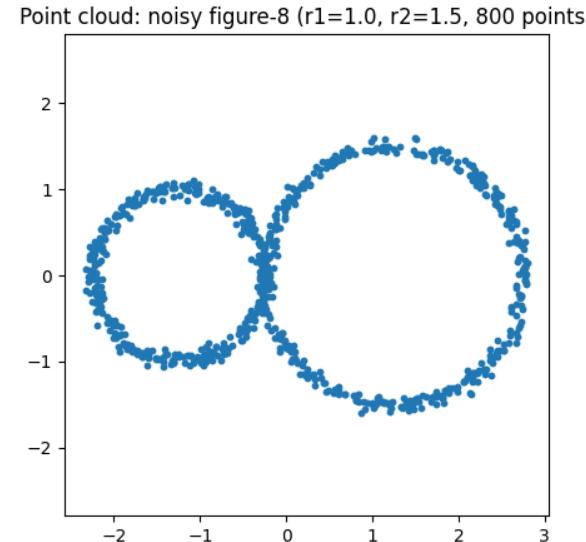


Similar Data => Similar PD



Similar Data => Similar PD

- Allows the use of subsampling to reduce runtime.



```
[uniform_square] n= 50, H0=50, H1=9, time=0.006s
[uniform_square] n= 100, H0=100, H1=19, time=0.015s
[uniform_square] n= 500, H0=500, H1=121, time=0.754s
[uniform_square] n=1000, H0=1000, H1=247, time=2.488s
[uniform_square] n=2000, H0=2000, H1=541, time=6.109s
[uniform_square] n=3000, H0=3000, H1=777, time=10.767s
-----
[gaussian_cloud] n= 50, H0=50, H1=7, time=0.004s
[gaussian_cloud] n= 100, H0=100, H1=18, time=0.006s
[gaussian_cloud] n= 500, H0=500, H1=120, time=0.248s
[gaussian_cloud] n=1000, H0=1000, H1=270, time=1.032s
[gaussian_cloud] n=2000, H0=2000, H1=527, time=5.476s
[gaussian_cloud] n=3000, H0=3000, H1=797, time=12.846s
-----
[circle_random] n= 50, H0=50, H1=1, time=0.005s
[circle_random] n= 100, H0=100, H1=1, time=0.020s
[circle_random] n= 500, H0=500, H1=1, time=3.363s
[circle_random] n=1000, H0=1000, H1=1, time=27.641s
```

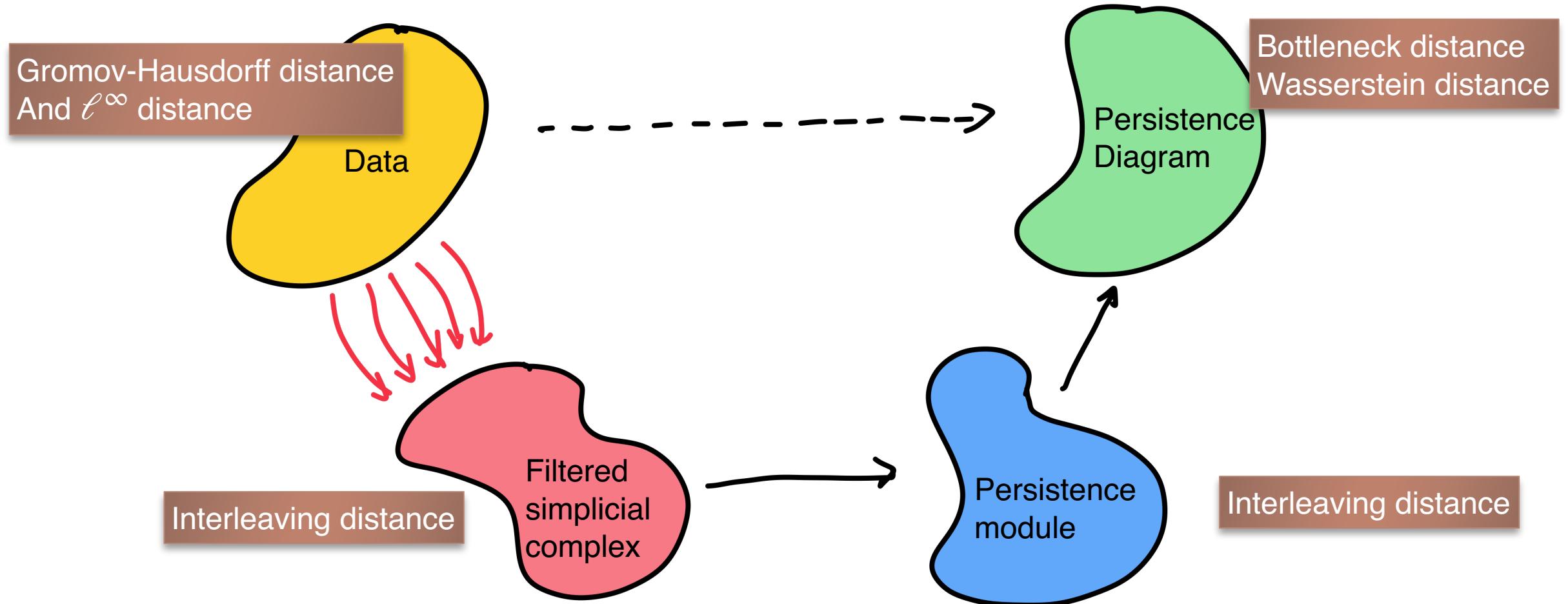
crashed...

```
[uniform_square] n= 50, H0=50, H1=9, H2=0, time=0.016s
[uniform_square] n= 100, H0=100, H1=24, H2=0, time=0.092s
[uniform_square] n= 200, H0=200, H1=55, H2=0, time=0.766s
[uniform_square] n= 400, H0=400, H1=96, H2=1, time=6.080s
[uniform_square] n= 600, H0=600, H1=155, H2=2, time=28.395s
```

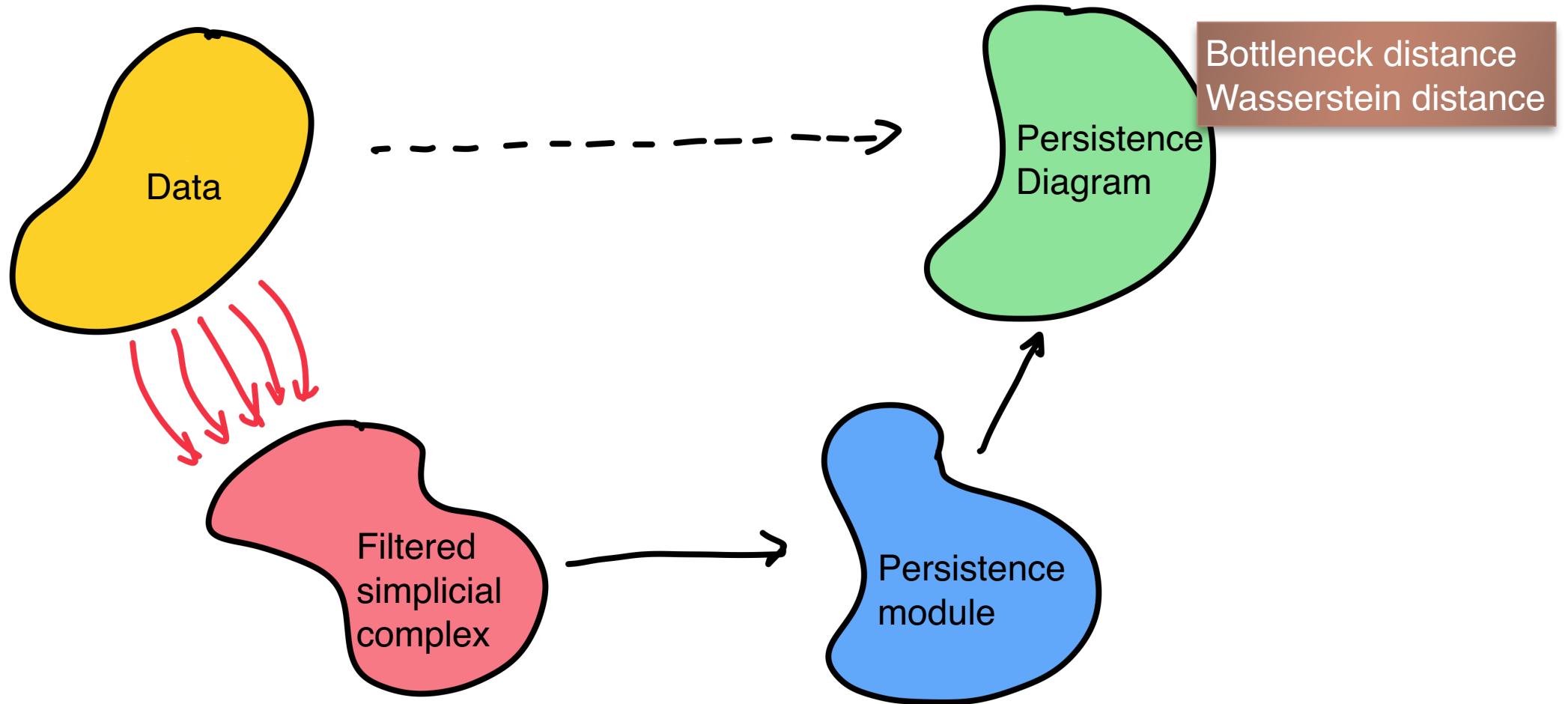
```
[gaussian_cloud] n= 50, H0=50, H1=8, H2=0, time=0.022s
[gaussian_cloud] n= 100, H0=100, H1=20, H2=1, time=0.186s
[gaussian_cloud] n= 200, H0=200, H1=51, H2=1, time=1.801s
[gaussian_cloud] n= 400, H0=400, H1=93, H2=1, time=13.391s
[gaussian_cloud] n= 600, H0=600, H1=167, H2=1, time=47.470s
```

```
[circle_random] n= 50, H0=50, H1=1, H2=6, time=0.022s
[circle_random] n= 100, H0=100, H1=1, H2=11, time=0.246s
[circle_random] n= 200, H0=200, H1=1, H2=24, time=2.134s
[circle_random] n= 400, H0=400, H1=1, H2=56, time=16.814s
[circle_random] n= 600, H0=600, H1=1, H2=75, time=62.211s
```

