# New Results on the Stability of Persistence Diagrams

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joint work with

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# Stability

Setting: K - finite simplicial complex

$$f,g:K\to\mathbb{R}$$

 $\mathsf{Dgm}(f)$  - sub-levelset PD  $(f^{-1}(-\infty, \alpha])$ 

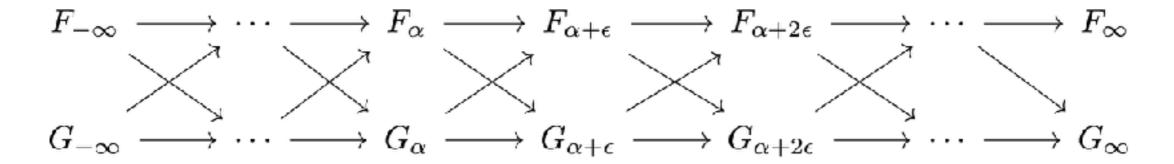
#### **Stability Theorem [CSEH06]**

Let X be a triangulable space with continuous tame functions  $f,g:X\to\mathbb{R}$ . Then the persistence diagrams  $\mathrm{Dgm}(f)$  and  $\mathrm{Dgm}(g)$  for their sublevel set filtrations satisfy

$$d_B(\mathsf{Dgm}(f), \mathsf{Dgm}(g))) \le ||f - g||_{\infty}.$$

## Extensions

Interleaving filtrations



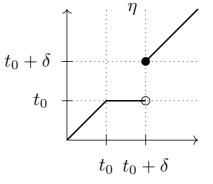
Categorical formulation

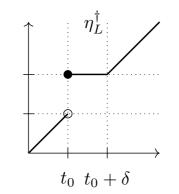
$$(\mathbb{R}, \leq) \xrightarrow{T_{\varepsilon}} (\mathbb{R}, \leq) \xrightarrow{T_{\varepsilon}} (\mathbb{R}, \leq)$$

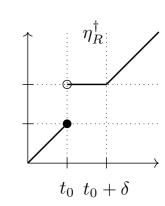
$$\downarrow F \qquad \Rightarrow \qquad \downarrow F \qquad \Rightarrow \qquad \downarrow F$$

$$D = D = D$$

Non-uniform interleaving t<sub>0</sub> + δ







# Applications of Stability

- Homological reconstruction
- Statistical inference

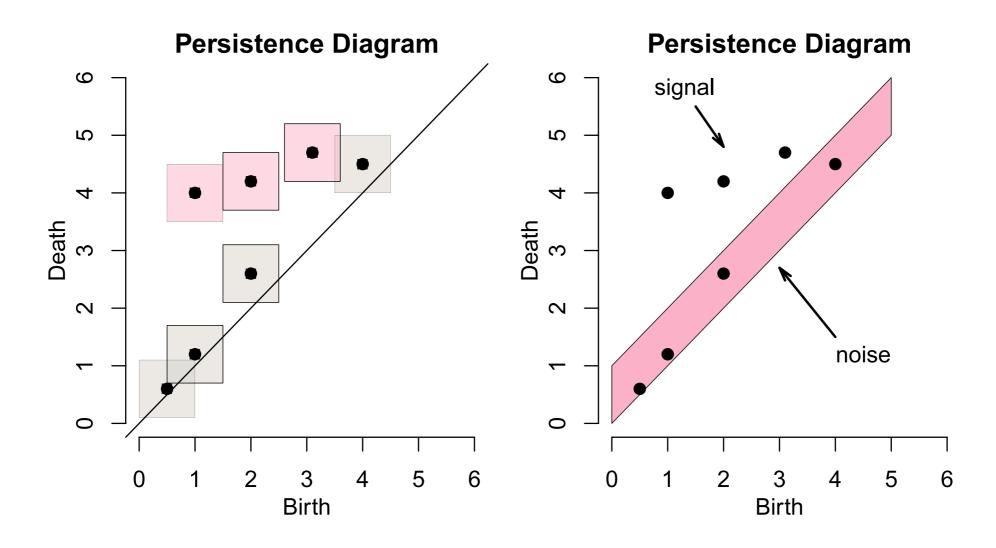
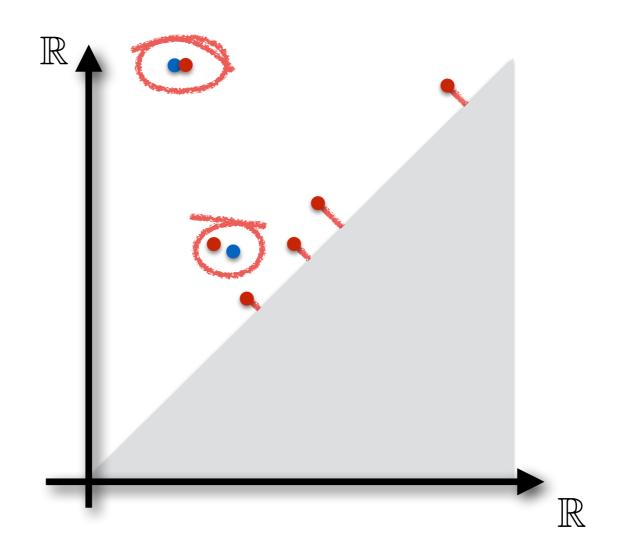


Image from Fasy, Brittany Terese, et al. "Confidence sets for persistence diagrams." The Annals of Statistics 42.6 (2014): 2301-2339.

## **Bottleneck Distance**

#### **Definition**

$$d_{\infty}(X,Y) = \inf_{\substack{\phi: X \to Y \\ \phi \in \text{bijections}}} \sup_{x \in X} \|x - \phi(x)\|_{\infty}.$$



## Problem: Outliers

- Bottleneck distance is defined by worst case
- To prove convergence, requires no outliers w.h.p.

