

TDA - fall 2025

Persistence
implementations



Last time: algorithm

$$R = B$$
for $j = 1 \dots m$ do

while $\exists j' < j$ with $low(j') = low(j)$ **do**

add column j' to column j

end while

end for

[illegible][illegible][illegible]

Complexity

With m simplices, matrix has size $m \times m$

```
R = B
for j = 1 ... m do
  while  $\exists j' < j$  with  $\text{low}(j') = \text{low}(j)$  do
    add column  $j'$  to column  $j$ 
  end while
end for
```

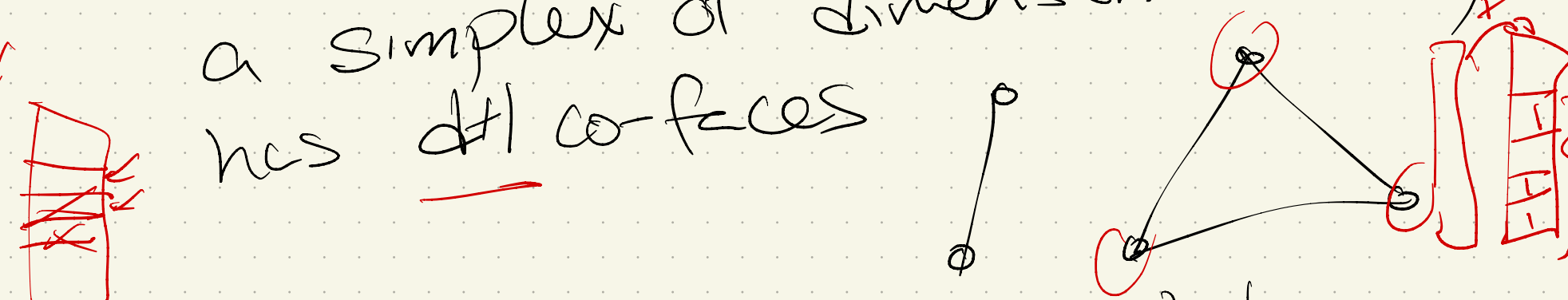
repeats m times

→ while loop:

each such operation must move $\text{low}(j)$ up, but other columns can become 1 during operation

Worst case: $O(m)$ additions to cancel a row
each "add" takes $O(m)$ $\Rightarrow O(m^3)$

Improving this runtime

- ① Note that the matrix is sparse:
a simplex of dimension d only
has $d+1$ co-faces
- 

So compressed representation helps in
prachce

- ② The algorithm can also be reduced
to Gaussian elimination

$\hookrightarrow O(n^w)$ time, $w = [2, 2.373]$

More speedups

Matrix algorithm last time:

- Sweep columns left to right
- every addition to col j moves $\text{low}[j]$ higher
- If additive, will 0 it out
↳ might take many column ops
- But, once we zero it out, it's never used again!

Practical improvement

Bauer et al 2014

Process filtration in backwards order

(highest dam first)

Then for pair (σ^{p-1}, σ^p)

\uparrow
 $\underbrace{\text{In } \delta_{p-1}}$
birth

\uparrow
 $\underbrace{\text{In } \delta_p}$
death

Result: can skip earlier column
(since we know it will be 0's)

Another method: Collapses

Boissonnat et al 2018

A simplicial cone for a complex L and a vertex a not in L is

$$aL = \{ \tau \mid \tau \in L \text{ or } \tau = \sigma \cup a, \sigma \in L \}$$

A vertex is dominated if $\text{Ink}(v)$

is a simplicial cone:

$\exists v' \neq v$ and $L \subseteq K$ s.t. $\text{Ink}_K(v) = v' L$

collapse:

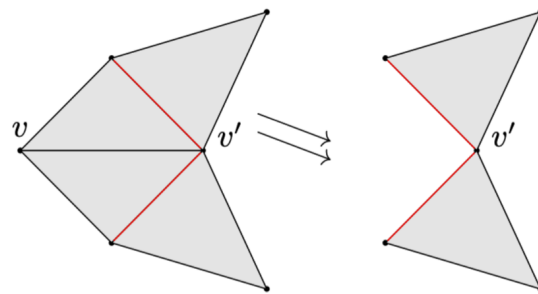


Figure 1 Illustration of an *elementary strong collapse*. In the complex on the left, v is dominated by v' . The link of v is highlighted in red. Removing v leads to the complex on the right.

