

Simplicial Manifold Reconstruction via Tangent Space Estimation

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Outline

- ▶ Building the triangulation
 - ▶ Weighted Delaunay complex
 - ▶ Restricted complex
 - ▶ A stability result for angles
- ▶ Statistical model and estimation procedure
 - ▶ Tangent space estimation *via* local PCA
 - ▶ Convergence rate
- ▶ A noisy model
 - ▶ Clustering *via* local PCA
 - ▶ Interpolation
 - ▶ Convergence rate

General Framework

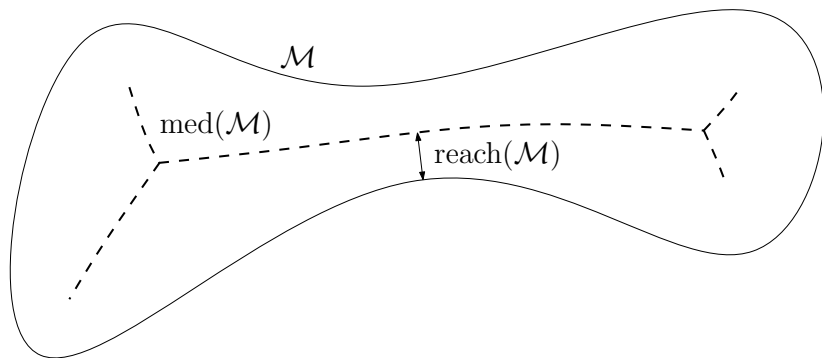
$\mathcal{M} \subset \mathbb{R}^D$ a d -dimensional sub-manifold.

The *reach* of \mathcal{M} is the minimal distance to its medial axis,

$$\text{reach}(X) = \inf_{x \in X} d(x, \text{med}(X)),$$

where

$$\text{med}(X) = \{p \in \mathbb{R}^D, \exists a \neq b \in \mathbb{R}^D, \|p - a\| = \|p - b\| = d(p, X)\}.$$



Reach Condition

Assume $\text{reach}(\mathcal{M}) \geq \rho$ for some fixed $\rho > 0$

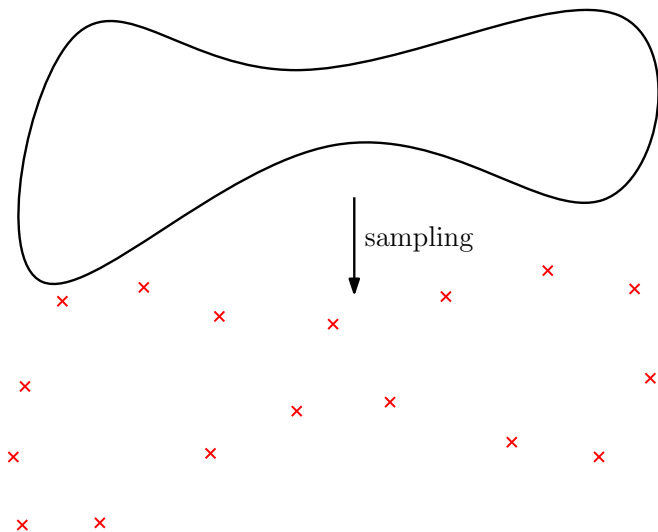


Figure: Reach and sampling

Reach Condition

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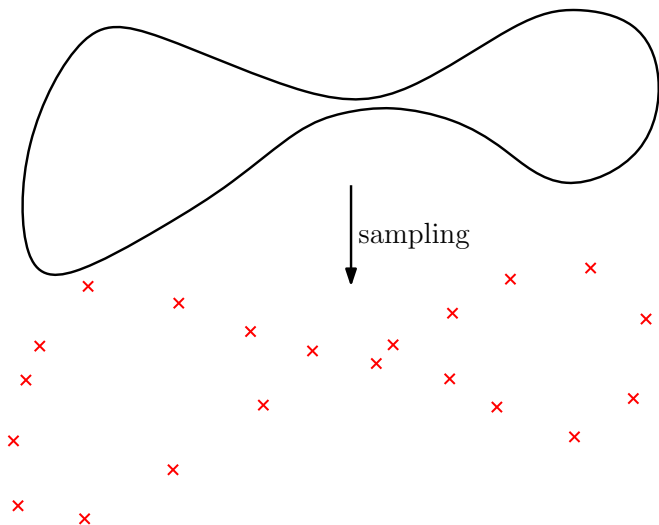