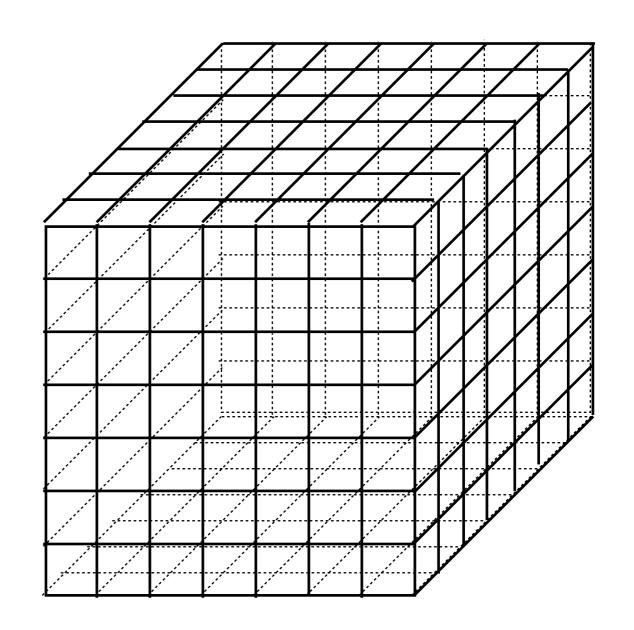
#### **Example**

Let  $\mathcal P$  be drawn by sampling M in i.i.d. fashion according to the uniform probability measure on M. Then with probability greater than  $1 - \delta_{\varepsilon}$  we have that  $\mathcal P$  is  $\frac{\varepsilon}{2}$ -dense  $(\varepsilon < \frac{\tau}{2})$  in M provided

$$|\mathcal{P}| > \beta_1(\log(\beta_2) + \log(\frac{1}{\delta}))$$

## Problem: Outliers

- Sub/super level set persistence
- Most errors are small, some are large
- Bottleneck distance is large



# p-Wasserstein Distance

#### **Definition**

The p-Wasserstein distance between two PDs X and Y is defined as

$$d_p(X,Y) = \left(\inf_{\substack{\phi: X \to Y \\ \phi \in \text{bijections}}} \sum_{x \in X} \|x - \phi(x)\|_p^p\right)^{1/p}$$

Often used in applications

# p-Wasserstein Distance

#### **Definition**

The p-Wasserstein distance between functions f and d defined on a simplicial complex K is defined as

$$||f - g||_p^p = \sum_{\Delta \in K} |f(\Delta) - g(\Delta)|^p.$$

Simplicial norm (compare to standard definition)

## Related Work

Wasserstein stability for Lipschitz functions

#### **Theorem**

Let X be a triangulable space with continuous tame functions  $f, g: X \to \mathbb{R}$ . Then the persistence diagrams Dgm(f) and Dgm(g) for their sublevel set filtrations satisfy

$$W_p(\mathsf{Dgm}(f), \mathsf{Dgm}(g))) \le C||f-g||_{\infty}^{1-\frac{\kappa}{p}}.$$

C depends on total persistence

Uses equivalence of norms (number of points)

D. Cohen-Steiner, H. Edelsbrunner, J. Harer and Y. Mileyko, *Lipschitz Functions Have Lpstable Persistence*, FOCM, April 2010

## Goal

#### **Theorem**

Let X be a triangulable space with continuous tame functions  $f, g: X \to \mathbb{R}$ . Then the persistence diagrams Dgm(f) and Dgm(g) for their sublevel set filtrations satisfy