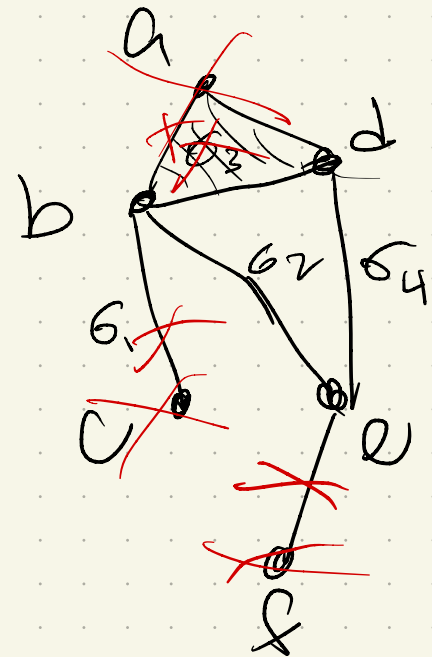
If we collapse all possible vertices,
get core K^C , which is unique
up to isomorphism & has same
homotopy type.
↳ via a retract

Can compute via matrix:
maximal simplices

	σ_1	σ_2	σ_3	σ_4	σ_5
a	0	0	1	0	0
b	1	1	1	0	0
c	1	0	0	0	0
d	0	0	1	1	0
e	0	1	0	1	1
f	0	0	0	0	1

	b	d	e
σ_1	1	0	0
σ_2	1	0	1
σ_3	1	1	0
σ_4	0	1	1
σ_5	0	0	1

	σ_2	σ_3	σ_4
b	1	1	0
d	0	1	1
e	1	0	1



Aside:

Lots of work on speeding up
persistence

↳ and on finding lower bounds:
cases where you really need
 $O(n^w)$ time.

Not mentioned: parallelization

The workflow so far

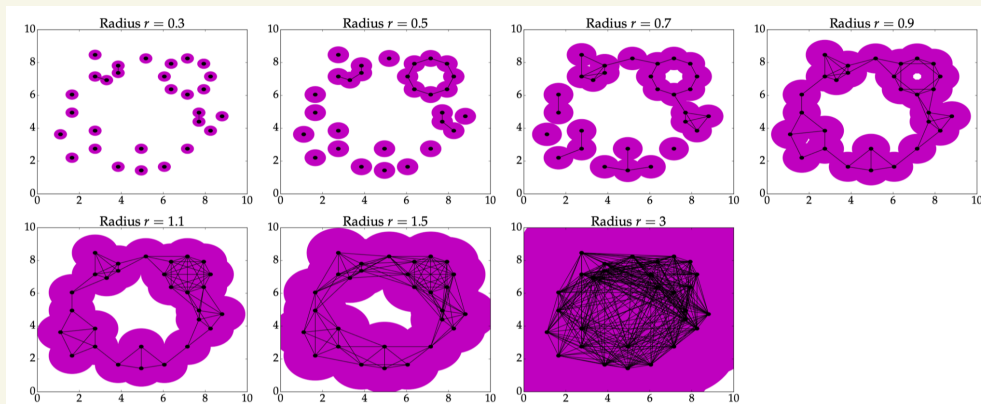
Build a filtration F from a simplicial complex