Using metrics to measure perturbations



Bottleneck Distance for PD



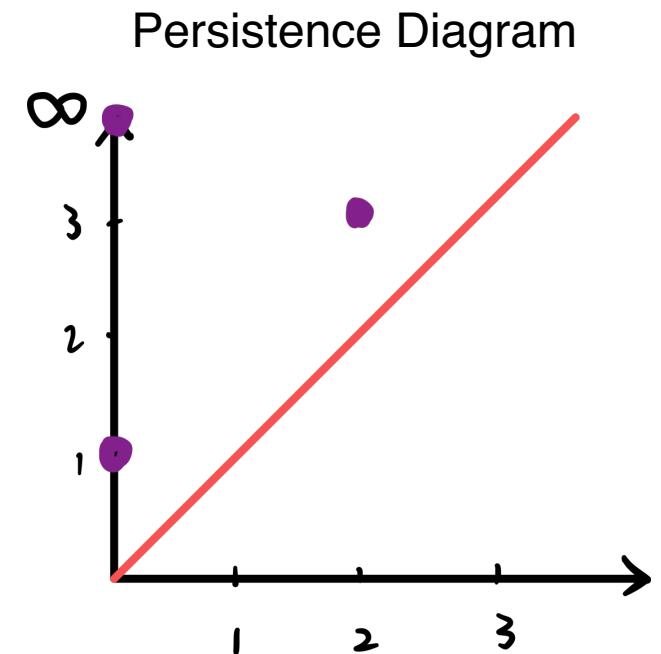
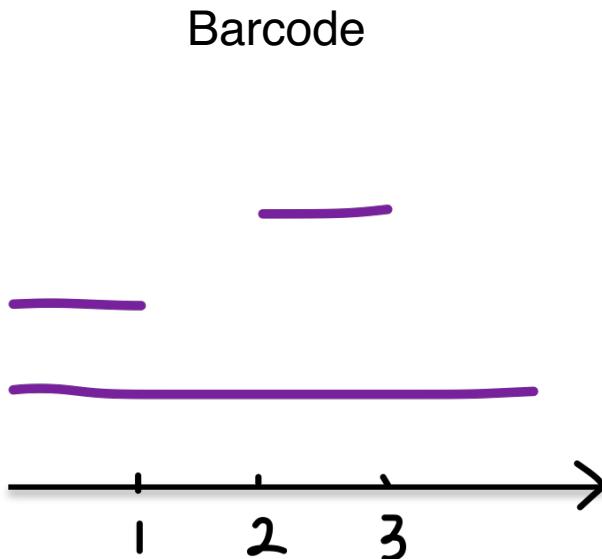
Recall: Persistence Diagram and Barcodes

- ▶ Assume we have decomposed a persistence module as

$$V_{\bullet} \cong I[b_1, d_1) \oplus I[b_2, d_2) \oplus \cdots \oplus I[b_M, d_M)$$

- ▶ The collection of intervals $\{[b_j, d_j)\}_{j=1,\dots,M}$ is called the **barcode** of V .
- ▶ The multiset $D = \{(b_j, d_j)\}_{j=1,\dots,M}$ is called the **persistence diagram** of V .

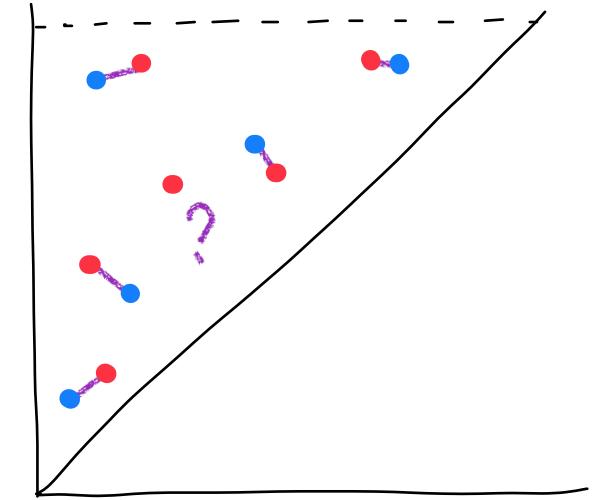
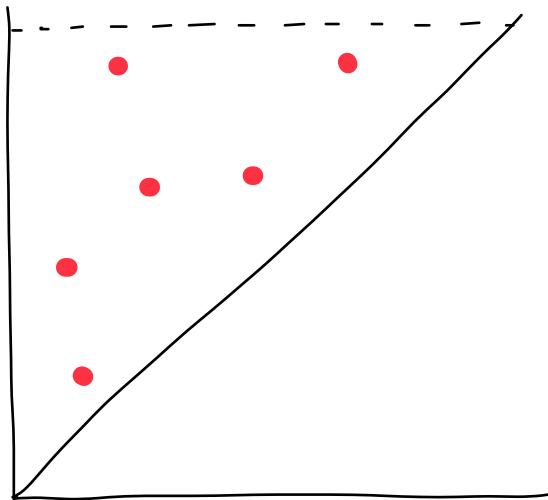
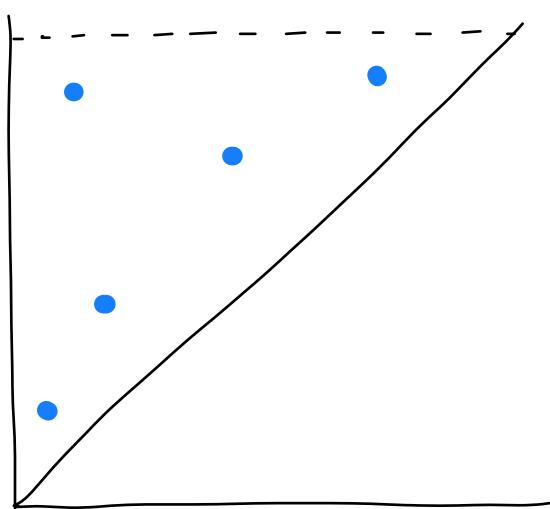
$$H_0(K_{\bullet}) \cong I[0, \infty) \oplus I[0, 1) \oplus I[2, 3)$$



(General) Persistence Diagram

- ▶ Any finite multiset $D = \{(b_j, d_j)\}_{j=1,\dots,M} \subseteq (\mathbb{R} \cup \infty)^2$ is called the **persistence diagram (PD)**, where $0 \leq b_i < d_i \leq \infty$ for each $i = 1, \dots, M$
- ▶ How to compare two different persistence diagrams?

Matching points



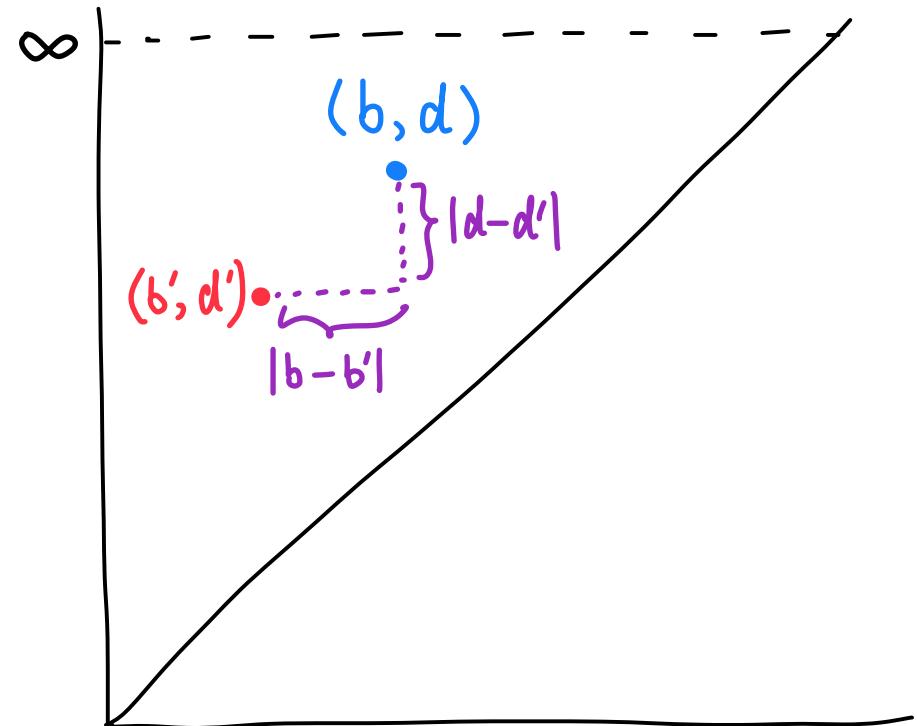
Distance Between Points

- Given two points $p = (b, d)$ and $q = (b', d') \in (\mathbb{R} \cup \infty)^2$
- Distance between two points is