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## Goal

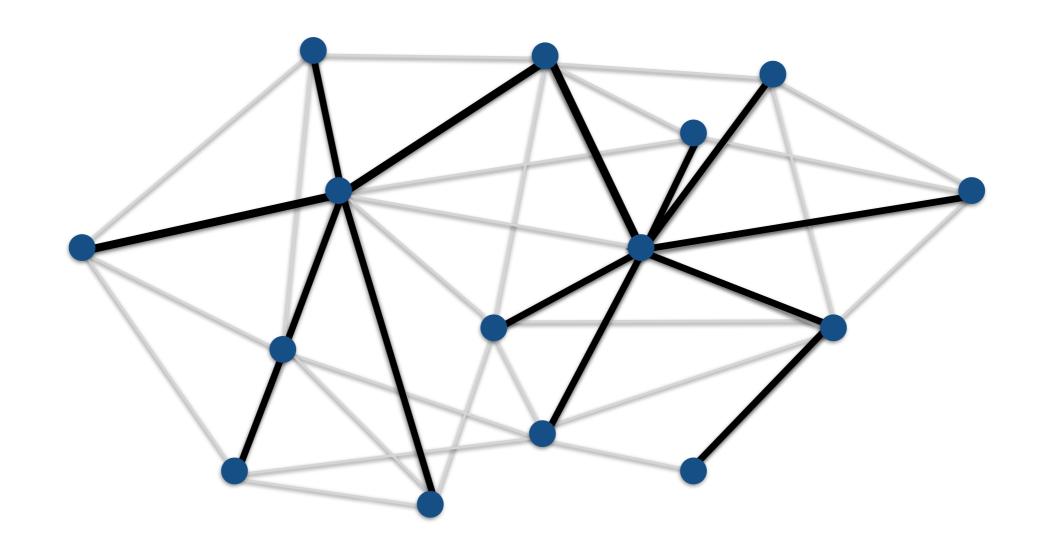
#### Theorem

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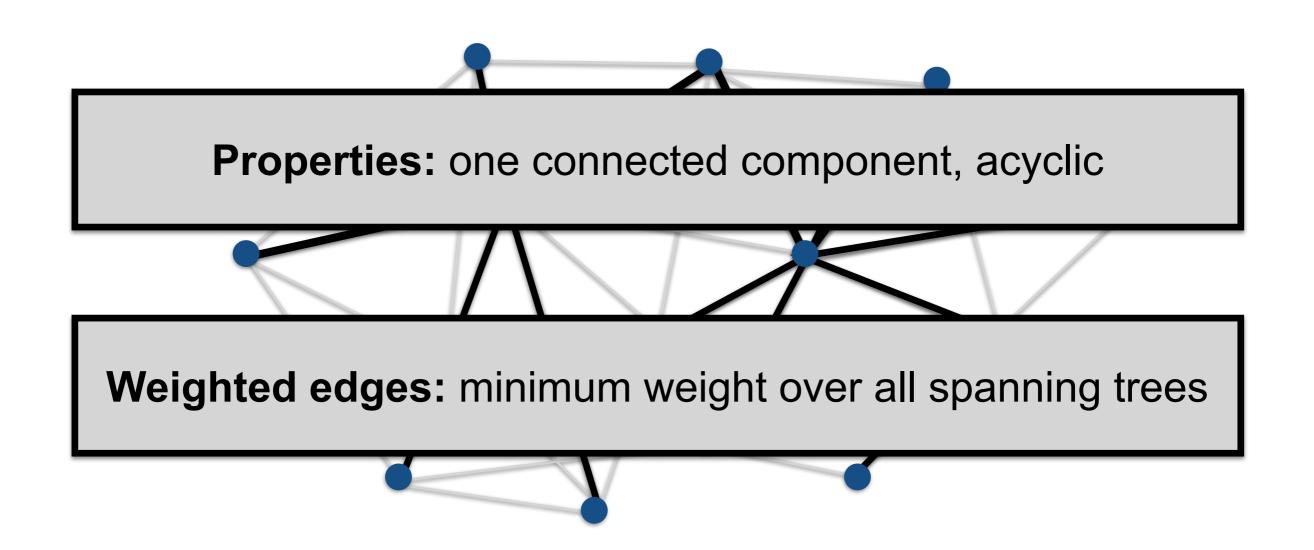
$$W_p(\mathsf{Dgm}(f), \mathsf{Dgm}(g))) \le C||f - g||_p$$

Why is this reasonable?

Generalization of a minimum spanning tree



Generalization of a minimum spanning tree



Fix dimension d - consider d-skeleton of a simplicial complex

#### **Algebraic properties:**

Acyclicity:  $\beta_d = 0$ 

Spanning:  $\beta_{d-1} = 0$ 

Fix dimension d - consider d-skeleton of a simplicial complex

#### **Algebraic properties:**

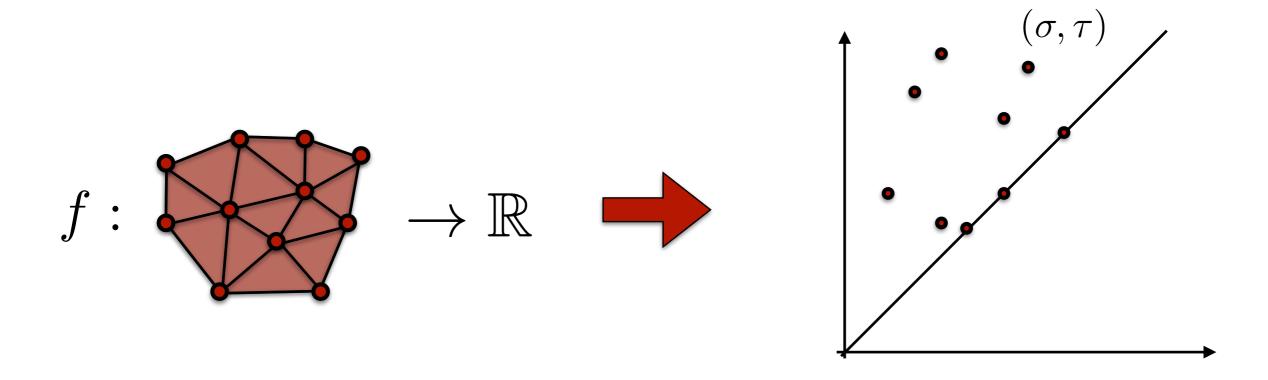
Acyclicity:  $\beta_d = 0$ 

Spanning:  $\beta_{d-1} = 0$ 

If (d-1)-Betti number cannot be zero, higher dimensional generalisation of spanning forest.

**Weighted version:** *d*-simplices have weights

Use weight function as a filtration



#### **Observation**

Negative simplicies (& death times) correspond to simplicies in MSA

## Birth and Death Times

#### **Theorem**

Let K be a finite complex with two monotone functions f, g. For any  $p \in \{0, \dots, \infty\}$ ,

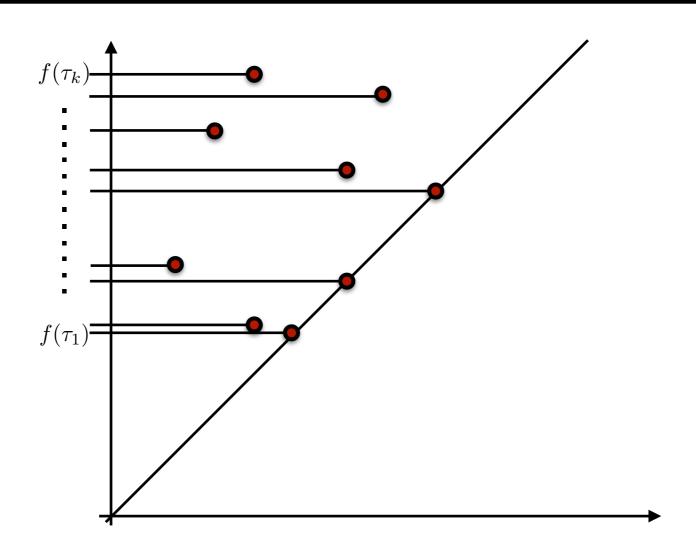
$$\inf_{\pi \in \Pi_D} \sum_i |D_i - \pi(D_i')|^p \le \sum_{\sigma \in K} |f(\sigma) - g(\sigma)|^p$$

$$\inf_{\pi \in \Pi_B} \sum_i |B_i - \pi(B_i')|^p \le \sum_{\sigma \in K} |f(\sigma) - g(\sigma)|^p$$