↳ usually parameterized by a function f , via sublevel sets

Example: $P \subseteq \mathbb{R}^n$

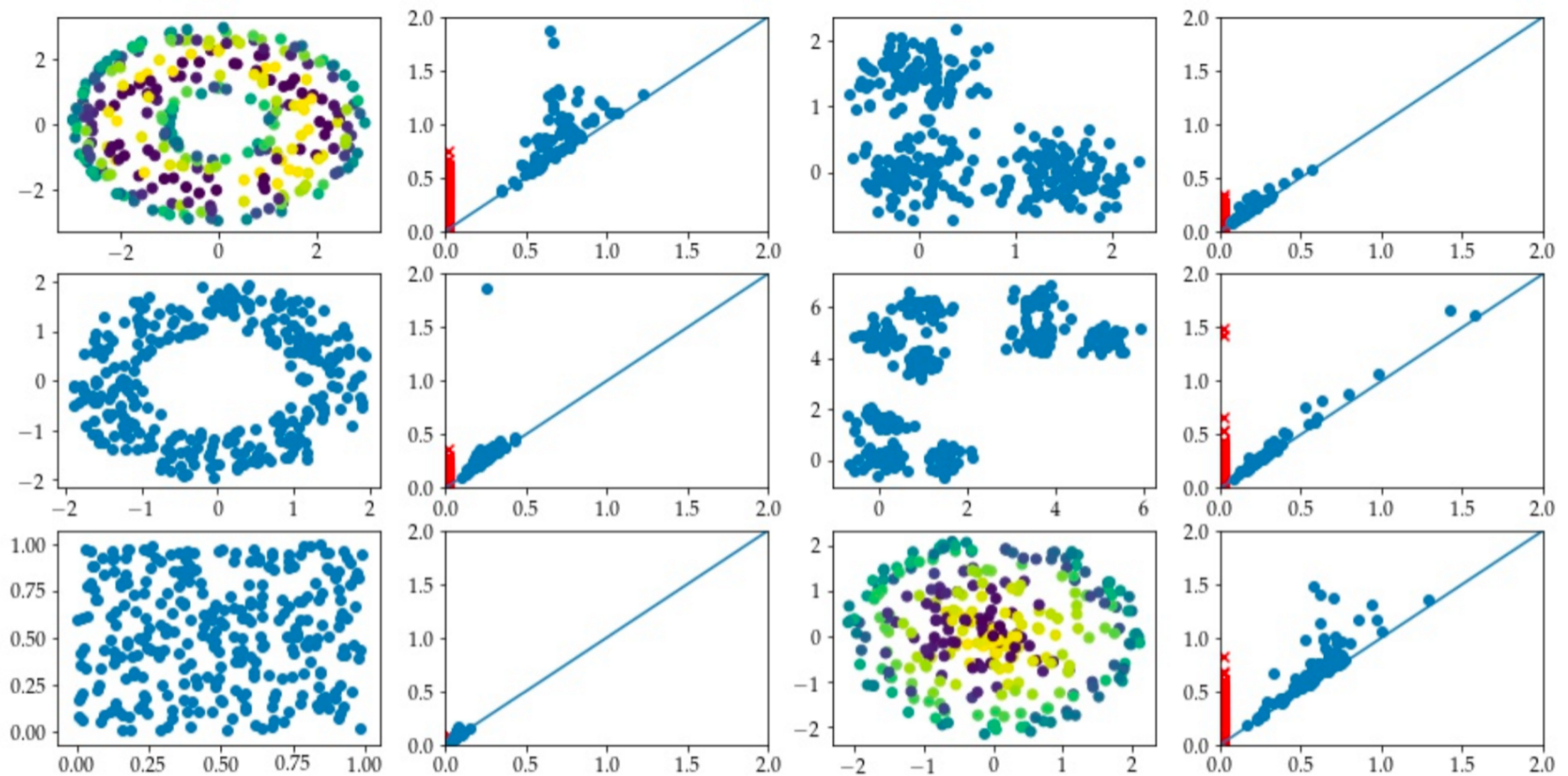
Build Rips filtration:

for $0 \leq r_1 \leq \dots \leq r_k$, $K_i = VR(P, r_i)$



Persistence diagram

Result: H_0 and H_1



... now what?

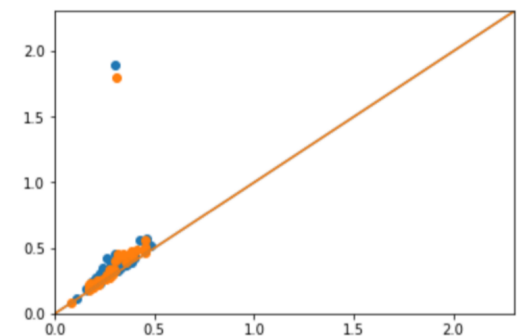
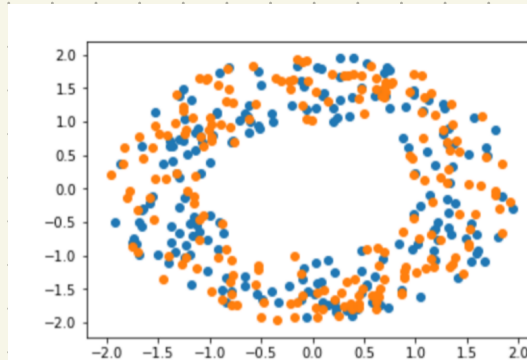
Distance measures

A distance on a set X is a function
 $d: X \times X \rightarrow \mathbb{R}_{\geq 0}$ s.t. $\forall x, y, z \in X$

