

Reminder: Please submit short project proposal on Canvas.

If you would like a partner, and haven't found one, just say so in your proposal, and I will try to match people based on interests.

Today: When can we recover homology from samples?

Recall: Hausdorff distance btw sets

$$d_H(X, Y) = \max \left\{ \sup_{x \in X} \inf_{y \in Y} \underline{d}(x, y), \sup_{y \in Y} \inf_{x \in X} d(y, x) \right\}$$

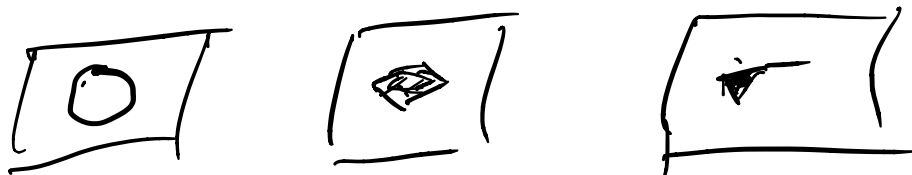
Last time:

$$d_H(\text{Proj}_K(R(x)), \text{Proj}_K(R(y))) \leq 2d_H(x, y)$$

this doesn't say anything abt "ground truth"

Order Ch. 4.

Let K be some compact set in Euclidean space \mathbb{R}^d . Examples: sub-manifolds of \mathbb{R}^d , convex sets, etc



K is "ground truth"

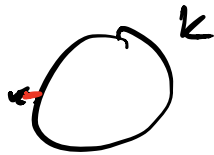
We sample points $X \sim K$, or $X \sim K + \epsilon$
|
noise



K has some ground truth $H_K(K)$
 can we recover $H_K(K)$ from samples X ?

we'll look at offsets:

1) we define distance to K : $d_K(x) = \min_{y \in K} \|x - y\|$



$$d_K: \mathbb{R}^d \rightarrow \mathbb{R}_+$$

r -offset of K : $K_r = d_K^{-1}((-\infty, r])$

$$= \emptyset \quad \text{if } r < 0$$

$$= K \quad \text{if } r = 0$$

$$= \mathbb{R}^d \quad \text{if } r = \infty$$

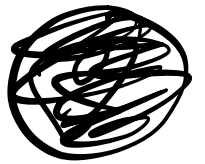
$K_r \subseteq K_s \quad \forall r \leq s \rightarrow$ filtration.



$\rightarrow K_{0.1}$



K_5 :



Note: i -offset in Cudot .

$$K_r \neq \mathcal{L}(K; r)$$

\rightarrow simplicial complex

\hookrightarrow continuous

Reach:

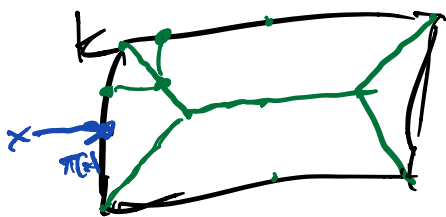
Projection set of $x \in \mathbb{R}^d$

$$\Pi_K(x) = \{ y \in K \mid d(x, y) = d_K(x) \}$$

for many $x \in \mathbb{R}^d$, this is a single pt.

on medial axis $|\Pi_K(x)| \geq 2$.

$$M(K) = \{ x \in \mathbb{R}^d \mid |\Pi_K(x)| \geq 2 \}$$



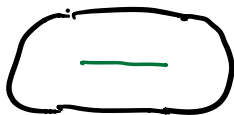
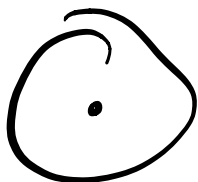
$$\text{rch}(K) = 0$$