## Outline of Proof

Death times represent when no new classes are created (including instantaneous classes)

Each simplex + or - ⇒ consequence of Mayer-Vietoris



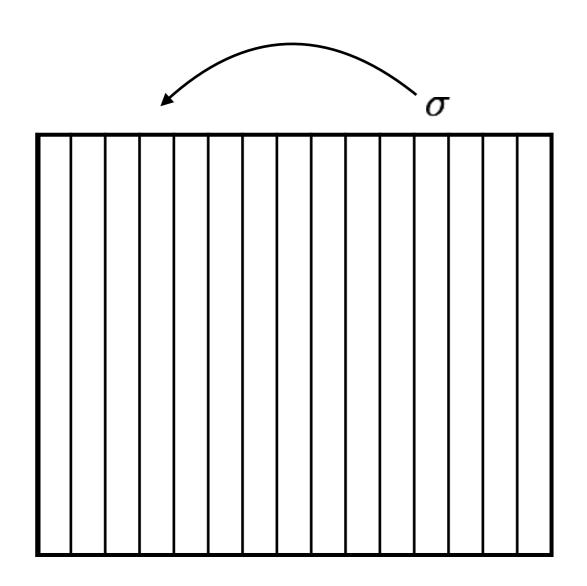
## Outline of Proof

#### 4 cases:

- 1. Moving a positive simplex forward
- 2. Moving a negative simplex backward
- 3. Moving a positive simplex backward
- 4. Moving a negative simplex forward

## Outline of Proof

Moving a positive simplex backward



If remains positive - trivial

 If negative - birth time moves less than how much the simplex moved

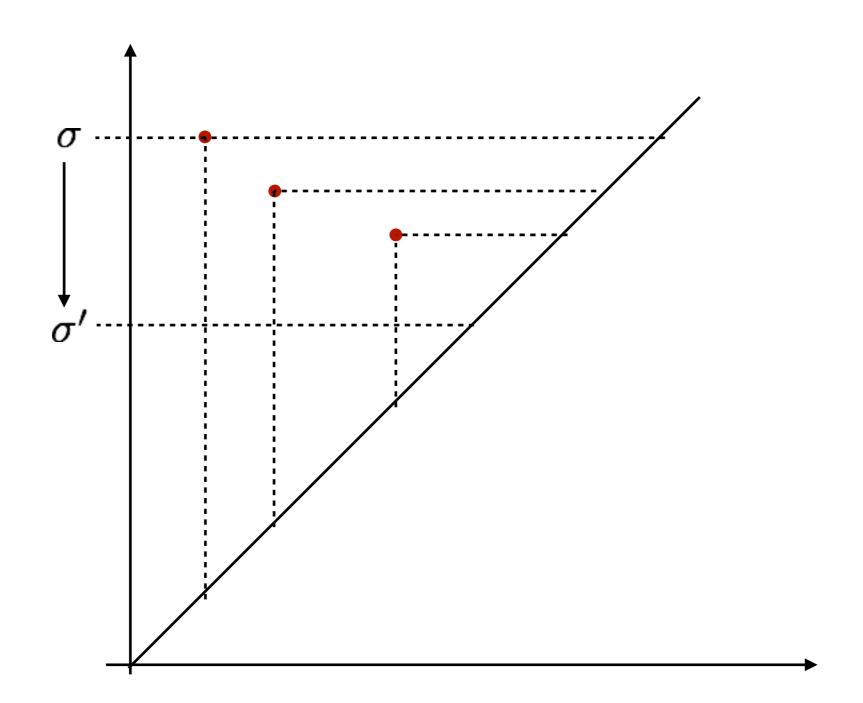
# Towards Persistence Diagrams

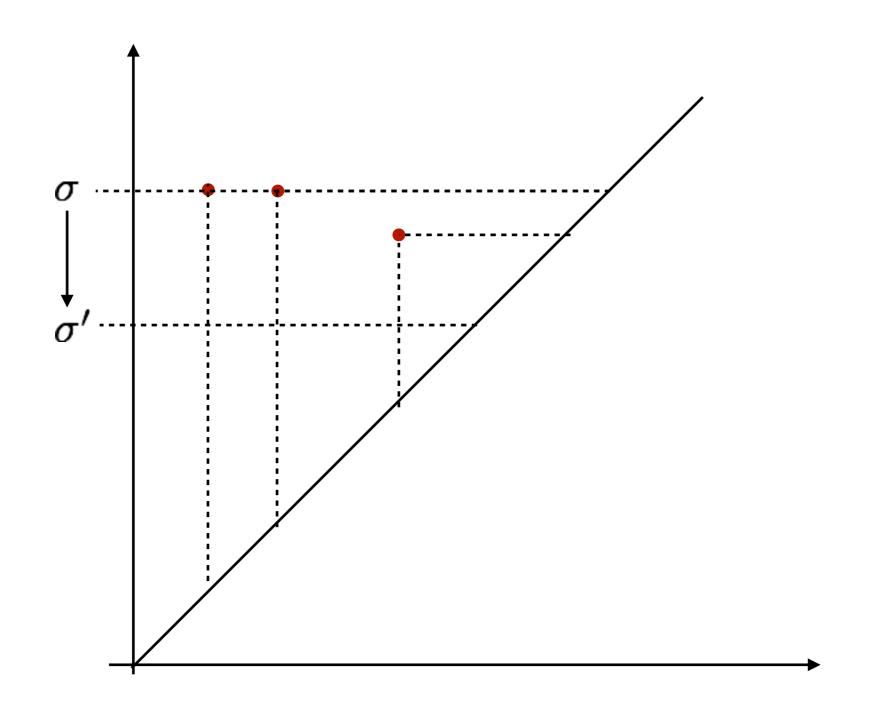
In persistence, need to take care of pairing between positive and negative simplicies

