

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 b/w sets

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

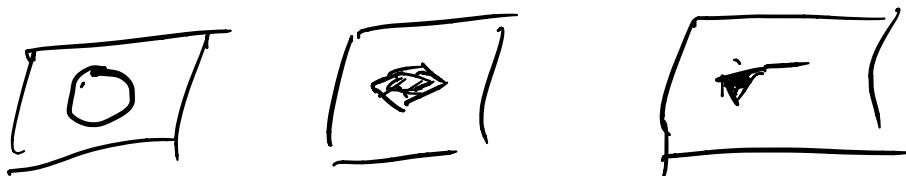
Last time:

$$d_H(PH_K(R(x)), PH_K(R(\gamma))) \leq 2d_H(x, \gamma)$$

this doesn't say anything abt "ground truth"

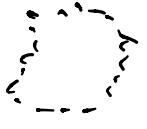
Out of Ch. 4.

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 $x \sim K + \frac{\epsilon}{n} \text{ noise}$



K

X

X

K has some ground truth $\text{f}_K(z)$
 Can we recover $\text{f}_K(x)$ from samples X ?

We'll look at offsets:

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

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

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

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

$K_r \subseteq K_S \wedge r \leq S \rightarrow$ filter, sm.

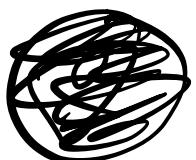


$$\rightarrow K_{0.1}$$



annulus

$$K_S$$



Note: i-offset in Octet.

$$K_r \neq \chi(K; r)$$

\hookrightarrow simplicial complex
 continuous

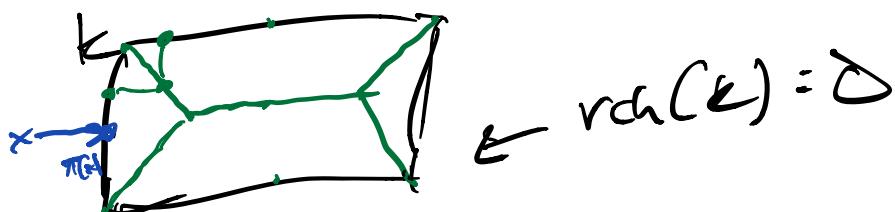
Reach:

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 several occasions $|\Pi_K(x)| \geq 2$.

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

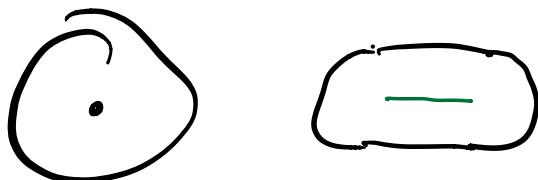


Reach of K : min. distance btw. K and $M(K)$

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

$$rch(K) = \max_{x \in \mathbb{R}^d} rch(K, x)$$

$$= \min_{x \in \mathbb{R}^d} d_{M(K)}(x)$$

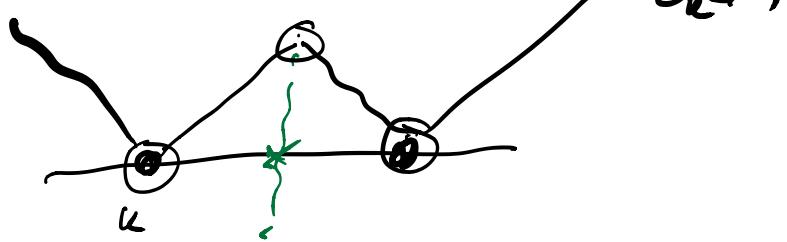


When K has $rch(K) > 0$, there is a nbhd of K which retracts onto K : if $r < rch(K)$

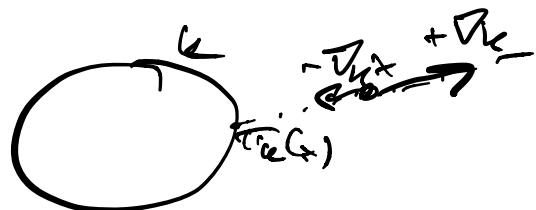
$$x_r \rightarrow K \quad \text{via} \quad x \mapsto \Pi_K(x)$$