Reach of K : min. distance btw. K and $M(K)$

at a point $\text{rch}(K, x) = d_{M(K)}(x)$

$$\text{rch}(K) = \min_{x \in K} \text{rch}(K, x)$$

$$= \min_{x \in K} d_{M(K)}(x)$$

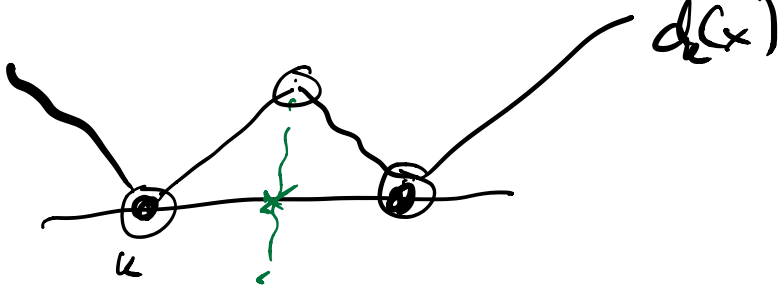


when K has $\text{rch}(K) > 0$, there is a nbhd of K which retracts onto K : if $r < \text{rch}(K)$

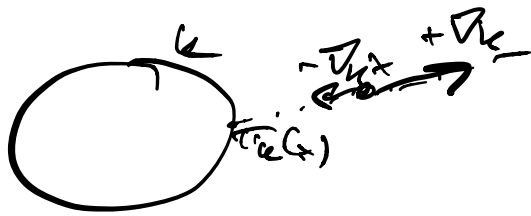
$$x_r \rightarrow K \quad \text{via} \quad x \mapsto \Pi_K(x)$$

flow along $\nabla d_K(x)$

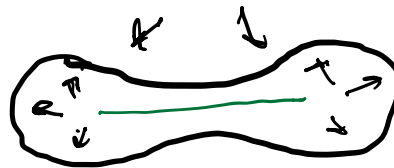
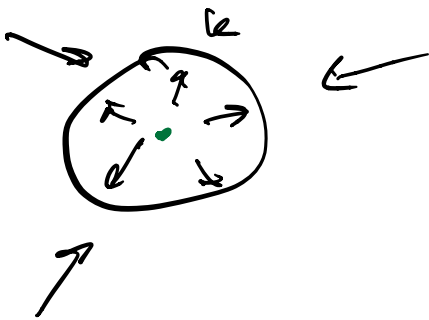
idea: d_K is 1-Lipschitz over \mathbb{R}^d
 differentiable over $\mathbb{R}^d \setminus (K \cup \partial K)$



$$D_K(x) = \frac{x - \pi_K(x)}{\|x - \pi_K(x)\|} = \frac{x - \pi_K(x)}{d_K(x)}$$



this is cts, so $-D_K$ can be integrated into a flow.



if $\text{rch}(K) > 0$, K_T retracts onto K
 using $-D_K$

Thm: Niyogi, Smale, Weinberger 05/06, 08.

Let K be cpt. submanifold of \mathbb{R}^d , with $\text{rch}(K) > 0$. Let P be a pointcloud $P \subseteq \mathbb{R}^d$

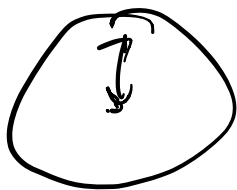
s.t. $d_H(K, P) = \epsilon < \sqrt{\frac{3}{20}} \text{rch}(K)$. Then

for any $r \in (\underline{2\epsilon}, \sqrt{\frac{3}{5}} \text{rch}(K))$

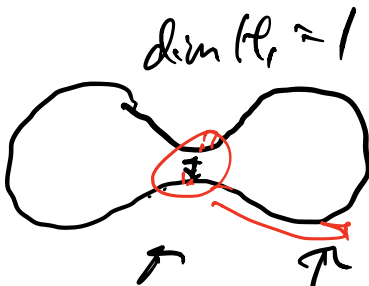
P_r deformation retracts onto K .