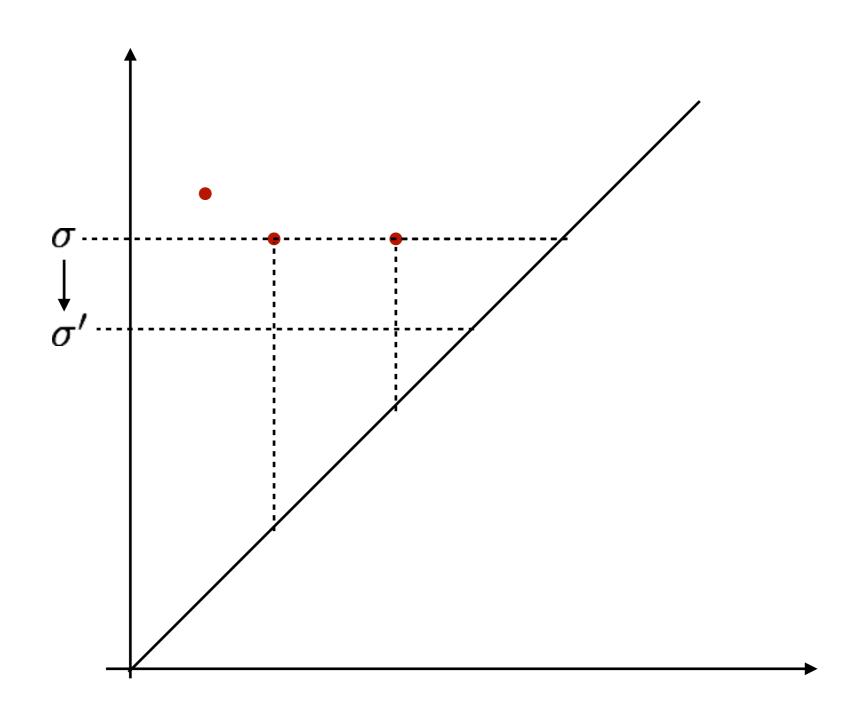
#### Lemma

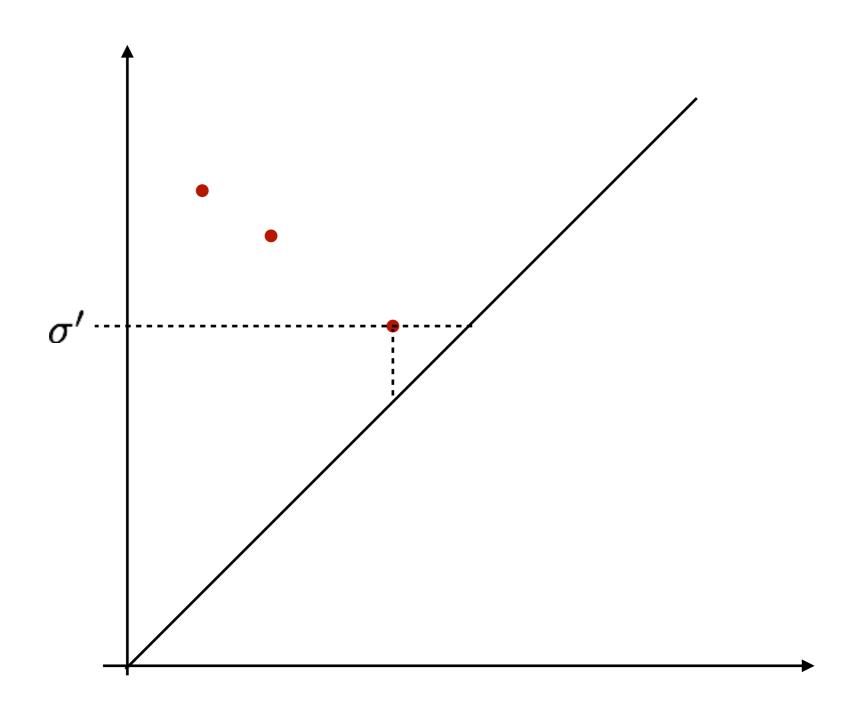
Let  $f:K\to\mathbb{R}$  be a monotone function over a simplicial complex K. There exists a map from  $\mathrm{Dgm}(f)$  to pairs of simplices  $(\sigma,\tau)$  such that for  $p\in\mathrm{Dgm}(f)$ , if  $\pi(p)=(\sigma,\tau)$ , then  $f(\sigma)=p_x$  and  $f(\tau)=p_y$ .

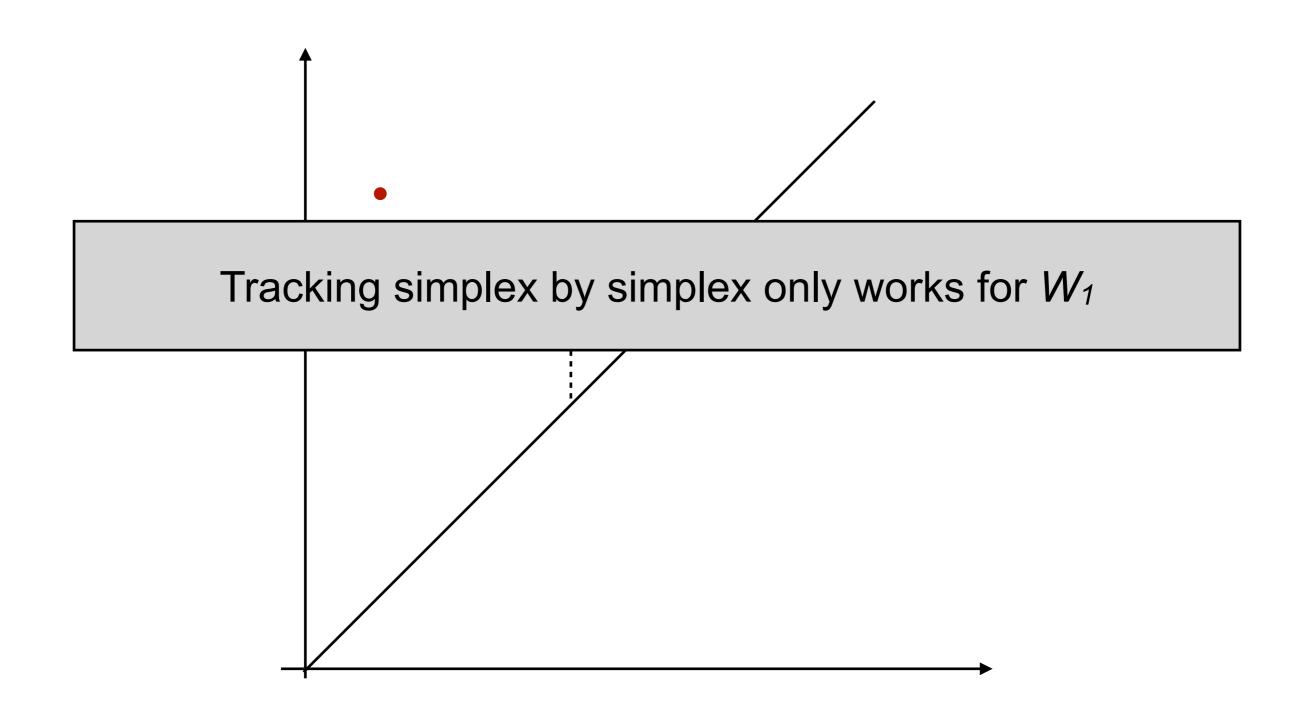
How can this map change?