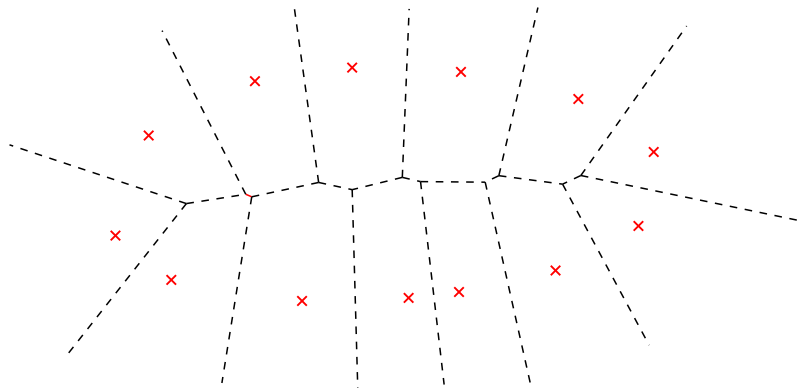
Figure: Reach and sampling

Weighted Delaunay Complex

For $\mathcal{P} = \{x_1, \dots, x_N\} \subset \mathbb{R}^D$ and a weight $\omega : \mathcal{P} \rightarrow [0, \infty)$, consider the weighted Voronoi $\text{Vor}^\omega(\mathcal{P})$, where the cell of $p \in \mathcal{P}$ is

$$\text{Vor}^\omega(p) = \{x \in \mathbb{R}^D : \|x - p\|^2 - \omega(p)^2 \leq \|x - q\|^2 - \omega(q)^2, \forall q \in \mathcal{P}\}.$$

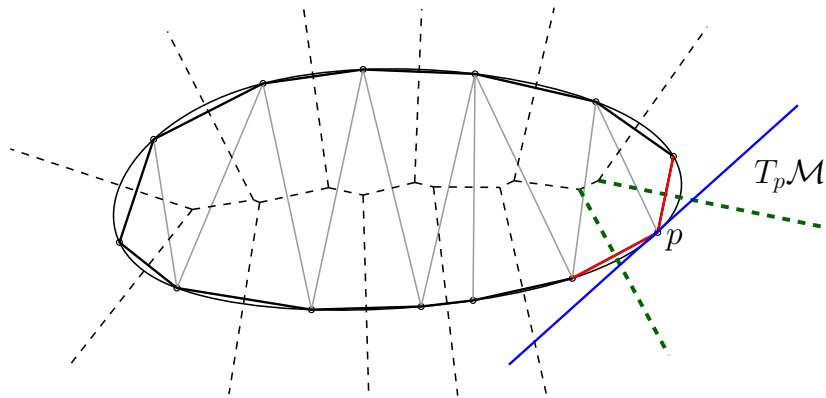
The *weighted Delaunay complex* $\text{Del}^\omega(\mathcal{P})$ is its dual triangulation.



Tangential complex

The tangential complex $\text{Del}^\omega(\mathcal{P}, T)$

$$\tau \in \text{Del}^\omega(\mathcal{P}, T) \Leftrightarrow \text{Vor}^\omega(\tau) \cap \left(\bigcup_{p \in \tau} T_p \mathcal{M} \right).$$



A Reconstruction Theorem

Theorem (Boissonnat, Ghosh 2014)

There exists ε_0 such that for all $\varepsilon \leq \varepsilon_0$, for all $\mathcal{P} \subset \mathcal{M}$, such that

- $d_H(\mathcal{P}, \mathcal{M}) \leq 2\varepsilon$,
- $d(p, \mathcal{P} \setminus \{p\}) \geq \varepsilon$ for all $p \in \mathcal{P}$,

there exists a weight function ω depending only on \mathcal{P} such that:

- $\text{Del}^\omega(\mathcal{P}, T)$ and \mathcal{M} are isotopic;
- $d_H(\text{Del}^\omega(\mathcal{P}, T), \mathcal{M}) \leq C\varepsilon^2$, where $C = C(d)$.