$$K \simeq P_r$$

$$H_k(K) \hat{=} H_k(P_r)$$



easy to reconstruct H_k



$\dim H_1 = 1$

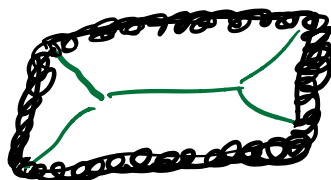


$\dim H_1 = 2$



hard

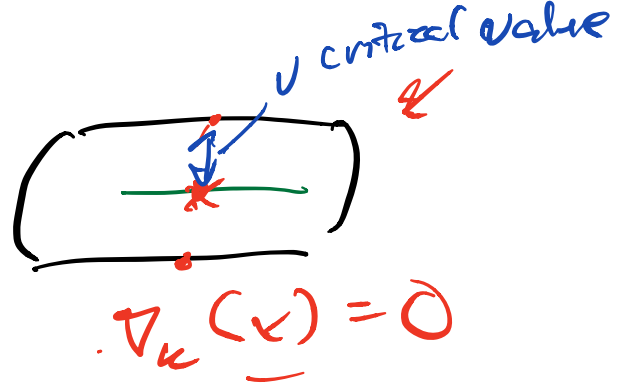
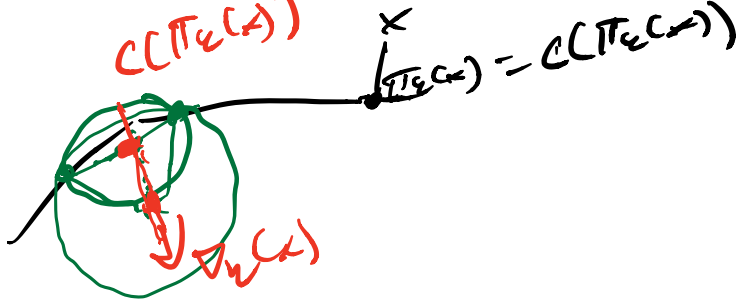
result above is useless



weak feature size. Chazal, Cohen-Steiner.
Lichtner - 2009

can extend D_K to medial axis?

$$D_K(x) = \frac{x - c(\Pi_K(x))}{d_K(x)} \quad c(\Pi_K(x)) \text{ center of } \Pi_K(x)$$



def: A point $x \in \mathbb{R}^d$ is critical if - its generalized gradient $\overline{\nabla_k c(x)} = 0$, or equivalently if $x \in \text{cvx hull of } \{x \mid \nabla_k c(x) = 0\}$

A value is critical if $v = d_k c(x)$ for some critical point x

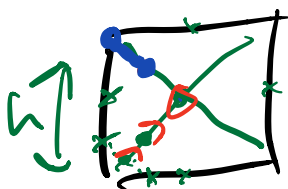
Lemma: if $0 \leq r \leq s$ are s.t. there are no crit. values of d_k in $[r, s]$, then k_s def. rel. onto k_r $k_r \cong k_s$

Weak feature size:

$$wfs(k) = \inf \{ r > 0 \mid r \text{ is a crit. value of } d_k \}$$

Comment: critical points all live on $\partial M(k)$

$$\Rightarrow \text{rch}(k) \leq wfs(k)$$



$$\text{rch}(k) = 0 \quad (\text{one crit. pt. at center})$$

$$wfs(k) = \frac{h}{2}$$

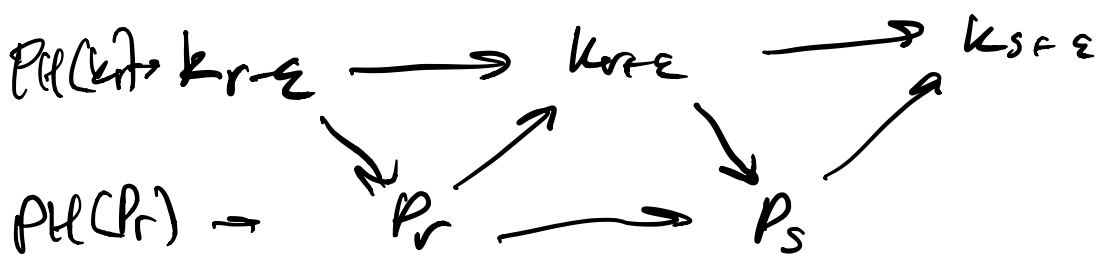
Remark: wfs is really what governs topy equivalence

