

Manifold Reconstruction

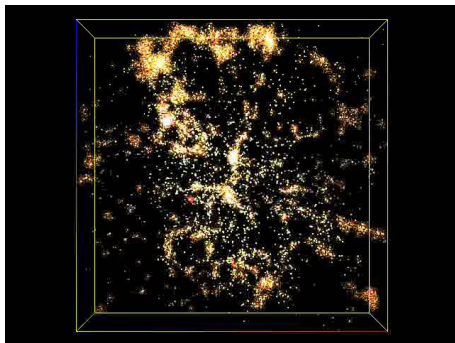
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Geometric data analysis

Images, text, speech, neural signals, GPS traces,...



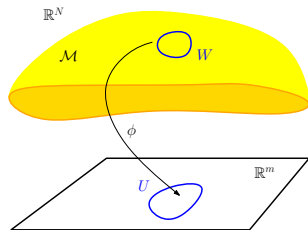
Geometrisation : Data = points + distances between points

Hypothesis : Data lie close to a structure of
“small” intrinsic dimension

Problem : Infer the structure from the data

Submanifolds of \mathbb{R}^d

A compact subset $M \subset \mathbb{R}^d$ is a submanifold without boundary of (intrinsic) dimension $k < d$, if any $p \in M$ has an open (topological) k -ball as a neighborhood in M



Intuitively, a submanifold of dimension k is a subset of \mathbb{R}^d that looks locally like an open set of an affine space of dimension k

A **curve** is a 1-dimensional submanifold

A **surface** is a 2-dimensional submanifold

Triangulation of a submanifold

We call triangulation of a submanifold $M \subset \mathbb{R}^d$ a (geometric) simplicial complex \hat{M} such that

- \hat{M} is embedded in \mathbb{R}^d
- its vertices are on M
- it is homeomorphic to M

Submanifold reconstruction

The problem is to construct a triangulation \hat{M} of some unknown submanifold M given a finite set of points $P \subset M$

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Issues in high-dimensional geometry

- Dimensionality severely restricts our intuition and ability to visualize data
 - ⇒ need for automated and provably correct methods
- Complexity of data structures and algorithms rapidly grow as the dimensionality increases
 - ⇒ no subdivision of the ambient space is affordable
 - ⇒ data structures and algorithms should be sensitive to the **intrinsic dimension** (usually unknown) of the data
- Inherent defects : sparsity, noise, outliers

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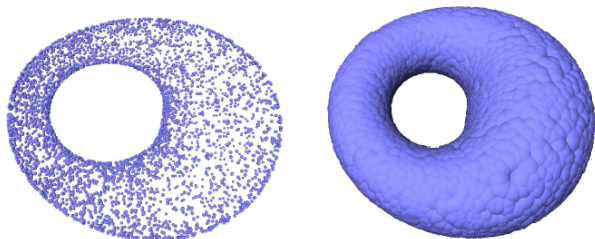
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Looking for small and faithful simplicial complexes

Need to compromise

- Size of the complex
 - ▶ can we have $\dim \hat{\mathbb{M}} = \dim \mathbb{M}$?
- Efficiency of the construction algorithms and of the representations
 - ▶ can we avoid the exponential dependence on d ?
 - ▶ can we minimize the number of simplices ?
- Quality of the approximation
 - ▶ Homotopy type & homology (RIPS complex, persistence)
 - ▶ Homeomorphism (Delaunay-type complexes)

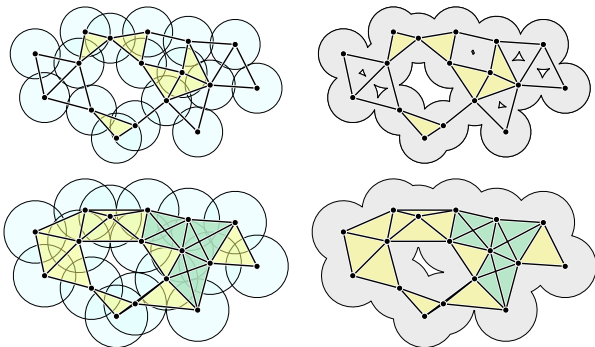
Distance to a compact K : $d_K : x \rightarrow \inf_{p \in K} \|x - p\|$



Stability

If the data points C are close (Hausdorff) to the geometric structure K , the topology and the geometry of the offsets $K_r = d^{-1}([0, r])$ and $C_r = d^{-1}([0, r])$ are close

Distance functions and triangulations



Nerve theorem (Leray)