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Problem: The $T_p\mathcal{M}$'s are unknown.

\Rightarrow We replace each $T_p\mathcal{M}$ by an estimated version \tilde{T}_p .

Stability of the Tangential Complex

Theorem (Stability of the tangential complex)

Let $(\tilde{T}_p)_{p \in \mathcal{M}}$ be a family of affine spaces of dimension d , with $p \in \tilde{T}_p$ and $\angle(\tilde{T}_p, T_p \mathcal{M}) \leq \pi/32$, for all $p \in \mathcal{P}$.

Under the conditions of the reconstruction theorem,

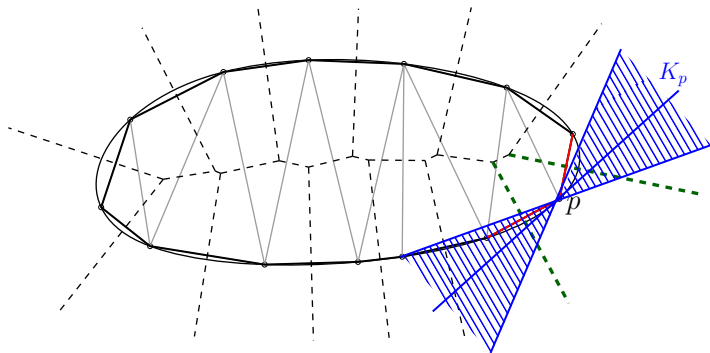
$$\mathrm{Del}^\omega(\mathcal{P}, \tilde{T}) = \mathrm{Del}^\omega(\mathcal{P}, T).$$

Stability: proof

The cocone complex $\text{Del}^\omega(\mathcal{P}, K)$

$$\tau \in \text{Del}^\omega(\mathcal{P}, K) \Leftrightarrow \text{Vor}^\omega(\tau) \cap \left(\bigcup_{p \in \tau} K_p \right),$$

where $K_p = \{q \in R^D, \angle(pq, T_p\mathcal{M}) \leq \pi/32\}$



$\text{Del}^\omega(\mathcal{P}, K)$ has dimension at most d .

Stability: Proof

$\text{Del}^\omega(\mathcal{P}, T), \text{Del}^\omega(\mathcal{P}, \tilde{T}) \subset \text{Del}^\omega(\mathcal{P}, K)$, and if there exists a d -simplex $\tau \in \text{Del}^\omega(\mathcal{P}, T)$ but $\tau \notin \text{Del}^\omega(\mathcal{P}, \tilde{\tau})$, we get the following configuration:

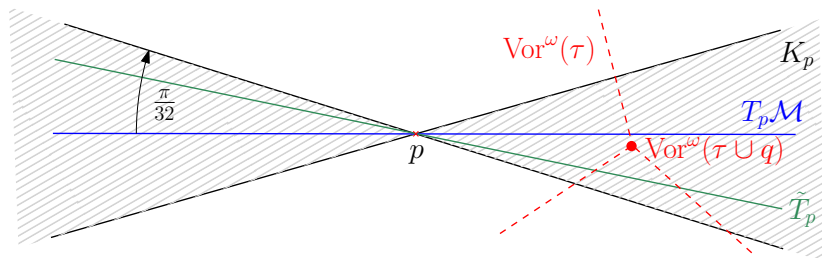


Figure: $\text{Vor}^\omega(\tau)$ cannot intersect K_p partially.