Thm (C, C-S, L): Let K be a compact set in \mathbb{R}^d , w/ $wfs > 0$, let $P \in \mathbb{R}^d$,
 $d_{\text{top}}(K, P) = \varepsilon < \frac{1}{4} wfs(K)$

Then for any r, s s.t. $\underline{\varepsilon} < r < r + 2\varepsilon \leq s < \underline{\varepsilon} + wfs(K)$

the map $H_*(P_r) \rightarrow H_*(P_s)$ induced by

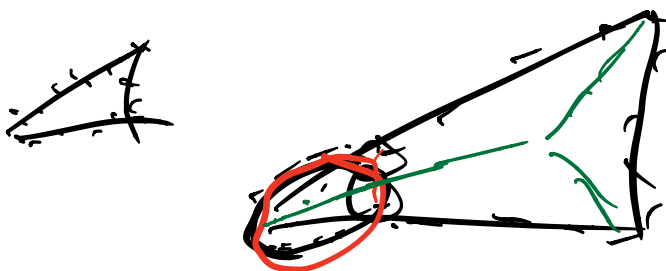
is isomorphic to $H_*(K_r) \xrightarrow{P_r \rightarrow P_s} H_*(K_s)$ + $\text{re}(0, wfs(K))$
 \downarrow
 K
 $H_*(K)$



any bars in barcode longer than 2ε in range $(0, wfs(K))$ are "real"

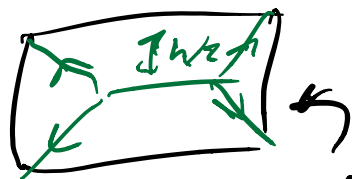
Topological SNR: $\frac{wfs(K) - 2\varepsilon}{2\varepsilon}$ ← signal bars
 ← noise bar

Note: P does not need to be $\leq K$ to have $d_{\text{top}}(P, K) \leq \varepsilon$.

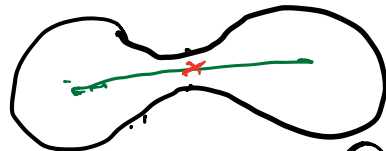


in presence of noise, no single value of r may have $P_r \cong K$, but we can still recover $H_k(K)$ using persistent homology.

wfs: what governs the ability to reconstruct.



$$rch(k) \neq wfs(k)$$



← still in trouble.

$$\partial rch(k) = wfs(k)$$

\mathbb{R}



need a lot of samples w/ no

noise

or noise ← $rch(k)$

P_r offset space,

$$\mathcal{C}(P; r)$$

P_r are continuous, unions of balls

$$P_r = \bigcup_{x \in P} B(x; r)$$

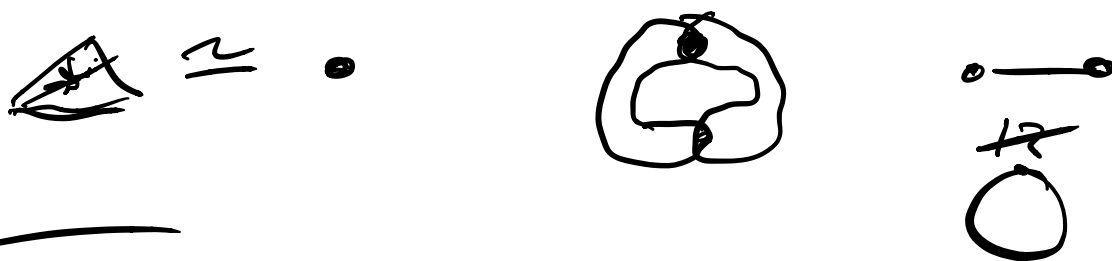
Each complex: Nerve of cover of P_r given by these balls.