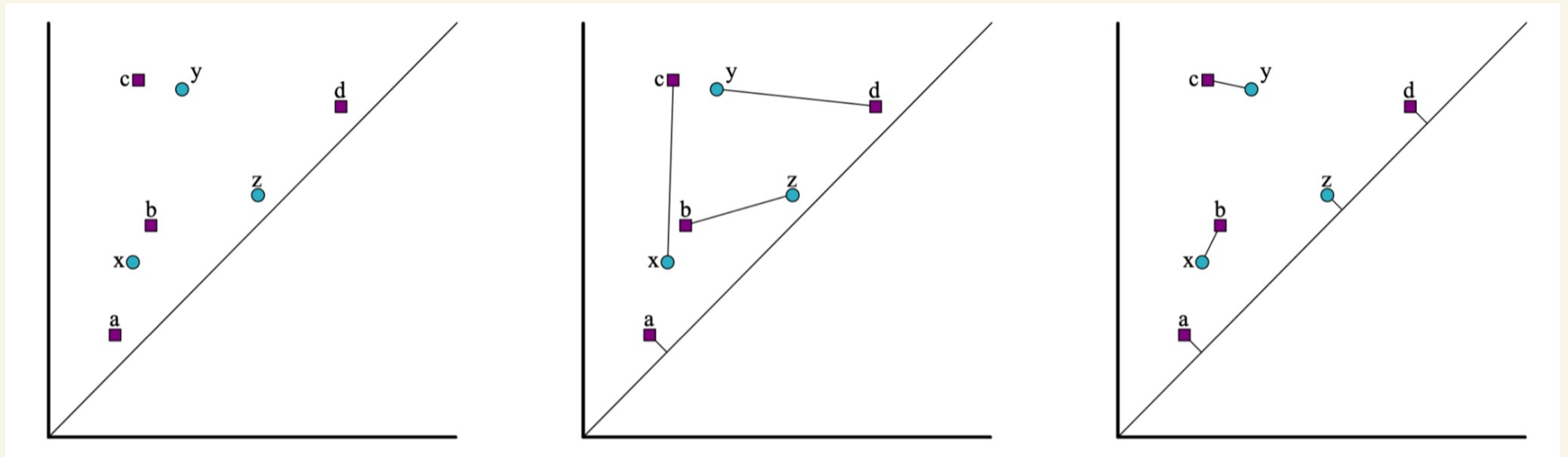
- $d(x, y) \geq 0$ & $d(x, y) = 0 \Leftrightarrow x = y$
- $d(x, y) = d(y, x)$
- $d(x, z) \leq d(x, y) + d(y, z)$

Our goal: distances for PDs

$$X = \{D_{\text{gm}} F(K)\}_{F, K}$$



How to quantify 'nearby' here?

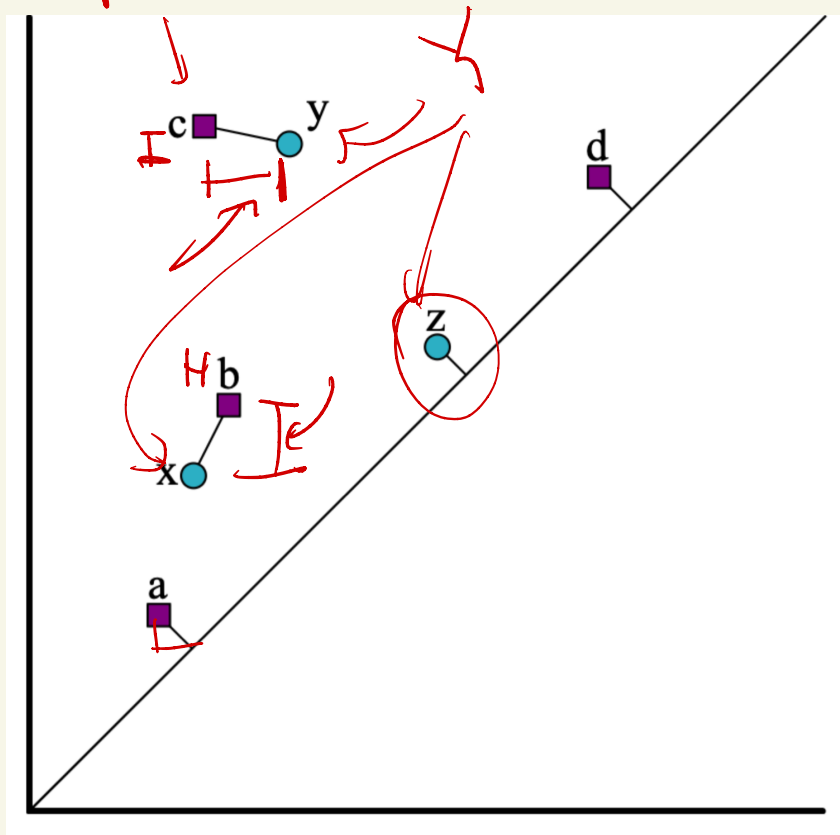


Major goal: Stability

Bottleneck distance (take 1)