Statistical Model

$X_1, \dots, X_n \stackrel{i.i.d.}{\sim} P$, where $\mathcal{M} = \text{supp}(P) \subset \mathbb{R}^D$ is a connected d -sub-manifold satisfies:

- \mathcal{M} has no boundary,
- $\text{reach}(\mathcal{M}) \geq \rho > 0$,
- P has a twice differentiable density f with respect to μ , the measure induced on \mathcal{M} by the d -dimensional Hausdorff measure over \mathbb{R}^D , with:

$$\begin{cases} f_{min} \leq f(x) \leq f_{max} \\ \|d_x^2 f\| \leq H. \end{cases}$$

Remarks on the Model

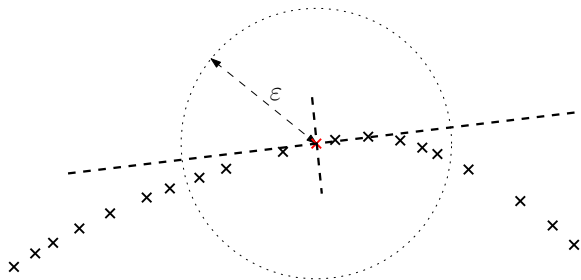
- P is (a, d) -standard: $P(\mathcal{B}(p, r) \geq ar^d)$ for small enough r .
Therefore,

$$\mathbb{E}d_H(\mathcal{M}, \{X_1, \dots, X_n\}) \leq C \left(\frac{\log n}{n} \right)^{1/d}.$$

- The model forces

$$\rho^d f_{min} \leq C_d, \quad \text{diam}(\mathcal{M}) \leq \frac{C'_d}{\rho^{d-1} f_{min}}.$$

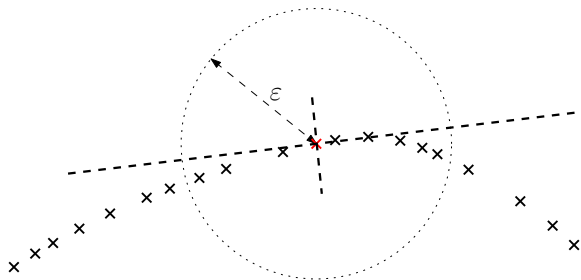
Tangent Space Estimation: Local P.C.A.



Define \hat{T}_j as the span of the d first eigenvectors of

$$\hat{O}_j = \frac{1}{n-1} \sum_{i \neq j} \mathbf{1}_{\|X_i - X_j\| < \epsilon} (X_i - X_j)(X_i - X_j)^T.$$

Tangent Space Estimation: Local P.C.A.



Theorem

Taking $\varepsilon \asymp \left(\frac{\log(n)}{n}\right)^{1/d}$, for n large enough, with probability larger than $1 - \left(\frac{1}{n}\right)^{1/d}$,

$$\max_j \angle(T_{X_j} \mathcal{M}, \hat{T}_j) \leq \frac{\pi}{32}.$$

Estimation Procedure & Convergence Rate

1. Estimate the $T_{X_j}\mathcal{M}$'s with local PCA.
2. Using the *farthest point sampling* algorithm, extract a sparse sample $\mathbb{Y}_n \subset \{X_1, \dots, X_n\}$.
3. Take as estimator $\hat{\mathcal{M}}$ the weighted Delaunay triangulation of \mathbb{Y}_n restricted to the estimated tangent spaces \hat{T}_j 's.

Theorem

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(d_{\text{H}}(\mathcal{M}, \hat{\mathcal{M}}) \leq \left(\frac{\log n}{n} \right)^{2/d} \text{ and } \mathcal{M} \cong \hat{\mathcal{M}} \right) = 1,$$

where \cong denotes the isotopy equivalence.

Moreover there exists a constant C such that for n large enough,

$$\mathbb{E}d_{\text{H}}(\mathcal{M}, \hat{\mathcal{M}}) \leq C \left(\frac{\log n}{n} \right)^{2/d}.$$

A Noisy Model: Clutter Noise

$$X \sim \beta P + (1 - \beta)U,$$

with $0 < \beta < 1$, P as previously and $U \sim \text{Uniform}(\mathcal{B}(0, M))$, for $\mathcal{B}(\mathcal{M}, \rho) \subset \mathcal{B}(0, M)$.