$$\|p - q\|_\infty = \max(|b - b'|, |d - d'|)$$

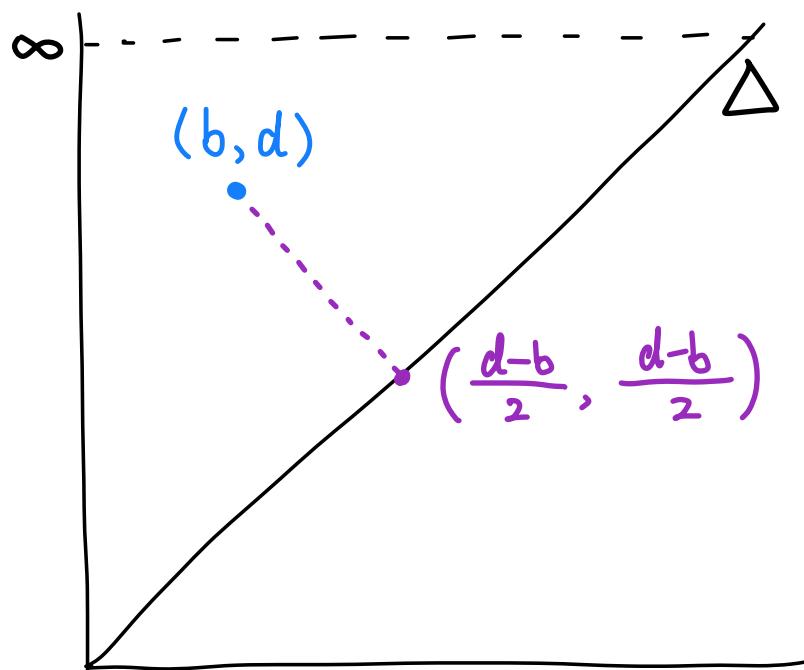
- Here, we assume:

- $\infty - \infty = 0$
- $\infty - \text{finite} = \infty$



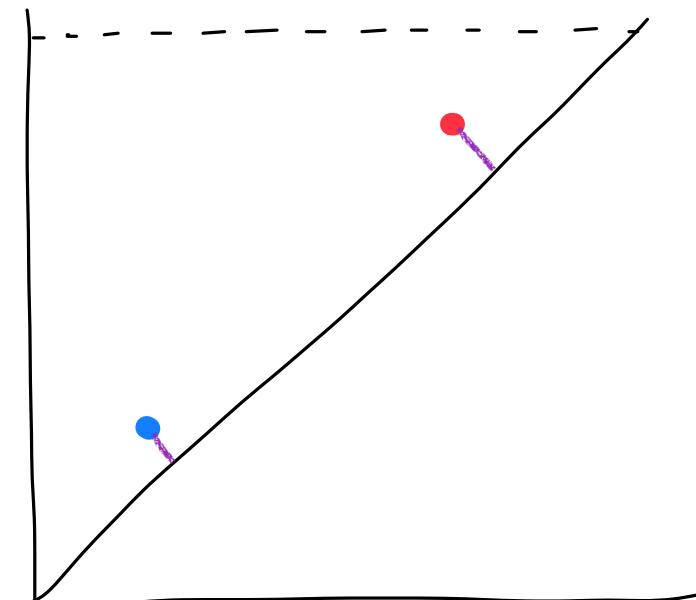
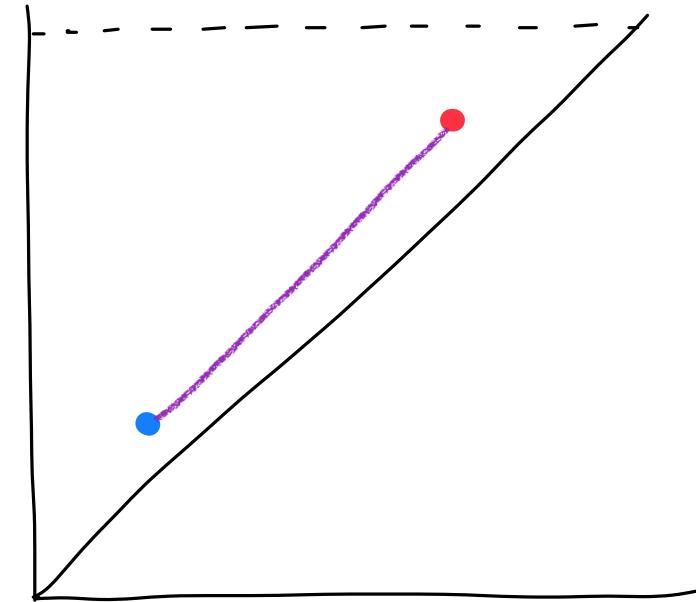
Distance Between Point and Diagonal

- ▶ Let $\Delta = \{(x, y) \mid x = y\}$ be the set of the diagonal points.
- ▶ Distance between a point and the diagonal is $\|p - \Delta\|_\infty = \frac{|b - d|}{2}$
- ▶ $\|p - \Delta\|_\infty$ is the distance between p and its closest point on the diagonal



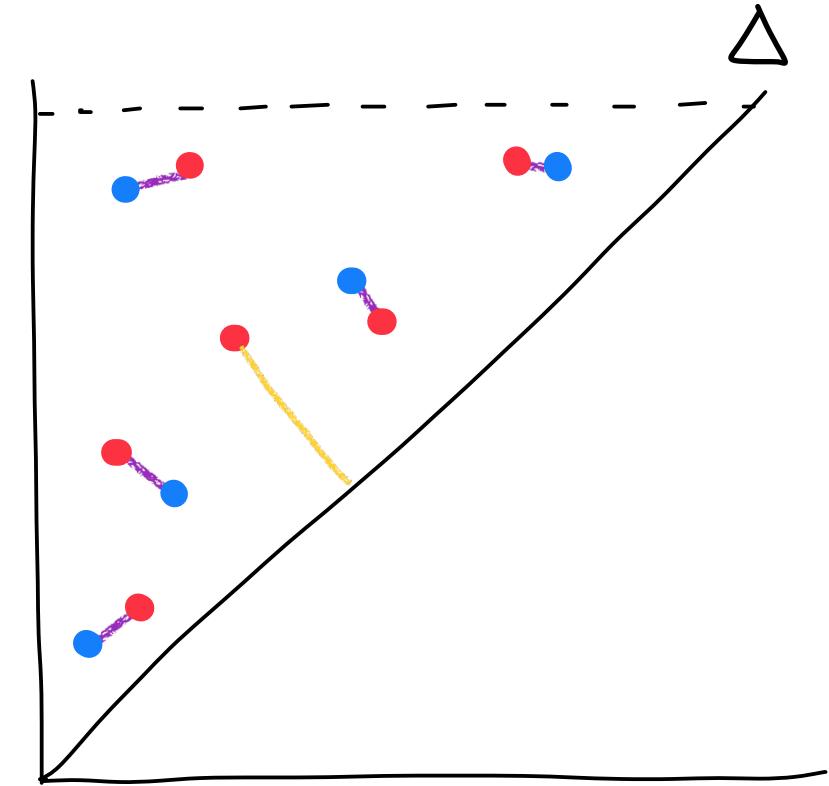
Motivation

- ▶ Points close to the diagonal
 - ▶ short bars, i.e. short lifetime
 - ▶ non-important features
- ▶ Non-important features should NOT be matched with large costs.



