

# From Geometry to Topology: Inverse Theorems for Distributed Persistence - Solomon, Wagner, Bendich

◦ Given a statistic  $\lambda$  of a finite pt cloud

$X \subset \mathbb{R}^d$ , what does computing  $\lambda$  on

Subsets of  $X$  tell us?

Definition ◦ For  $k \in \mathbb{N}$  we define

$$\lambda_k := \left\{ (S, \lambda(S)) \mid S \subseteq X, |S| = k \right\}$$

◦ we say  $\lambda$  is  $k$ -distributed if  $\lambda_k(X)$  determines  $\lambda(X)$  for any  $|X| \geq k$ .

Example

$$\begin{aligned} \text{Consider } \lambda(S) &:= \text{diameter}(S) \\ &= \max_{x, y \in S} d(x, y) \end{aligned}$$

then we see that  $\lambda$  is  $k$ -distributed

for any  $k \geq 2$ :

$$\lambda(X) = \max \lambda_k(X)$$

Moreover,  $\lambda_k$  may contain more information — here  $\lambda_2$  is the entire distance matrix.

- ① Are any TDA statistics  $\lambda$   $k$ -distributed?
- ② Does  $\lambda_k$  contain more geometric information than  $\lambda$ ?
- ③ Do we really need to compute  $\lambda$  on all subsets of size  $k$ ?

They consider

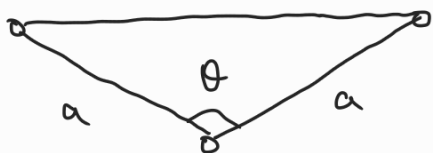
$$\lambda = \begin{array}{ll} \text{VR persistence} & (\text{RP}) \\ \text{VR Euler curve} & (\text{RE}) \\ \text{Čech Persistence} & (\text{CP}) \\ \text{Čech Euler curve} & (\text{CE}) \end{array} .$$

proposition • Each of these  $\lambda$  is  $\mathbb{Z}$ -distributed.

Moreover,  $\lambda_2(X)$  determines the distance matrix of  $X$  and thus  $X$  up to isometry.

$\lambda_3$  does not determine  $X$  up to isometry.

proof:



RP



RP

$\lambda_2$  contains more geometric information than even  $\lambda_3$ . Can we quantify the change from increasing  $k$ ?

### Definition

$\lambda$  is  $(k_1, \dots, k_r)$ -distributed

if  $\lambda_{k_1}, \dots, \lambda_{k_r}$  determine  $\lambda$ .

For any of the four  $\lambda$  considered, let  $\lambda^m(X)$  denote  $\lambda$  ( $m$ -skeleton of  $X$ ).

## Theorem

For any of the four  $\lambda$  we have that  $\lambda^m$  is  $(k, k-1, \dots, k-m-1)$ -distributed for all  $k \geq m+1 \geq 2$ .

Moreover,  $\{\lambda_k^m(X), \dots, \lambda_{k-m-1}^m(X)\}$  determine  $X$  up to isometry.

If the  $\lambda_k^m(X)$  are 'close' to  $\lambda_k^m(Y)$  does that mean the geometry of  $X$  is 'close' to that of  $Y$ ?

Definition  $\circ \phi: (X, d_X) \rightarrow (Y, d_Y)$  is

an  $\varepsilon$ -quasi isometry if we have

that

$$|d_X(x_1, x_2) - d_Y(\phi(x_1), \phi(x_2))| \leq \varepsilon$$

for all  $x_1, x_2 \in X$ .

## Theorem

Let  $\lambda = \mathbb{R}P$  or  $\mathbb{C}P$  and take  $k > m > 0$ .

Let  $\phi: X \rightarrow Y$  be a bijection s.t.

$\forall S \subseteq X$  with  $|S| \in \{k, k-1, \dots, k-m-1\}$

we have

$$d_{\text{Bottleneck}}(\lambda^m(S), \lambda^m(\phi(S))) \leq \varepsilon.$$

Then

$$\phi \text{ is a } \begin{cases} 112 k^2 \varepsilon - QI & \text{if } \lambda = \mathbb{R}P \\ 224 S(k, m) k^{m+1} \varepsilon - QI & \text{if } \lambda = \mathbb{C}P \end{cases}$$

$$\text{where } S(k, m) = \binom{k}{2} + \binom{k}{3} + \dots + \binom{k}{m+1}.$$

So increasing  $k$  interpolates between capturing geometric info and capturing topological info.

In fact, a tighter bound linear in  $k$  is available by considering  $w(\lambda'(x), \lambda'(\phi(x)))$ .

This seems expensive. Do we need to compute  $\lambda$  on every subset?

Proposition The conditions of the previous theorem can be weakened. It suffices that

$$d_B(\lambda^m(S), \lambda^m(\phi(S))) \leq \epsilon$$

for all  $S \in C$ , where  $C$  is any collection of subsets of  $X$  satisfying:

• (covering)

$$\forall \sigma \subseteq X \text{ with } |\sigma| \leq 2 \quad \exists S \in C, |S| = k \\ \text{s.t. } \sigma \subseteq S.$$

• (closure)

$\forall S \in C$  s.t.  $|S| = k$  we have that

$$\forall S' \subseteq S \text{ with } |S'| \geq k - m - 1, \text{ then } S' \in C.$$

Moreover, we can bound the probability that randomly sampled subsets will satisfy

the covering property. (we can just fill in the necessary subsets to satisfy closure).

Proposition let  $|X| = n$  and choose  $M$

subsets  $S_1, \dots, S_M$  of size  $k$  by uniform sampling without replacement.

let  $A_2$  be the event that every pair of points  $(x_1, x_2)$  is contained in at least one  $S_i$  (i.e.  $S_1, \dots, S_M$  cover),

then

$$P(A_2) \geq 1 - \binom{n}{2} \left[ 1 - \left( \frac{k-1}{n-1} \right)^2 \right]^M.$$

In particular, given  $p \in (0, 1)$  then

setting

$$M \geq \left[ 2 \log \left( \frac{n}{2} \right) - \log(1-p) \right] \left( \frac{n-1}{k-1} \right)^2$$

ensures  $P(A_2) \geq p$ .