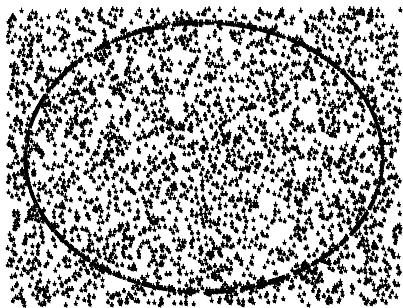
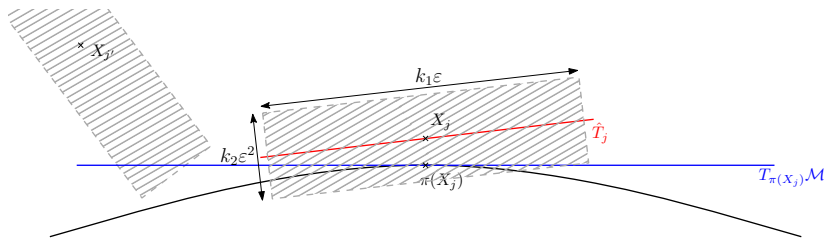


Figure: A realization of the clutter model

Clustering before Estimation

We define boxes S_j centred at each X_j :



To determine if $X_j \in \mathcal{M}$, consider $P_n(S_j) = |S_j \cap \{X_1, \dots, X_n\}|$.
As $\epsilon \rightarrow 0$,

$$P_n(S_j) \sim \begin{cases} \epsilon^D & \text{if } X_j \text{ is far from } \mathcal{M} \\ \epsilon^d \gg \epsilon^D & \text{if } X_j \in \mathcal{M} \end{cases}$$

Clustering Result

Proposition

There exists constants $k(d, D)$ and $t(d, D, \rho)$ such that, for n large enough, if

$$\varepsilon = k \left(\frac{\log(n)}{\beta n} \right)^{\frac{1}{d+2}},$$

then, with probability larger than $1 - \left(\frac{1}{n}\right)^{\frac{2}{d}} - \left(\frac{1}{n}\right)^{2D}$, we have

$$\left(\frac{n}{\log(n)} \right) P_n(S_j) \begin{cases} \leq t & \text{if } d(X_j, \mathcal{M}) \geq \varepsilon^2 \\ > t & \text{if } X_j \in \mathcal{M} \end{cases}$$

Moreover, on the same event, for every X_j such that $d(X_j, \mathcal{M}) \leq \sqrt{\frac{1}{2d+5}}\varepsilon$, we have

$$\angle(\hat{T}_j, T_{\pi(X_j)}\mathcal{M}) \leq \left(\frac{K_1(d, H)}{\rho} + \frac{K_2(d, D)}{\beta} \right) \varepsilon.$$

Reminder: The Reconstruction Theorem

Theorem (Boissonnat, Ghosh 2014)

There exists ε_0 such that for all $\varepsilon \leq \varepsilon_0$, for all $\mathcal{P} \subset \mathcal{M}$, such that

- $d_H(\mathcal{P}, \mathcal{M}) \leq 2\varepsilon$,
- $d(p, \mathcal{P} \setminus \{p\}) \geq \varepsilon$ for all $p \in \mathcal{P}$,

there exists a weight function ω depending only on \mathcal{P} such that:

- $\text{Del}^\omega(\mathcal{P}, T)$ and \mathcal{M} are isotopic;
- $d_H(\text{Del}^\omega(\mathcal{P}, T), \mathcal{M}) \leq C\varepsilon^2$, where $C = C(d)$.