Motivating examples

- ▶ Two persistence diagrams D and D' may have different number of points
- ▶ There may be no bijection between D and D'
- ▶ Strategy:
 - ▶ Match part of D and part of D'
 - ▶ Compute ℓ^∞ between matched pairs
 - ▶ Record “importance” of unmatched points;
(equivalently, these points are matched to Δ)



Partial-matching: how to match points

- Given two persistence-diagrams (multiset of points in $(\mathbb{R} \cup \{\infty\})^2$)
 - $D_1 = \{p_1, p_2, \dots, p_s\}$ and $D_2 = \{q_1, q_2, \dots, q_t\}$
- A **partial-matching** between D_1 and D_2 is $M \subseteq D_1 \times D_2$ s.t.
 - $\forall p \in D_1, \exists \text{ at most one } (p, x) \in M$
 - $\forall q \in D_2, \exists \text{ at most one } (x, q) \in M$
 - If $(p, q) \in M$, we say p is **matched** with q.

$$M_1 = \{(p_1, q_1)\}$$

$$M_2 = \{(p_1, q_1), (p_1, q_2)\}$$

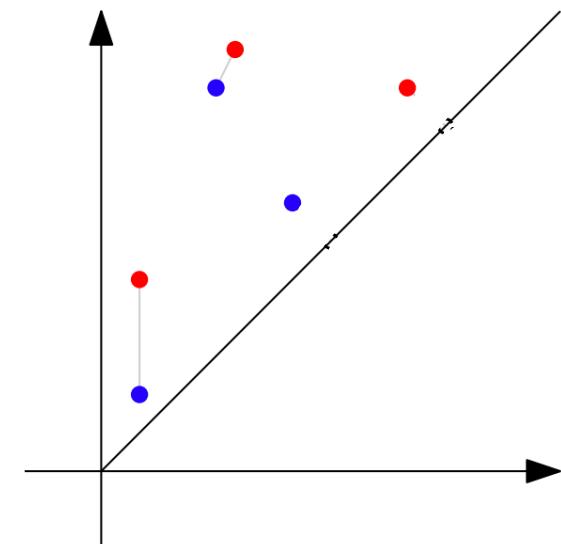
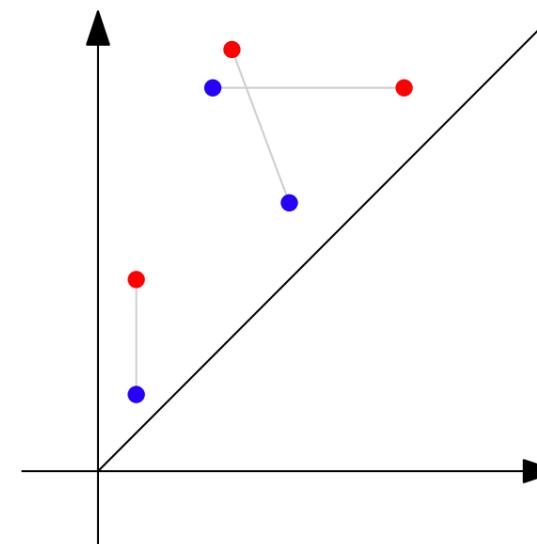
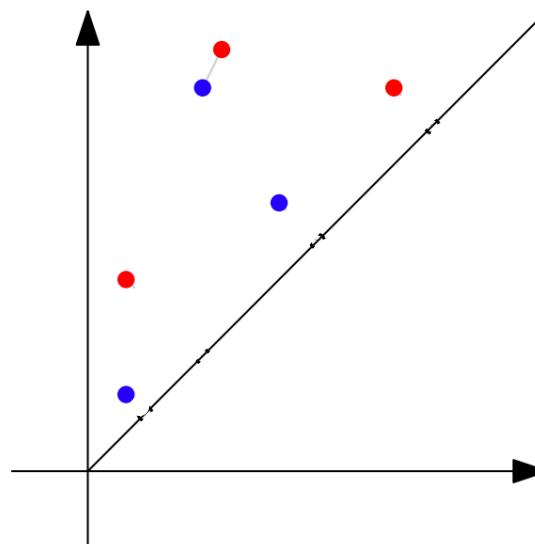
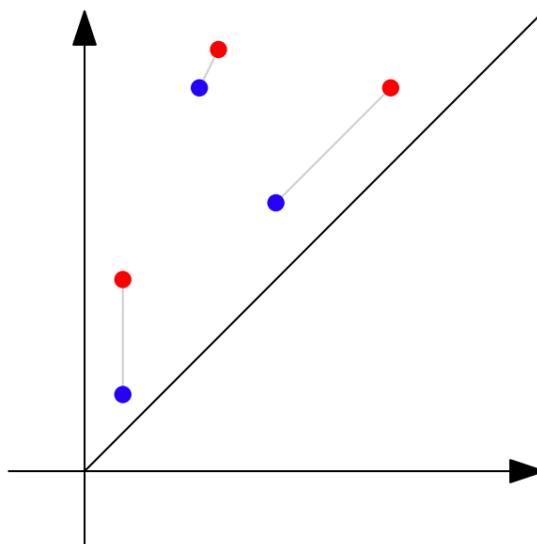
$$M_3 = \{(p_1, q_2), (p_2, q_1)\}$$

Example $D_1 = \{p_1, p_2\}$

$$D_2 = \{q_1, q_2\}$$

Partial-matching: how to match points

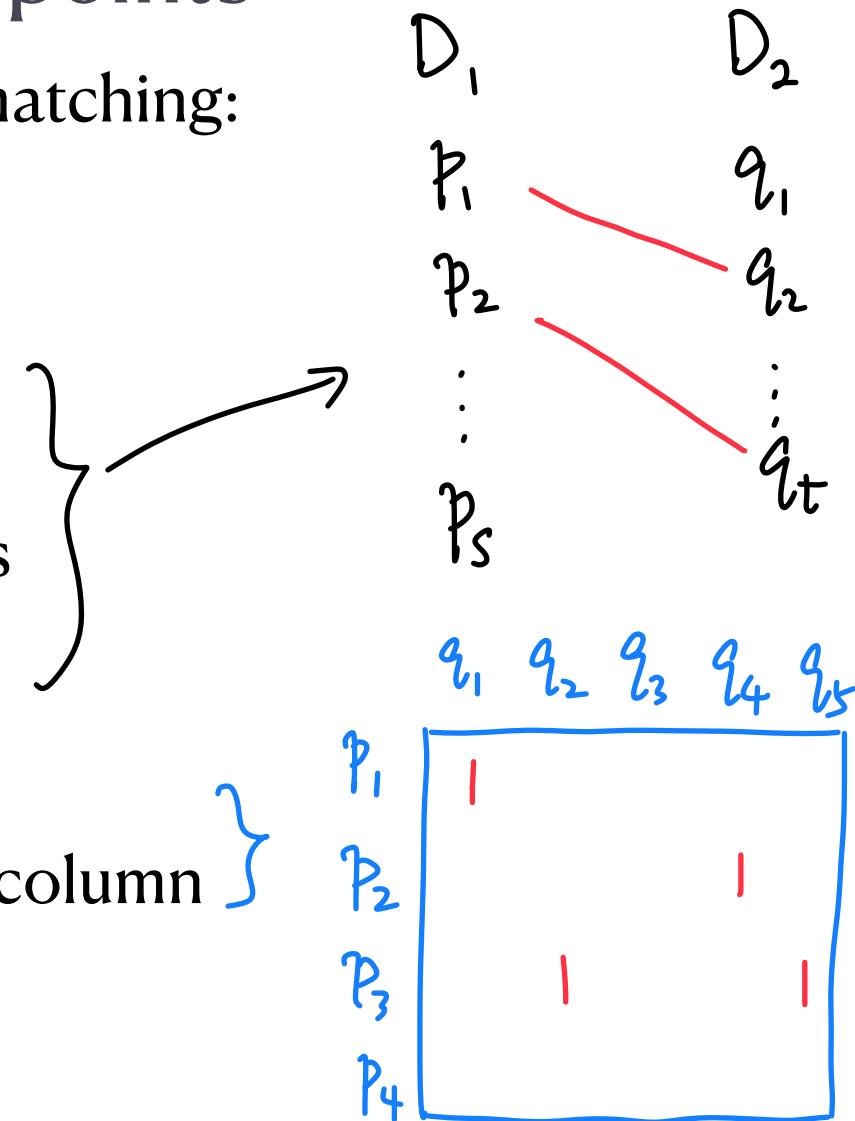
Examples of partial matchings:



Partial-matching: how to match points

- ▶ Different ways of understanding partial matching:

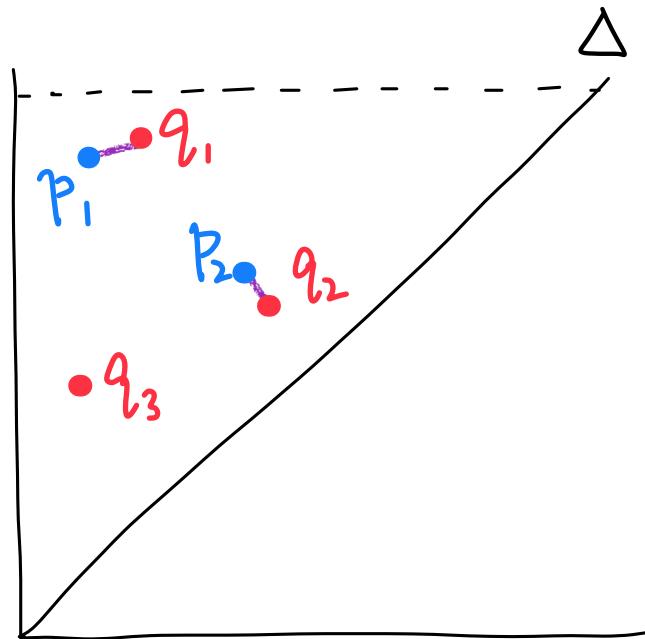
- ▶ subset of $D_1 \times D_2$
- ▶ bipartite graph
 - ▶ two lists of nodes
 - ▶ edges connect nodes in different lists
 - ▶ every node has at most one edge
- ▶ binary matrix
 - ▶ at most one 1 in every row and every column



Cost of partial-matching

- The cost of a partial matching $M \subset D_1 \times D_2$ is

$$cost(M) = \max \left(\max_{(p,q) \in M} \|p - q\|_\infty, \max_{p \text{ unmatched}} \|p - \Delta\|_\infty \right)$$