## **Critical Pairs**

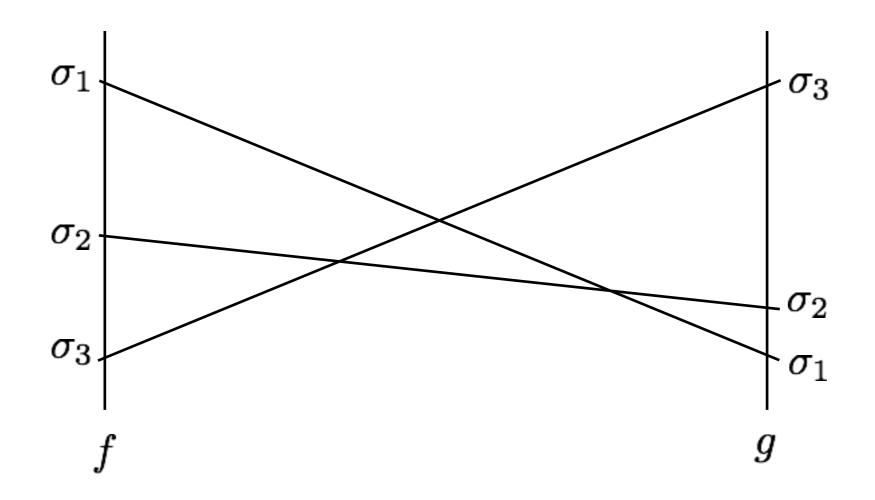
#### **Definition**

A *critical pair* is a pair of simplicies in the image of  $\pi$  such that  $f(\tau) - f(\sigma) > 0$ . We call any simplex critical if it part of a critical pair.

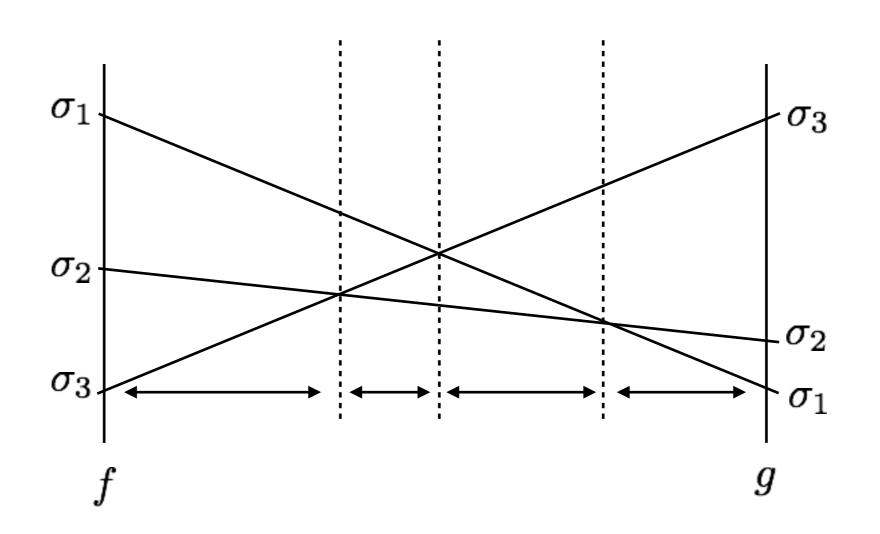
Correspondence with points off the diagonal