Interpolation Lemma

Lemma

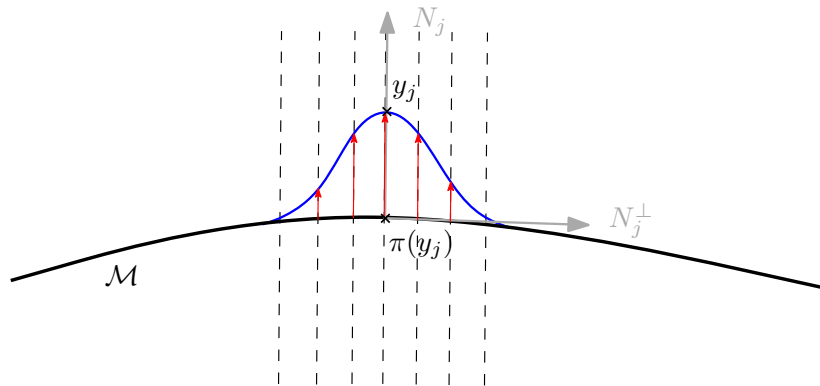
Let $\mathbb{Y} = \{y_1, \dots, y_q\} \subset \mathbb{R}^D$ be a sample such that:

- \mathbb{Y} ε -approximates \mathcal{M} : $d_H(\mathbb{Y}, \mathcal{M}) < \varepsilon$,
- \mathbb{Y} is δ -sparse: $\min_{i \neq j} \|y_j - y_i\| \geq \delta > 0$ for all j ,
- the y_j 's are η -close to \mathcal{M} : $\max_{1 \leq j \leq q} d(y_j, \mathcal{M}) < \eta$,

Provided that $\eta < \text{reach}(\mathcal{M}) \wedge \delta/25$, there exists a smooth sub-manifold $\mathcal{M}' \subset \mathbb{R}^D$ such that:

1. \mathcal{M}' interpolates the points: $\mathbb{Y} \subset \mathcal{M}'$;
2. $T_{y_j} \mathcal{M}' = (y_j - \pi(y_j)) + T_{\pi(y_j)} \mathcal{M}$ for all $1 \leq j \leq q$;
3. $d_H(\mathbb{Y}, \mathcal{M}') \leq \varepsilon + \eta$;
4. \mathcal{M} and \mathcal{M}' are ambient isotopic;
5. $\text{reach}(\mathcal{M}') \geq \rho \cdot \frac{(1 - 25\frac{\eta}{\delta})^2}{1 + 1475\frac{\eta}{\delta^2}\rho + 25\frac{\eta}{\delta}}$.

Interpolation Lemma: Proof



Interpolation Lemma

Proposition

Taking

$$\varepsilon \asymp \left(\frac{\log(n)}{n} \right)^{\frac{1}{d+2}}, \quad \delta = \varepsilon/2, \quad \eta \asymp \varepsilon^2$$

in last lemma, we get for n large enough,

1. \mathcal{M}' interpolates the points: $\mathbb{Y} \subset \mathcal{M}'$;
2. $T_{y_j} \mathcal{M}' = (y_j - \pi(y_j)) + T_{\pi(y_j)} \mathcal{M}$ for all $1 \leq j \leq q$;
3. $d_H(\mathbb{Y}, \mathcal{M}') \leq 2\varepsilon$;
4. \mathcal{M} and \mathcal{M}' are ambient isotopic;
5. $\text{reach}(\mathcal{M}') \geq C\rho$.

Convergence Result

1. Partition the sample into noise/data with slab counting,
2. Using the *farthest point sampling* algorithm, extract a sparse sample $\mathbb{Y}_n \subset \{X_1, \dots, X_n\}$.
3. Take as estimator $\hat{\mathcal{M}}$ the weighted Delaunay triangulation of \mathbb{Y}_n restricted to the estimated tangent spaces \hat{T}_j 's.

Theorem

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(d_{\text{H}}(\mathcal{M}, \hat{\mathcal{M}}) \leq \left(\frac{\log n}{n} \right)^{2/(d+2)} \text{ and } \mathcal{M} \cong \hat{\mathcal{M}} \right) = 1,$$

where \cong denotes the isotopy equivalence.

Moreover there exists a constant C such that for n large enough,

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