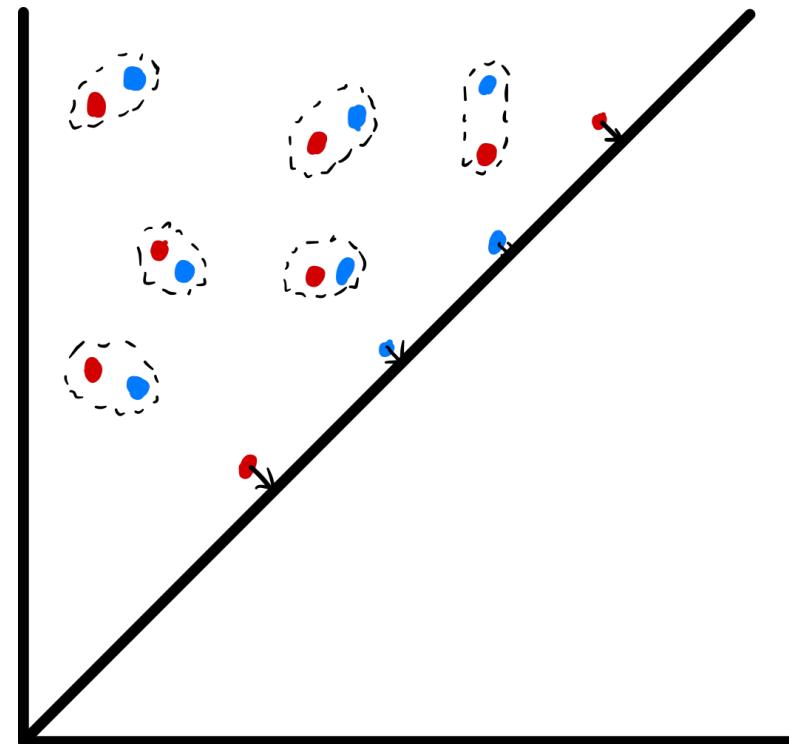
$$D_1 = \{p_1, p_2\} \quad D_2 = \{q_1, q_2, q_3\}$$

$$M = \{(p_1, q_1), (p_2, q_2)\}$$

$$\begin{aligned} cost(M) = \max \{ & \|p_1 - q_1\|_\infty, \\ & \|p_2 - q_2\|_\infty, \\ & \|\Delta - q_3\|_\infty \} \end{aligned}$$

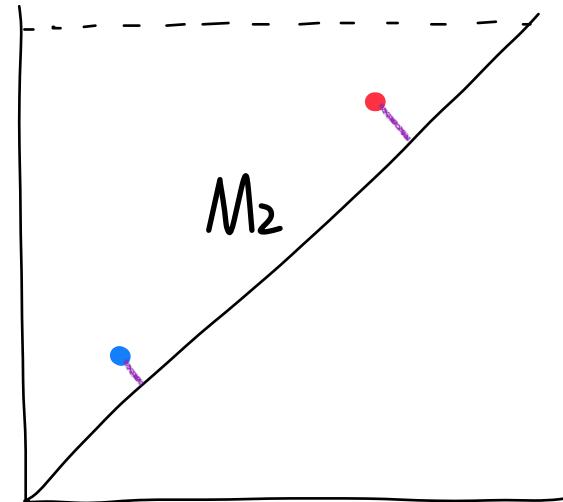
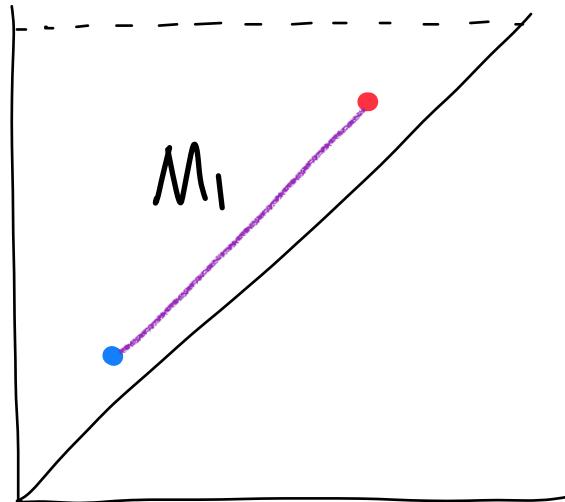
Bottleneck distance: optimal partial-matching

- ▶ [Cohen-Steiner, Edelsbrunner, Harer, DCG 2007]
- ▶ The bottleneck distance between D_1 and D_2 is
 - ▶ $d_B(D_1, D_2) := \min\{cost(M) \mid M \subset D_1 \times D_2 \text{ a partial matching}\}$



Bottleneck distance between 1-point PDs

- ▶ Assume that $D = \{p\}$ and $D' = \{q\}$ with $p=(b,d)$ and $q=(b', d')$.
- ▶ There are only two possible partial matchings:
 - ▶ $M_1 = \{(p, q)\}$ with $\text{cost}(M_1) = \|p - q\|_\infty$
 - ▶ $M_2 = \emptyset$ with $\text{cost}(M_2) = \max(\|p - \Delta\|_\infty, \|q - \Delta\|_\infty)$
- ▶ Thus, $d_B(D, D') = \min \left(\max(|b - b'|, |d - d'|), \max \left(\frac{|b - d|}{2}, \frac{|b' - d'|}{2} \right) \right)$



Bottleneck distance is an extended metric

- ▶ The bottleneck distance defines a metric on the set of PDs:
 - ▶ $d_B(D, D') = 0$ iff $D = D'$
 - ▶ $d_B(D, D') = d_B(D', D)$
 - ▶ $d_B(D, D') \leq d_B(D, D'') + d_B(D'', D')$
- ▶ We call d_B an *extended* metric, because it can take value ∞

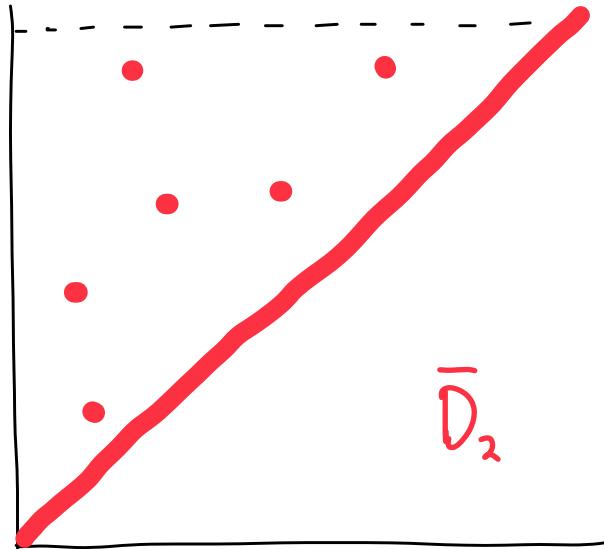
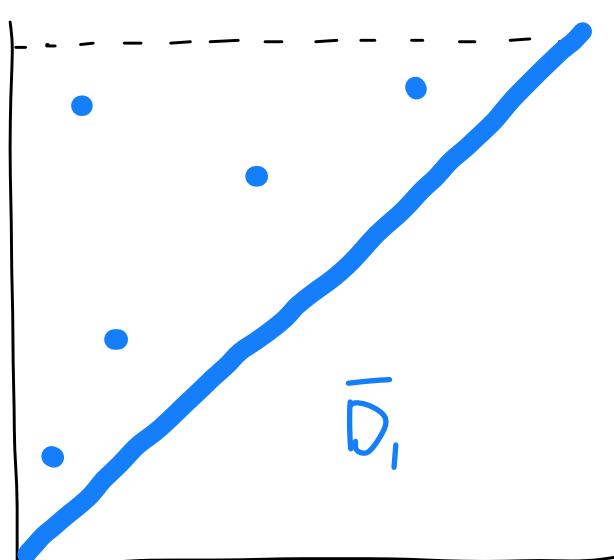
Example $D = \{(\underline{0}, \underline{\infty})\}_{\Leftrightarrow}^p, D' = \emptyset.$

The only partial-matching is $M = \emptyset$

$$\Rightarrow d_B(D, D') = \text{cost}(M) = \|p - \Delta\|_\infty = \frac{\infty - 0}{2} = \infty$$

Alternative formulation of the bottleneck distance

- Given two persistence-diagrams (multiset of points in $(\mathbb{R} \cup \{\infty\})^2$)
 - $D_1 = \{p_1, p_2, \dots, p_s\}$ and $D_2 = \{q_1, q_2, \dots, q_t\}$
- Augment $\bar{D}_1 := D_1 \cup \Delta^\infty$ and $\bar{D}_2 := D_2 \cup \Delta^\infty$
 - where Δ^∞ is the diagonal points each with **infinite multiplicity**



- When PDs have different cardinalities, we cannot talk about bijections.
- Adding the extra copies of diagonal points forces the same cardinality.