Given 2 diagrams $X, Y \subseteq \mathbb{R}^2$,

$$d_B(X, Y) = \inf_{\varphi: X \rightarrow Y} \sup_{x \in X} \|x - \varphi(x)\|_\infty$$



Here!

$$\varphi(a) = \text{diagonal}$$

$$\varphi(b) = x$$

$$\varphi(c) = y$$

$$\varphi(d) = \text{diagonal}$$

Alternate definition

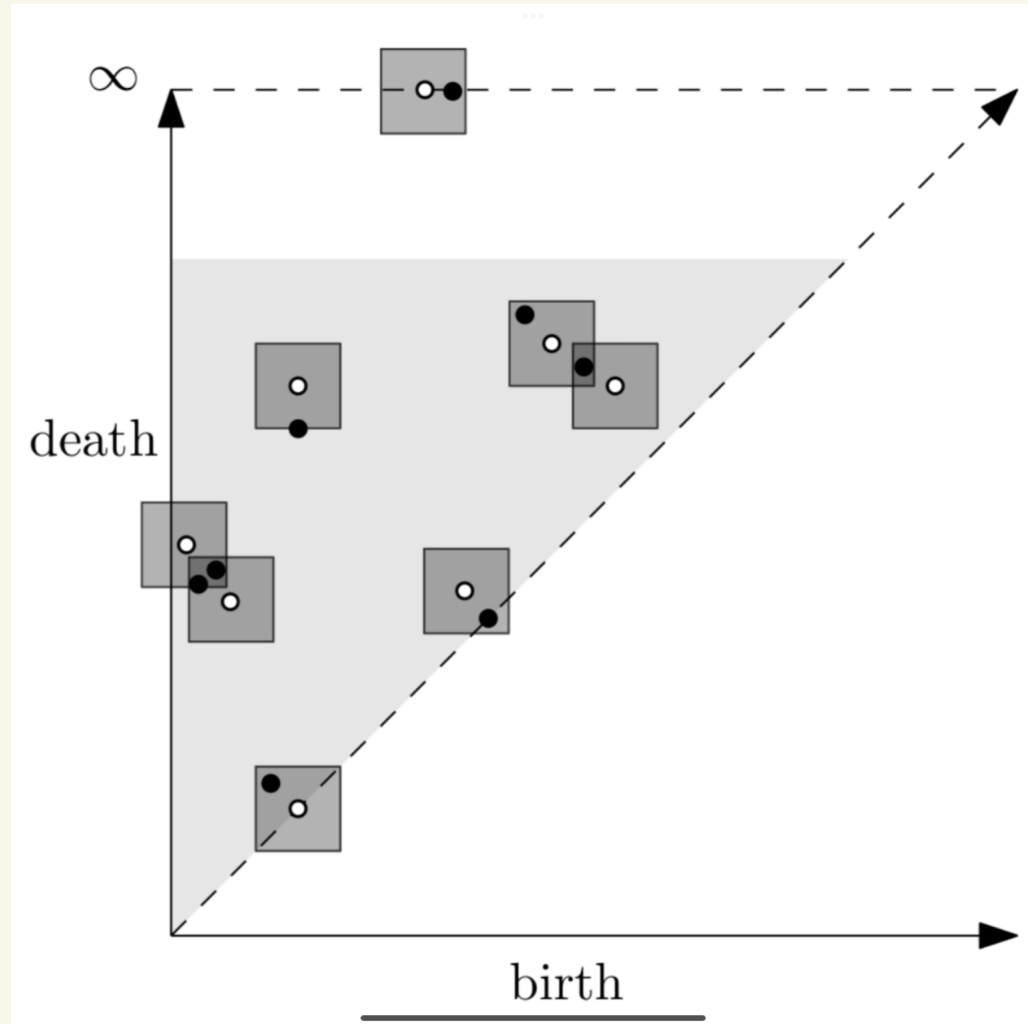
A matching between X & Y is a bijection on a subset of the off-diagonal points of the 2 diagrams, $X' \subseteq X$ & $Y' \subseteq Y$.