# Interpolation

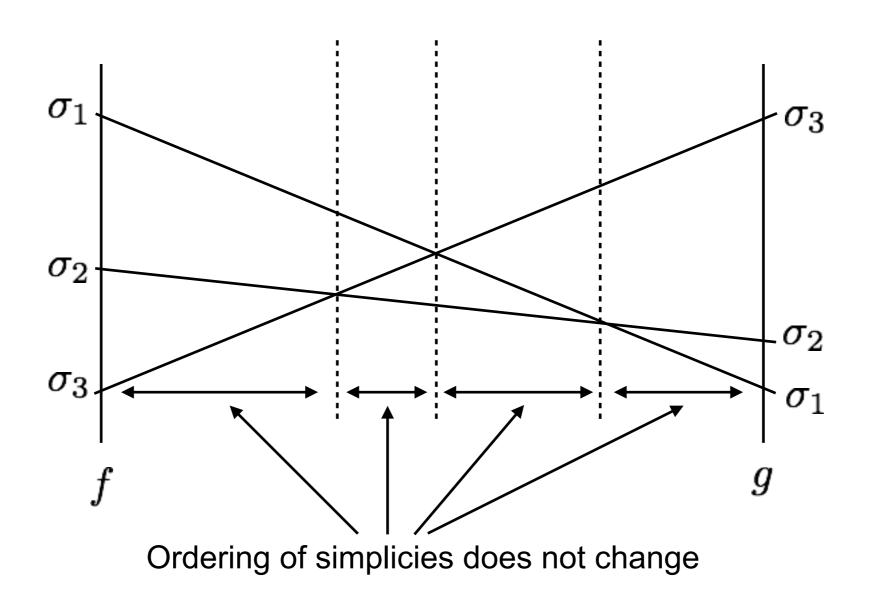
Linearly interpolate between functions



# Dividing Up the Problem



# Dividing Up the Problem



# Easy Case

#### Lemma

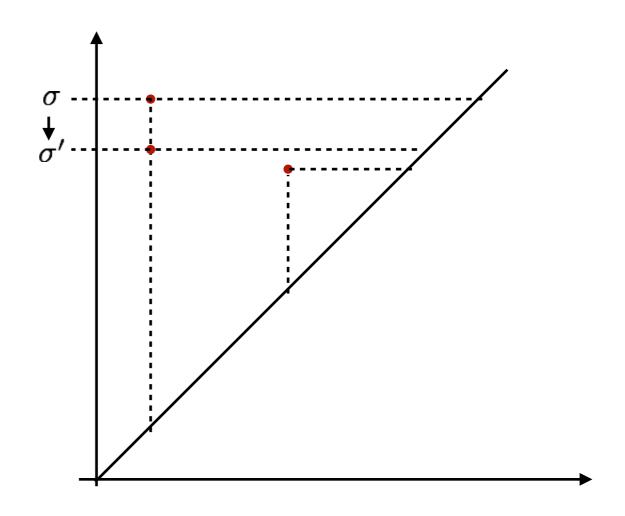
Let  $f_t: K \to \mathbb{R}$ ,  $t \in [a,b]$  be a continuous family of monotone functions over a simplicial complex K such that for all a < s < b the order (potentially with equality) of the function values of the simplices remains the same. Then

$$W_p(\mathsf{Dgm}(f_a), \mathsf{Dgm}(f_b)) \leq ||f_a - f_b||_p.$$

## Proof Idea

For  $t \in (a, b)$ , the ordering does not change so pairing map does not change (we extend to a and b by taking one-sided limits)

Since pairing map does not change, movement of points is equal to movement of simplices



# Combining Intervals

Within each interval we have the desired bound

For each interval (a, b),  $0 \le a \le b \le 1$  we have

$$||f_a - f_b||_p = |b - a| \cdot ||f - g||_p$$

$$\begin{split} W_p(\mathsf{Dgm}(f), \mathsf{Dgm}(g)) & \leq \sum_{i=0}^n W_p(\mathsf{Dgm}(f_{a_i}), \mathsf{Dgm}(f_{a_{i+1}})) \\ & \leq \sum_{i=0}^n \|f_{a_i} - f_{a_{i+1}}\|_p \\ & = \sum_{i=0}^n (a_{i+1} - a_i) \|f - g\|_p \\ & = \|f - g\|_p \end{split}$$

## Main Result

#### **Theorem**

Let  $f, g: K \to \mathbb{R}$  be monotone functions,

$$W_p(\mathsf{Dgm}(f), \mathsf{Dgm}(g)) \le ||f - g||_p.$$

## Observations

- Outliers do not affect Wasserstein distances between persistence diagrams too much
- Requires fixed simplicial complex (since we use a simplicial norm)
- It does not use the equivalence of norms

# Vietoris-Rips Filtrations

- Often we build complexes from point sets
- Can a similar result hold if we move/perturb points?

## Result

#### Theorem

Fix k,d If  $C_{d,k}$  is finite then for all  $p\geq 1$ , assuming  $X_0,X_1\subset \mathbb{R}^d$ 

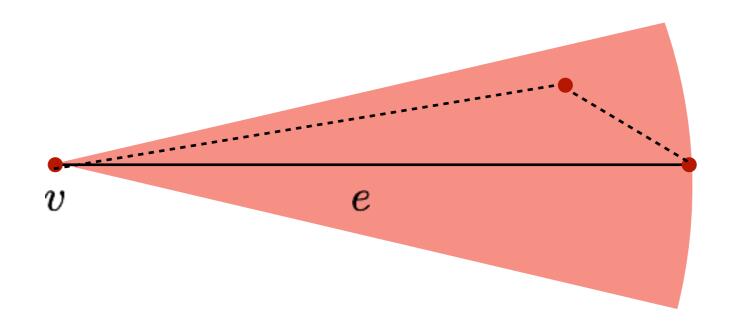
$$W_p(\mathsf{Dgm}_k(X_0), \mathsf{Dgm}_k(X_1) \le C_{d,k}^{1/p} D_p(S_0, S_1).$$

where  $\operatorname{Dgm}_k(X_i)$  is the k-dimensional persistence diagram for the Vietoris-Rips filtration on the point set  $X_i$ .

When can we bound  $C_{k,d}$ ?