Alternative formulation of the bottleneck distance

- ▶ A partial-matching between D_1 and D_2 is a **bijection** $\bar{M} \subset \bar{D}_1 \times \bar{D}_2$ with

$$cost(\bar{M}) := \max_{(p,q) \in \bar{M}} \|p - q\|_\infty$$

For any p , let $\tilde{p} \in \Delta$ be p 's closest point in Δ

$$\begin{aligned} \forall M \subset D_1 \times D_2, \text{ define } \bar{M} := M \cup & \left\{ (p, \tilde{p}) : p \in D_1, \text{ unmatched} \right\} \\ & \cup \left\{ (\tilde{q}, q) : q \in D_2, \text{ unmatched} \right\} \\ & \cup \left\{ (\tilde{p}, \tilde{p}) : \tilde{p} \in \Delta \right\} \end{aligned}$$

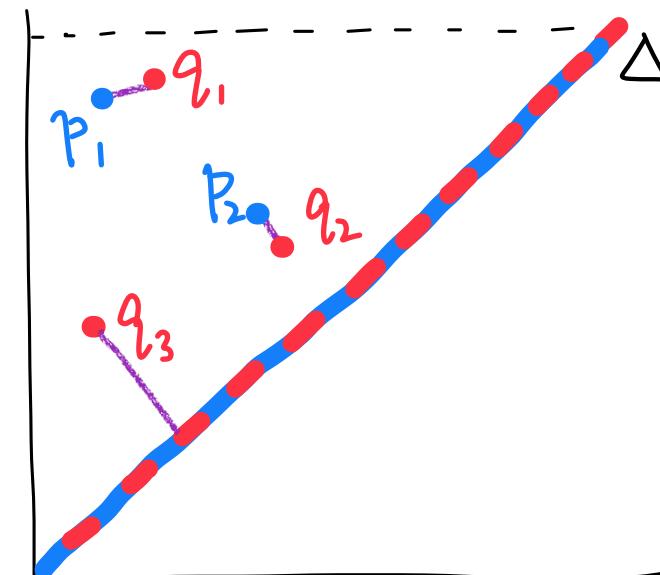
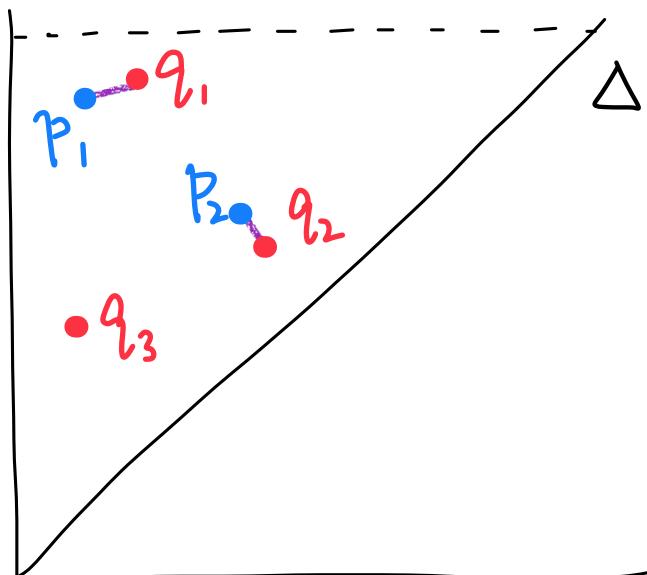
$$\forall \bar{M} \subset (D_1 \cup \Delta^\infty) \times (D_2 \cup \Delta^\infty), \text{ define } M = \{(p, q) \in \bar{M} \mid p \notin \Delta, q \notin \Delta\}$$

Alternative formulation of the bottleneck distance

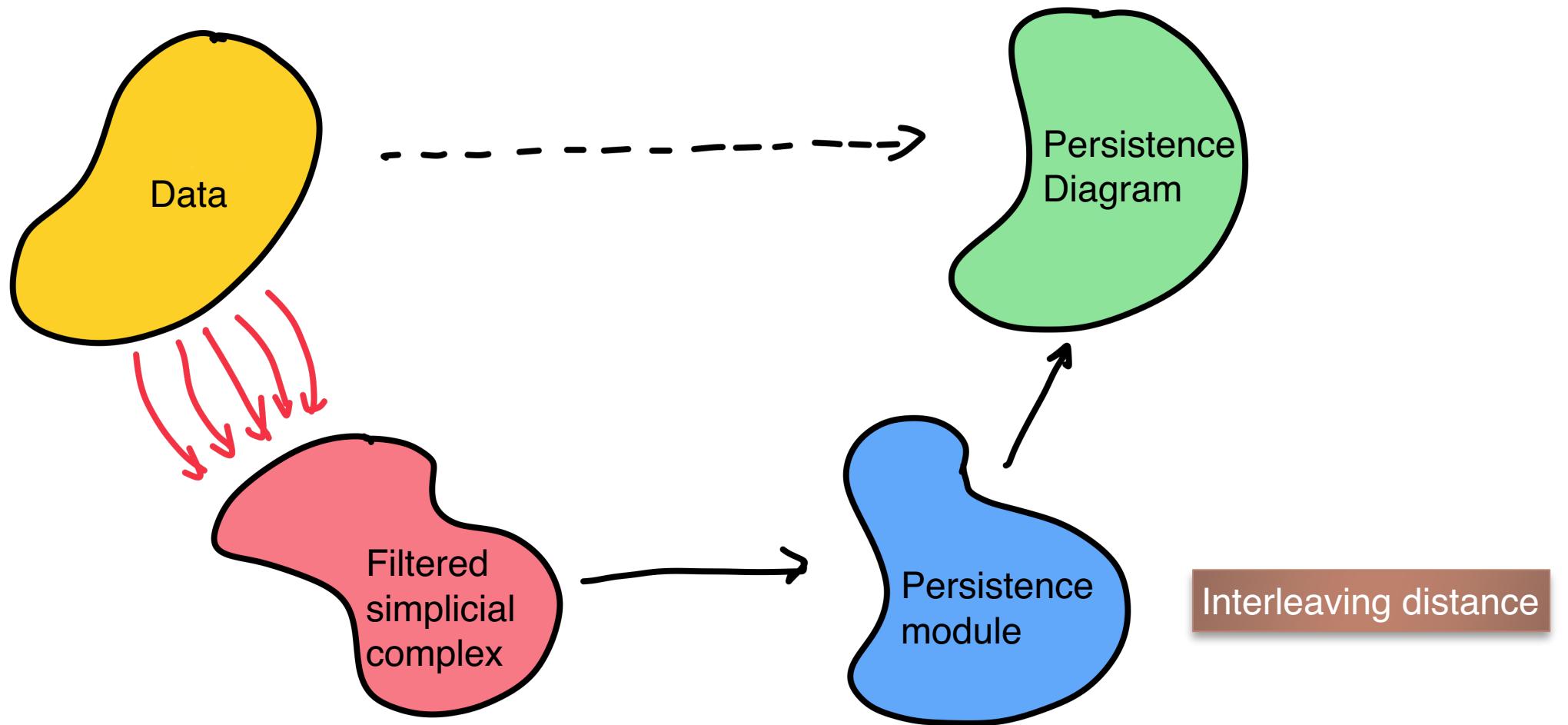
- The **bottleneck distance** between D_1 and D_2 is

$$d_B(D_1, D_2) := \min\{cost(\bar{M}) \mid \bar{M} \subset \bar{D}_1 \times \bar{D}_2 \text{ a bijection}\} = \min_{\bar{M}} \max_{(p,q) \in \bar{M}} \|p - q\|_\infty$$

$$\begin{array}{c} \left\{ \begin{array}{l} \text{partial-matchings} \\ M \subset D_1 \times D_2 \end{array} \right\} \xleftrightarrow{1-1} \left\{ \begin{array}{l} \text{bijections} \\ \bar{M} \subset \bar{D}_1 \times \bar{D}_2 \end{array} \right\} \end{array}$$