flow along $\nabla d_K(x)$

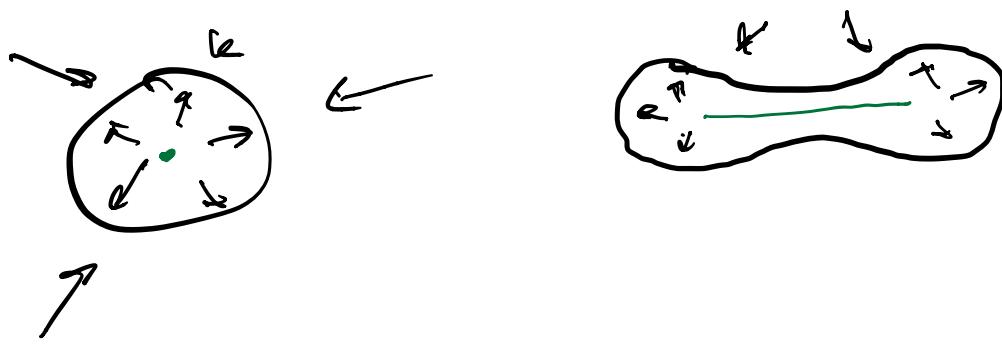
Idea: d_k is ℓ -Lipschitz over \mathbb{R}^d
 differentiable over $\mathbb{R}^d \setminus (k \cup M(k))$



$$D_k(x) : \frac{x - \pi_k(x)}{\|x - \pi_k(x)\|} = \frac{x - \pi_k(x)}{d_k(x)}$$



This is cfs, so $-D_k$ can be integrated
 into a flow.



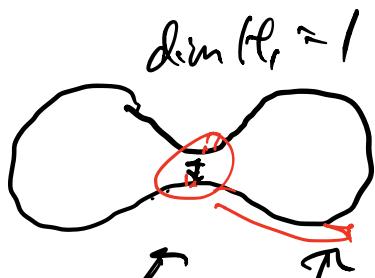
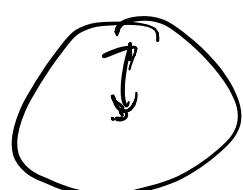
If $r_{ch}(k) > 0$, k_g refracts onto k
 using $-D_k$

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 $P \subseteq \mathbb{R}^{cl}$ s.t. $d_{\mathbb{R}^d}(K, P) = \varepsilon < \sqrt{\frac{3}{20}} \text{rch}(K)$. Then for any $r \in (\underline{2\varepsilon}, \overline{\sqrt{\frac{3}{5}} \text{rch}(\varepsilon)})$ P_r deformation retracts onto K .

$$K \cong P_r$$

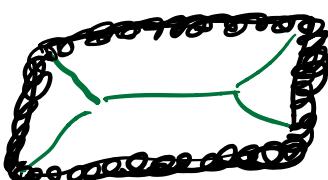
$$H_k(K) \cong H_k(P_r)$$



$$\dim H_r \approx 2 \approx \infty$$

t_{easy} to reconstruct the

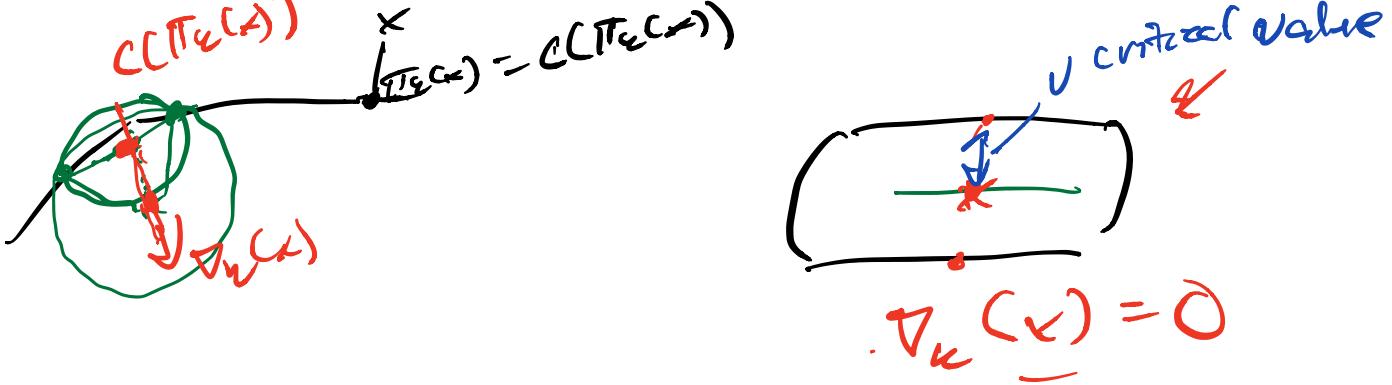
result above is useless



weak feature size. Chazal, Lohier-Sterner.
Lieutier - 2008

can extend D_K to medial axis.