$$\mathcal{C}(P; r) = \mathcal{N}(\{B(x; r)\})$$

Recall: simplex $(x_0 \dots x_k) \in \mathcal{L}(P; r)$ if $\bigcap B(x_i; r) \neq \emptyset$

Recall Nerve Thm: Let X be paracompact space, and \mathcal{U} an open cover of X , s.t. intersections of sets either empty or contractible. Then $N(\mathcal{U}) \simeq X$

$B(x; r)$ is convex: so if $\bigcap_{x \in A} B(x; r) \neq \emptyset$ then A is also convex. Convex sets are contractible:



Persistent Nerve thm:

Let $X \subseteq X'$, and $\mathcal{U}, \mathcal{U}'$ be open covers $\mathbb{T}_{\text{paracompact}}$

w/ same parameter set A . Assume that $U_a \subseteq U'_a \forall a \in A$, and $\mathcal{U}, \mathcal{U}'$ satisfy conditions of Nerve thm. Then Nerve equivalences commute w/ inclusions

$$H_k(x) \rightarrow H_k(x')$$

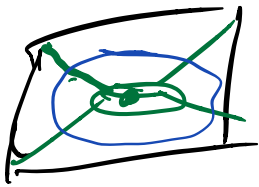
$$\cong \uparrow \quad \cup \quad \uparrow \cong$$

$$H_k(N(\mathcal{U})) \rightarrow H_k(N(\mathcal{U}'))$$

Corollary: Let K be compact set in \mathbb{R}^d
w/ $wfs(K) > 0$. Let $P \subseteq \mathbb{R}^d$, $d_H(P, K) = \epsilon$
 $\epsilon < \frac{1}{4} wfs(K)$, then there is a
"sweet range" $J: (\epsilon, wfs(K) - \epsilon)$, where
barcode of $\mathcal{Z}(P, r)$ has property:
bars w/ length $> 2\epsilon$ in sweet range
encode $H_k(K)$

Moral: Each complex filtrations have nice
theoretical properties.

μ -reach: a point is called μ -critical if
 $\|V_k(x)\| < \mu$. $M_\mu(K)$ the μ -medial axis.
 μ -reach: $r_\mu(K) = \min_{x \in K} d_{M_\mu(K)}(x)$



interpolates btw. $\frac{rch(\kappa)}{\mu}$ and $\frac{wfs(\kappa)}{\mu}$.

Thm: Let κ be compact w/ positive μ -reach for some $\mu \in (0, 1]$ let $d_H(P, \kappa) = \epsilon$
 $\epsilon < \frac{\mu^2}{5\mu^2 + 2} r_\mu(\kappa)$. Then for any

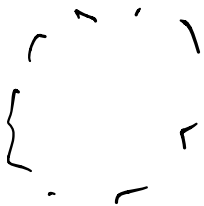
$$r \text{ s.t. } \frac{4\epsilon}{\mu^2} \leq r \leq r_\mu(\kappa) - 3\epsilon,$$

$$P_r \cong \kappa_S \text{ for any } S \in (0, wfs(\kappa))$$

$\Rightarrow \mu$ -reach tells us there is a nice range where there is no topological noise in barcode.

Difficulties:

- 1) if we don't know κ , we don't nec. know $rch(\kappa)$ or $wfs(\kappa)$ but might be able to bound



2) we need $d_H(P, K) < \epsilon$

we can accommodate bounded noise models,

but we don't have theory to deal with tails in noise.

$$X \sim K + N(0, \sigma^2)$$

not robust to outliers.

3) $\mathcal{C}(P, r)$ is hard to compute.

checking intersections of balls grows exp. in d , where \mathbb{R}^d is ambient space.

$\mathcal{C}(P, \infty)$ is very large.

$DC(P, r)$: restriction of \mathcal{C} to Delaunay

triangulation is http://

construction in high-dims difficult.