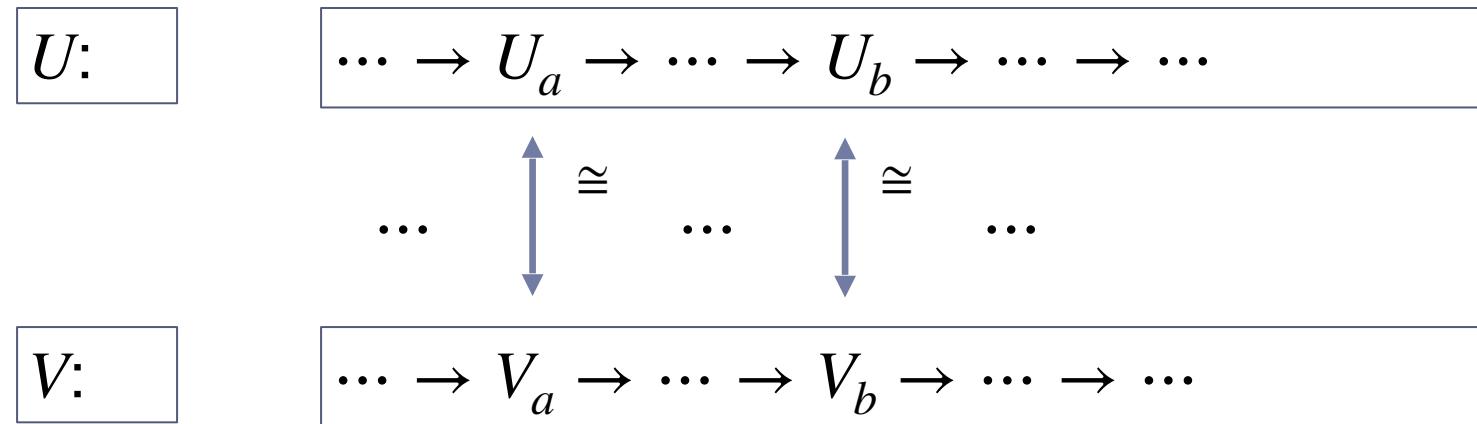


Interleaving distance between
Persistence Modules



Intuition

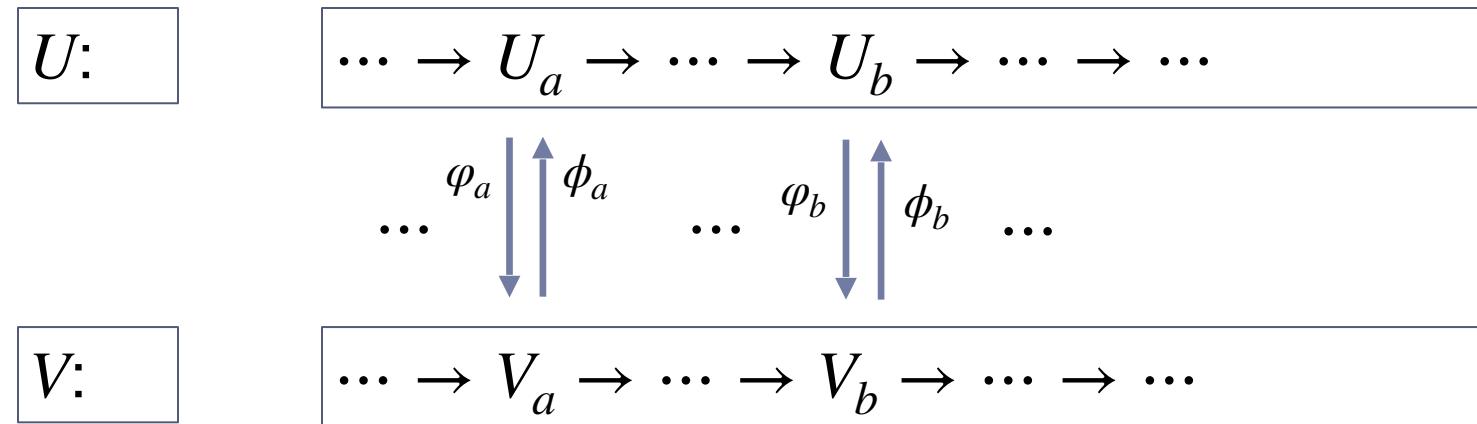
- ▶ Isomorphic persistence modules



- ▶ Vertical maps also have to commute with horizontal maps (in all possible combinations)

Intuition

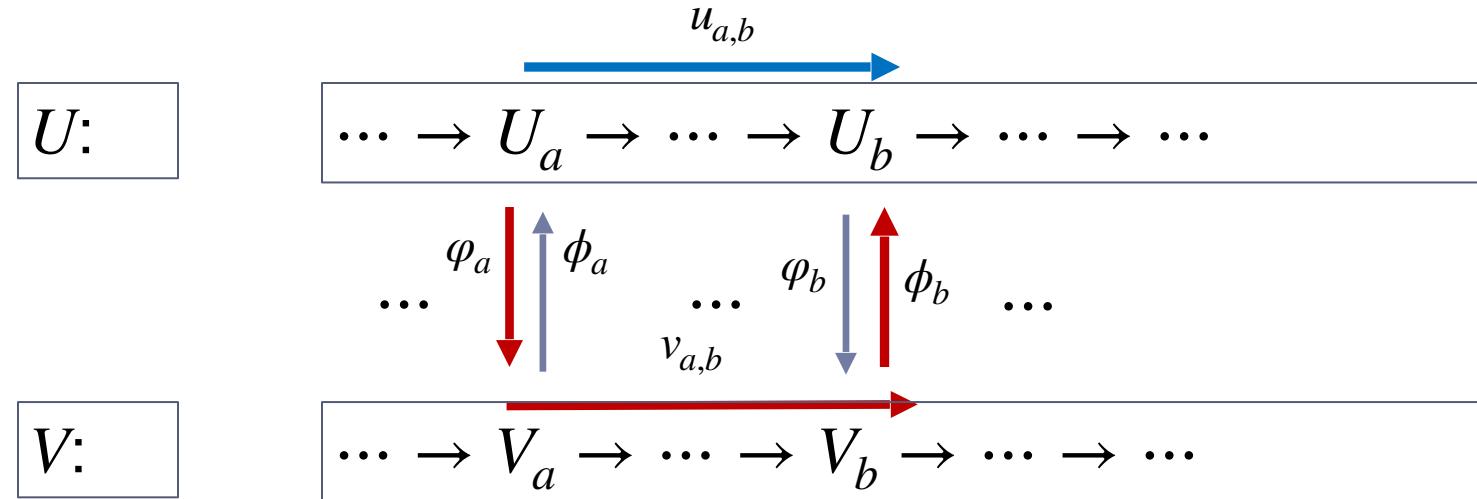
- ▶ Isomorphic persistence modules



- ▶ Vertical maps also have to commute with horizontal maps (in all possible combinations)

Intuition

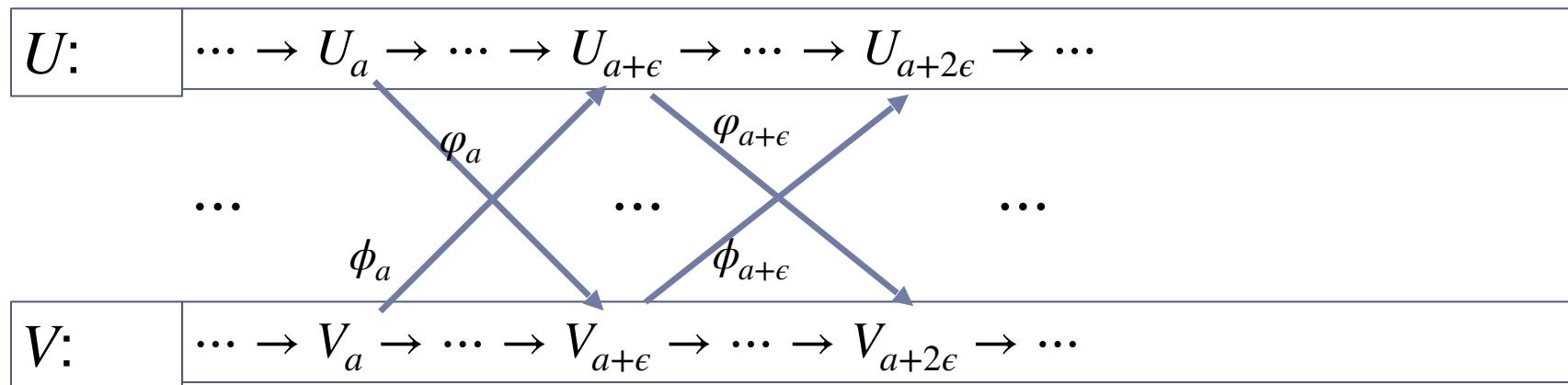
- ▶ Isomorphic persistence modules



- ▶ Vertical maps also have to commute with horizontal maps (in all possible combinations)

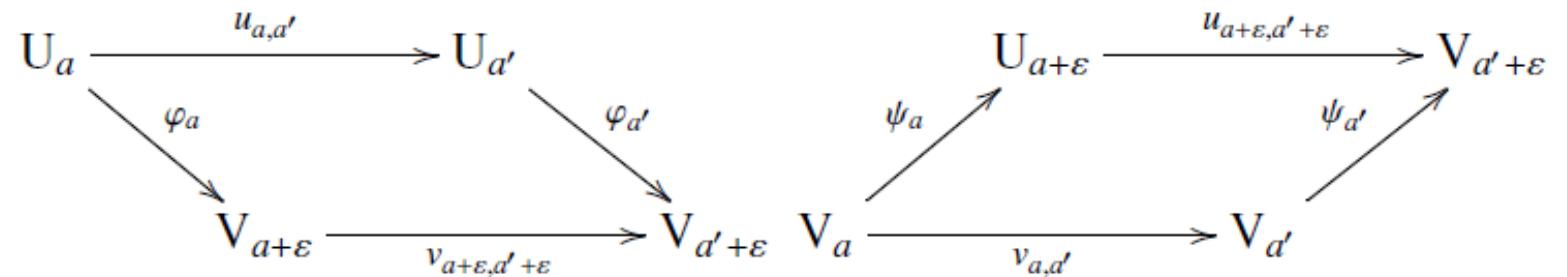
ϵ -Interleaving

- ▶ U and V are **ϵ -interleaved** if there exists maps
 - ▶ $\varphi_a : U_a \rightarrow V_{a+\epsilon}$ and $\phi_a : V_a \rightarrow U_{a+\epsilon}$ for any $a \in \mathbb{R}$
 - ▶ s.t. these maps commute with horizontal maps u 's and v 's
- ▶ The maps φ_a and ϕ_a are called an **ϵ -interleaving**