$$D_K(x) = \frac{x - c(\pi_K(x))}{\text{dist}(x)} \quad c(\pi_K(x)) \cdot \text{center of } (\pi_K(x))$$



def: A point $x \in \mathbb{R}^d$ is critical, if its generalized gradient $\overline{\nabla}_k(x) = \emptyset$, or equivalently if $x \in \text{cvx hull of } t_k(x)$

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

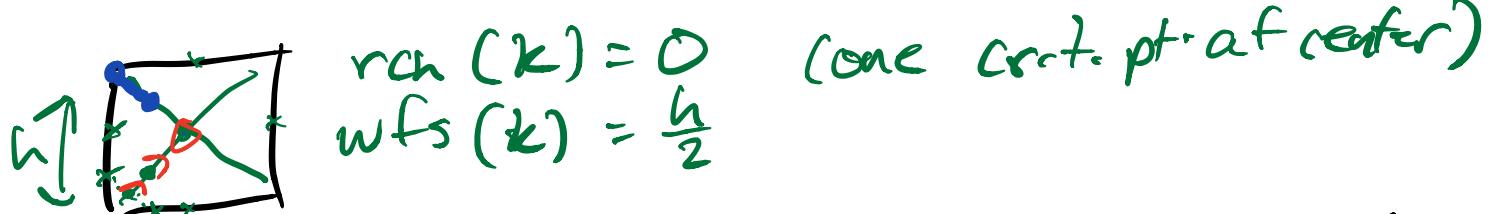
Lemma: if $0 \leq r \leq s$ are s.t. there are no crit. values of d_k in $[r, s]$, then r_k def. ref. onto k $r_k \approx k s$

weak feature size:

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

comment: critical points all live on $M(k)$

$$\Rightarrow rch(k) \leq wfs(k)$$



Final: wfs is really what governs hptg equivalence

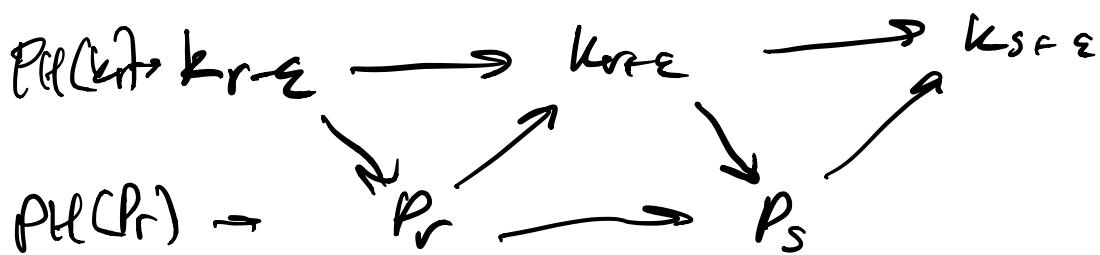
Thm $(C, C-S, L)$: Let K be a compact set in \mathbb{R}^d , w/ $wfs > 0$, let $P \subseteq \mathbb{R}^d$,

$$d_{CP}(K, P) = \varepsilon < \frac{1}{4} wfs(K)$$

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

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

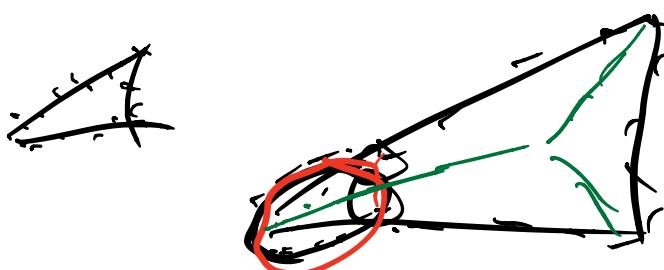
\cong isomorphic to $h_*(K_r) \xrightarrow{\begin{array}{c} P_r \rightarrow P_s \\ \cong \\ K \end{array}} h_*(K_s)$



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

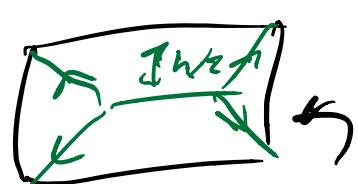
Topological SNR: $\frac{wfs(K) - 2\varepsilon}{2\varepsilon}$ ← signal bars
← noise bar

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



in presence of noise, no single value of r may have $\text{Pr} \simeq k$, but we can still recover $H_k(\text{Pr})$ using persistent homology.

wfs: what governs the ability to reconstruct.



$$\text{rch}(k) \neq \text{wfs}(k)$$



$$\sigma \text{rch}(k) = \text{wfs}(k)$$

∞ need a lot of samples w/o noise or noise $\ll \text{rch}(k)$

Pr offset space.