# Components(H<sub>0</sub>)

- Cannot have too many classes around one vertex
- We do not points, so only consider critical edges (which for 0-dim only kill homology classes)
- Assuming e is critical (adjacent to a critical simplex), we can exclude a cone



# Čech Filtrations

#### **Distance filtration:**

Let  $\mathcal{P} \subset \mathbb{R}^d$  be a finite point set, let  $f: \mathbb{R}^d \to \mathbb{R}$ , defined by

$$f(x) = \min_{p \in \mathcal{P}} d(x, p)$$

Homotopic to the Čech Filtration

# Counterexample

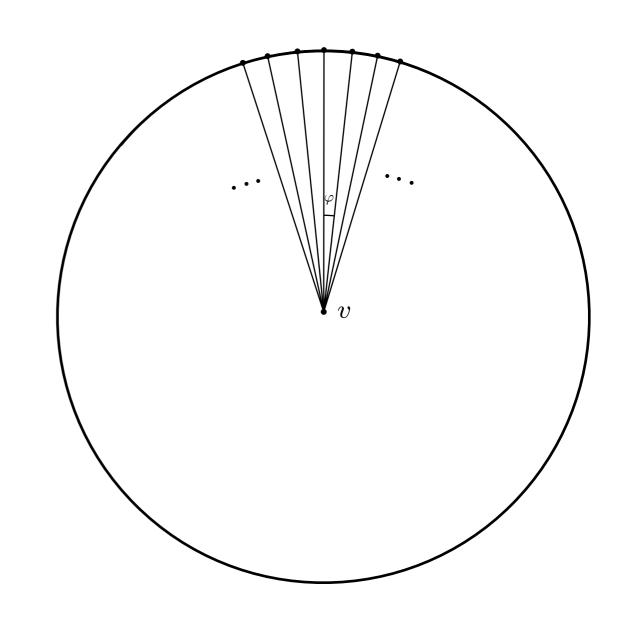
Points placed along a circle

Configuration is generic

Constant depends on n,

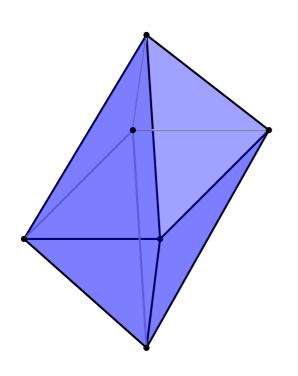
$$C_{2,2} = O(n)$$

$$C_{2,2} = O(n)$$
$$C_{2,1} = O(n)$$

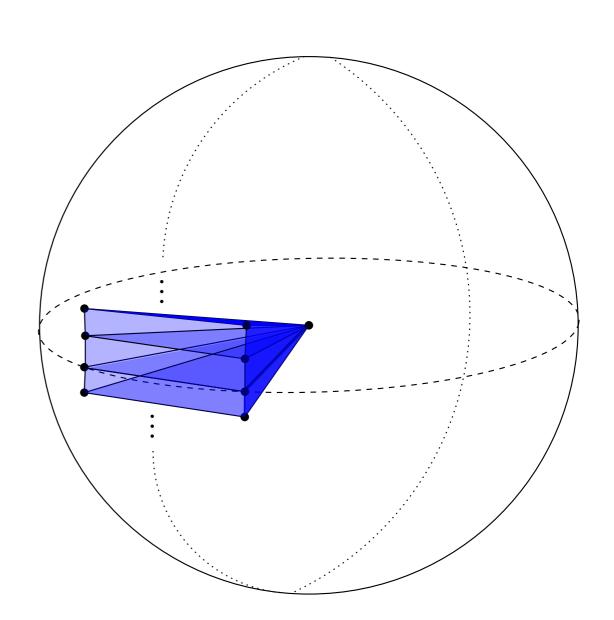


# Bounding C<sub>d,k</sub>

Counterexample



minimal nontrivial 2-cycle



packing critical simplifies

# Bounding $C_{2,k}$

Generalize 0-dimensional case: cannot be adjacent to too many critical simplifies

#### Lemma

If  $\tilde{H}_k(\mathsf{Lnk}(e)) = 0$  for all k, then e cannot be a critical edge.

**Proof:** Mayer-Vietoris

# Bounding C<sub>2,1</sub>

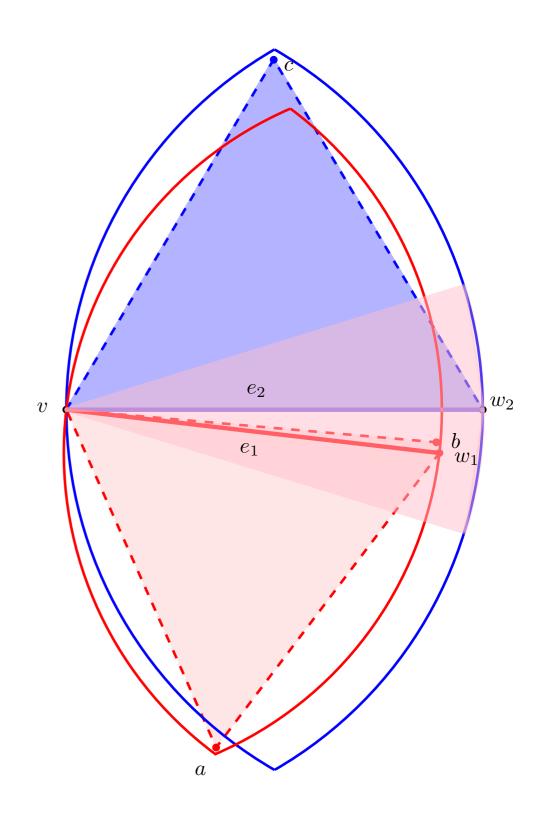
#### **Corollary**

If e an edge is critical, either  $\mathsf{Lnk}(e)$  consists only of the endpoints of e or  $\mathsf{Lnk}(e)$  (and so  $\mathsf{Cl}\ \mathsf{St}(e)$ ) must contain at least two vertices,  $v_1$  and  $v_2$  such that  $v_1, v_2 \not\in e$  and  $d(v_1, v_2) > f(e)$ . In  $\mathbb{R}^2$ , this implies that  $v_1$ , lies in the half-plane above e and  $v_2$  in the half-plane below e.

**Implication:** There are at most three 1-critical edges within an angle of  $\pi/12$  of each other.

If there are four 1-critical edges, then any triangle which includes the 4th (longest) edge, is homologous to an existing triangle.

Should extend to  $C_{2,k}$ 



## What's Next?

- Relating simplicial norm to more classical norm, e.g. Sobelov norm, recent work by Polterovich et. al. (Persistence barcodes and Laplace eigenfunctions)
- Expected Wasserstein bounds (bad cases are generic but unlikely)
- Combining with approximate Nerve Theorem