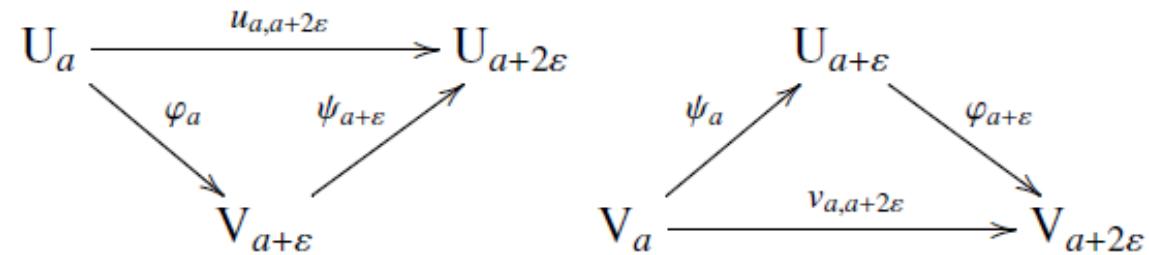


ϵ -Interleaving

- ▶ U and V are **ϵ -interleaved** if there exists maps
 - ▶ $\varphi_a : U_a \rightarrow V_{a+\epsilon}$ and $\phi_a : V_a \rightarrow U_{a+\epsilon}$ for any $a \in \mathbb{R}$
 - ▶ s.t. these maps commute with horizontal maps u 's and v 's



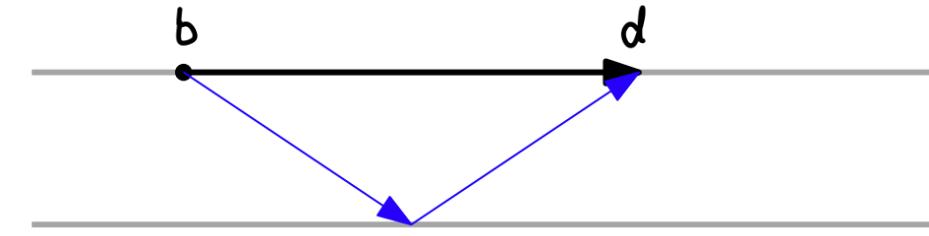
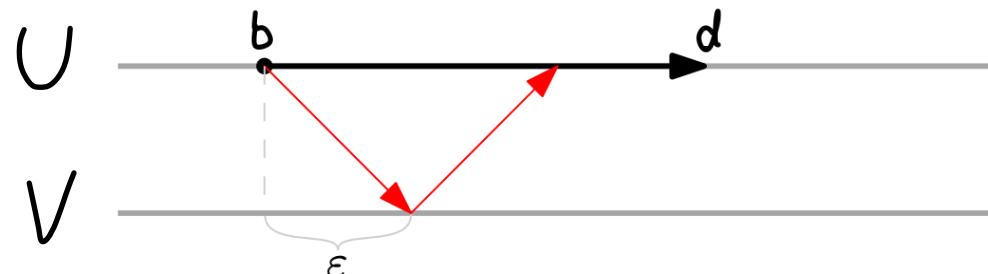
- ▶ To verify commutativity of maps, only need to check four configurations:



ϵ -Interleaving

- ▶ U and V are **ϵ -interleaved** if there exists maps
 - ▶ $\varphi_a : U_a \rightarrow V_{a+\epsilon}$ and $\phi_a : V_a \rightarrow U_{a+\epsilon}$ for any $a \in \mathbb{R}$
 - ▶ s.t. these maps commute with horizontal maps u 's and v 's

Example $U = \mathbb{F}[b, d)$, $V = \emptyset$ Let $\varphi_a = 0$ & $\phi_a = 0$



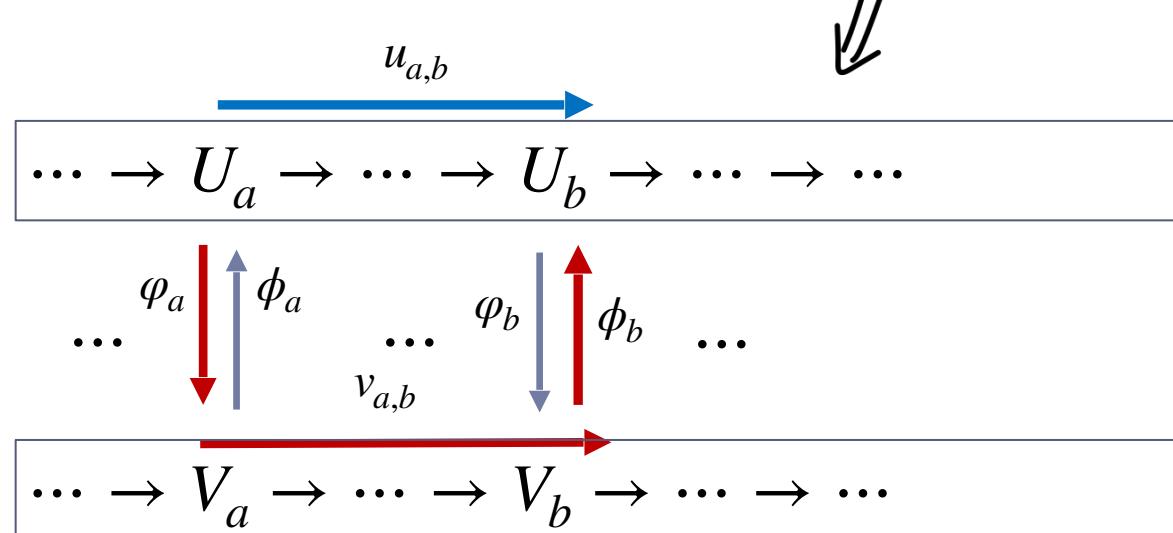
NOT ϵ -interleaved for $\epsilon < \frac{d-b}{2}$

ϵ -interleaved for $\epsilon \geq \frac{d-b}{2}$

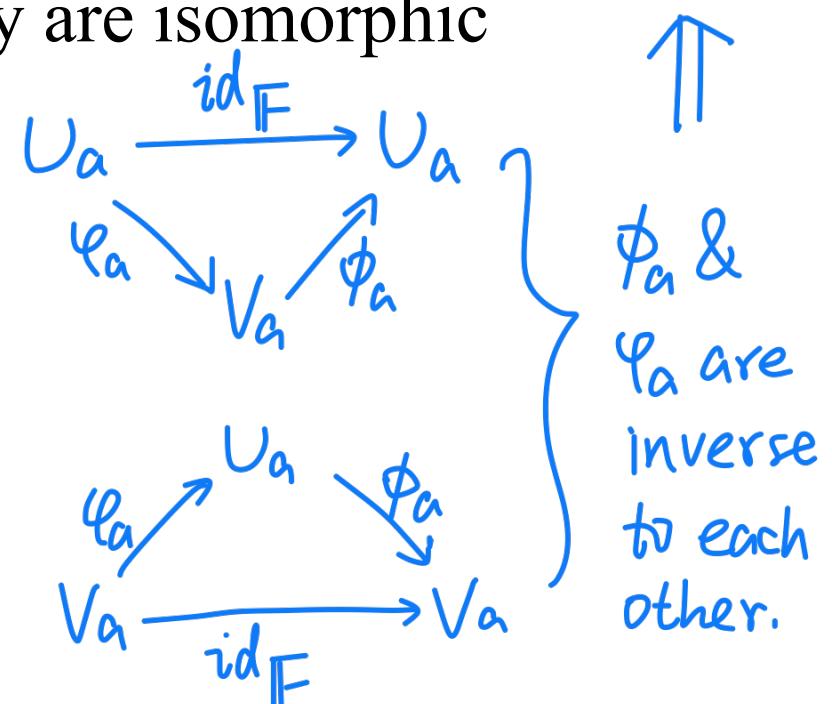
ϵ -Interleaving

- ▶ U and V are **ϵ -interleaved** if there exists maps
 - ▶ $\varphi_a : U_a \rightarrow V_{a+\epsilon}$ and $\phi_a : V_a \rightarrow U_{a+\epsilon}$ for any $a \in \mathbb{R}$
 - ▶ s.t. these maps commute with horizontal maps u 's and v 's
- ▶ Proposition: If U and V are 0-interleaved, then they are isomorphic

$U:$



$V:$



Interleaving Distance

- ▶ The **interleaving distance** between two persistence modules V and U is
$$d_I(V, U) := \inf\{\epsilon > 0 \mid V \text{ and } U \text{ is } \epsilon\text{-interleaved}\}$$
- ▶ It is an extended pseudo-metric
 - ▶ It is symmetric and it satisfies triangle inequality
 - ▶ $d_I(U, W) \leq d_I(U, V) + d_I(V, W)$
 - ▶ (extended) Can take value ∞
 - ▶ (pseudo) Non isomorphic persistence modules can have zero distance

Interleaving Distance

- ▶ The **interleaving distance** between two persistence modules V and U is
 $d_I(V, U) := \inf\{\epsilon > 0 \mid V \text{ and } U \text{ is } \epsilon\text{-interleaved}\}$

Isometry Theorem [Lesnick 2015], [Chazal, de Silva, Gliss and Oudot, 2016]

Given two finitely represented persistence modules U and V , let D_U and D_V be their corresponding persistence diagrams. We then have:

$$d_B(D_U, D_V) = d_I(U, V)$$