Cost of matching:

$$C(M) = \sup \left\{ \|x - M(x)\|_{\infty} \mid x \in X' \right\} \cup \left\{ \frac{1}{2} |x_1 - x_2| \text{ s.t. } (x_1, x_2) \in X \setminus X' \cup Y \setminus Y' \right\}$$

$$\text{Bottleneck } d_B(X, Y) = \inf_M C(M)$$

Another view: L_∞ balls in \mathbb{R}^2

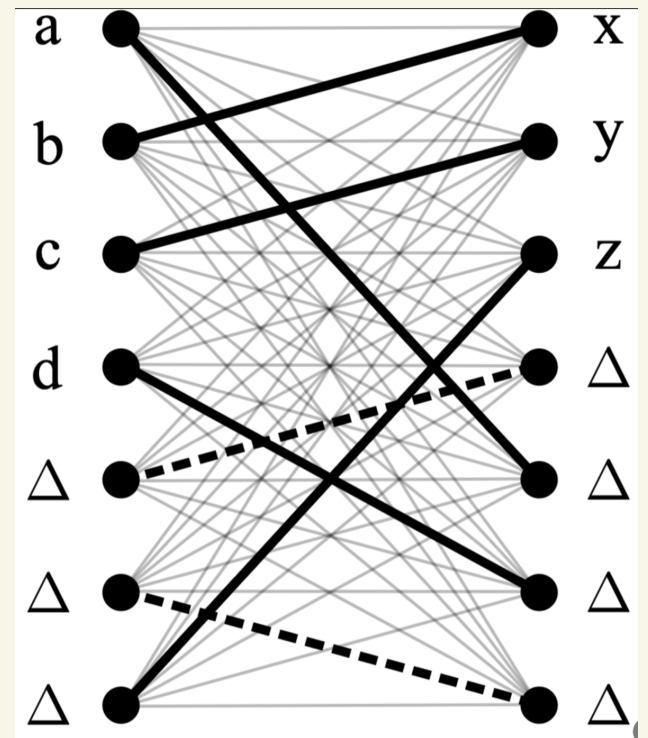
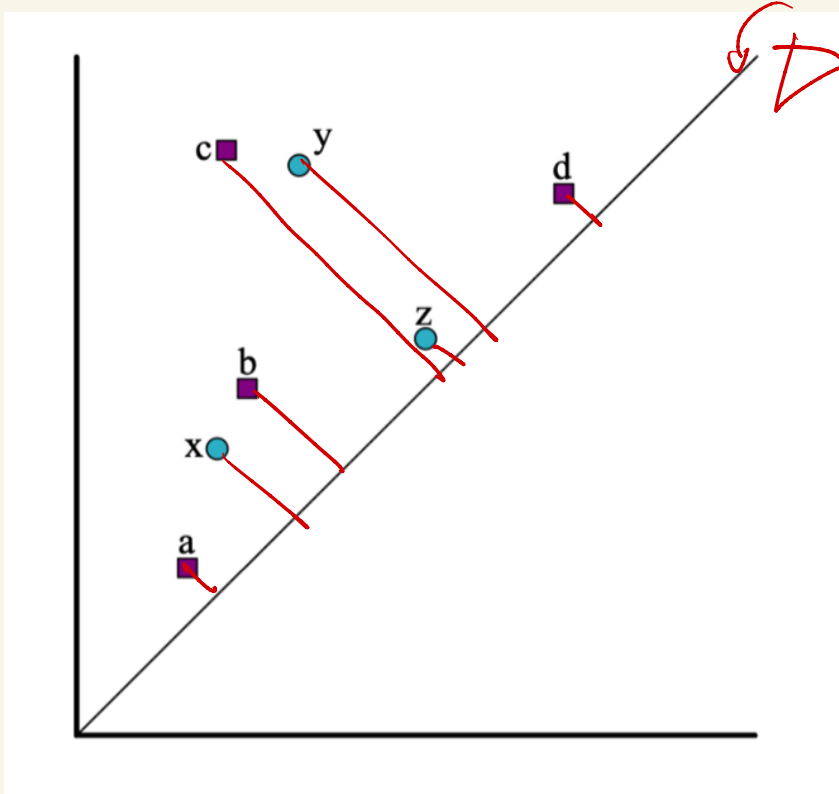


Min ϵ s.t.
 ϵ balls around
every point
either:

- includes at least 1 point from other set
- touches the diagonal

How to compute?

Reduce to a graph problem



Min cost matchings: use network flow
on graph $\leadsto O((\# \text{ per points})^2)$