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

Pr are continuous, unions of balls

$$\text{Pr} = \bigcup_{x \in \mathcal{P}} B(x; r)$$

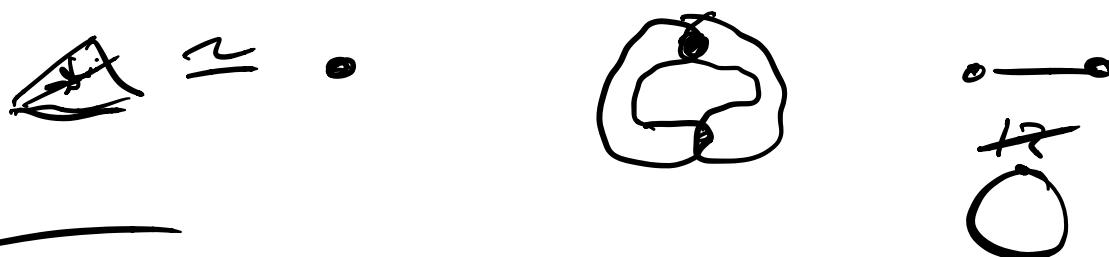
Cech complex: Nerve of cover of Pr given by these balls.

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

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

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 $N(\mathcal{U}) \cong K$

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



Persistent Nerve thm:

Let $X \subseteq X'$, and $\mathcal{U}, \mathcal{U}'$ be open covers
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
htpy equivalences commute w/ inclusions

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

$$\cong \quad \alpha \uparrow \cong$$

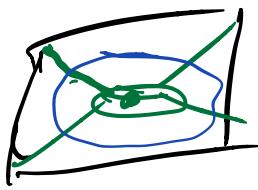
$$H_k(N(u)) \rightarrow H_k(N(u'))$$

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

Moral: Čech complex filtrations have nice theoretical properties.

μ -reach: a point is called μ -critical if $|N_{\mu}(x)| \leq \mu$. $N_{\mu}(K)$ the μ -medial axis.

μ -reach: $r_{\mu}(k) = \min_{x \in k} d_{N_{\mu}(x)}(x)$



interprets b/w $\frac{\text{rch}(x)}{\mu}$ and
 $\frac{\text{wfs}(k)}{\mu}$.

Then: Let k be compact w/ positive μ -reach
 for some $\mu \in [0, 1]$ (if $d_{\mathbb{R}^2}(P, k) = \epsilon$
 $\epsilon < \frac{\mu^2}{5\mu+2} r_{\mu}(k)$). Then for any

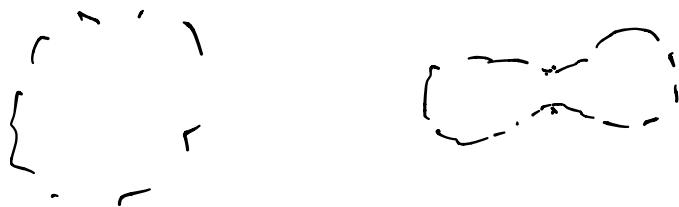
$$r \text{ s.t. } \frac{4\epsilon}{\mu^2} \leq r \leq r_{\mu}(k) - 3\epsilon,$$

$P_r \cong k_S$ for any $S \in C_0(\text{wfs}(k))$

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

Difficulties:

- (1) if we don't know k , we don't nec.
 know $\text{rch}(x)$ or $\text{wfs}(k)$
 but might be able to bound



- 2) we need $d_{\mathcal{H}}(P, K) < \varepsilon$
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) $\mathcal{E}(P, r)$ is hard to compute.

Checking intersections of balls grows
exp. in d , where R^d is ambient space.

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

$D(P, r)$: restriction of \mathcal{E} to delaunay
triangulation is highly over-
construction in high-dims difficult.