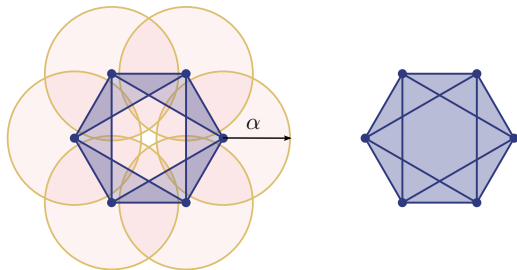
The nerve of the balls (Cech complex) and the union of balls have the same homotopy type

Questions

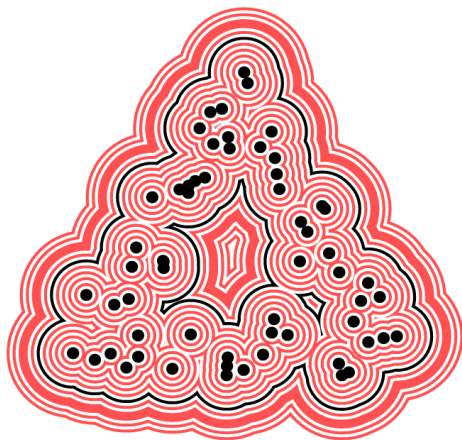
- + The homotopy type of a compact set X can be computed from the Čech complex of a sample of X
- + The same is true for the α -complex
- The Čech and the α -complexes are **huge** ($O(n^d)$ and $O(n^{\lceil d/2 \rceil})$) and very difficult to compute
- Both complexes are **not** in general homeomorphic to X (i.e. not **a triangulation** of X)
- The Čech complex **cannot be realized** in general in the same space as X

Čech and Rips complexes

The Rips complex is easier to compute but still very big, and less precise in approximating the topology



An example where no offset has the right topology !

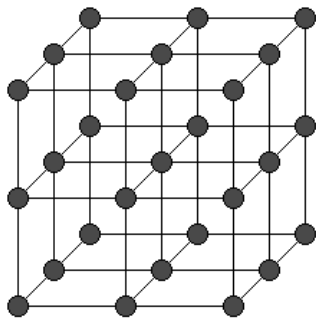


Persistent homology at rescue !

The curses of Delaunay triangulations in higher dimensions

- Their complexity depends exponentially on the ambient dimension. Robustness issues become very tricky
- Higher dimensional Delaunay triangulations are not thick even if the vertices are well-spaced
- The restricted Delaunay triangulation is no longer a good approximation of the manifold even under strong sampling conditions (for $d > 2$)

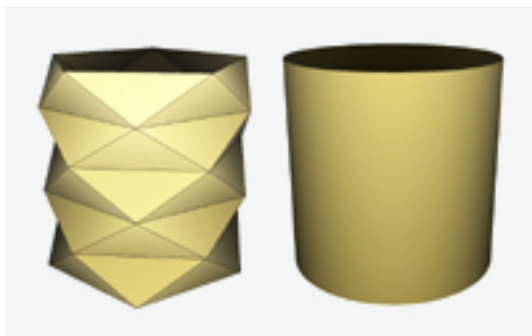
3D Delaunay Triangulations are not thick even if the vertices are well-spaced



- Each square face can be circumscribed by an empty sphere
- This remains true if the grid points are slightly perturbed therefore creating thin simplices

Badly-shaped simplices

Badly-shaped simplices lead to bad geometric approximations



which in turn may lead to topological defects in $\text{Del}_{|\mathcal{M}}(\mathbf{P})$ [Oudot]

see also [Cairns], [Whitehead], [Munkres], [Whitney]

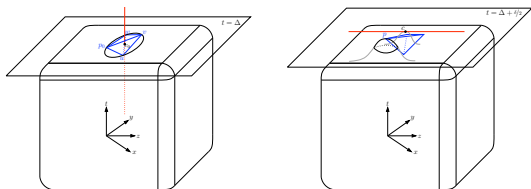
Tangent space approximation

\mathbb{M} is a smooth k -dimensional manifold ($k > 2$) embedded in \mathbb{R}^d

Bad news

[Oudot 2005]

The Delaunay triangulation restricted to \mathbb{M} may be a bad approximation of the manifold even if the sample is dense



Whitney's angle bound and tangent space approximation

Lemma

[Whitney 1957]

If σ is a j -simplex whose vertices all lie within a distance η from a hyperplane $H \subset \mathbb{R}^d$, then

$$\sin \angle(\text{aff}(\sigma), H) \leq \frac{2j\eta}{D(\sigma)}$$

Corollary

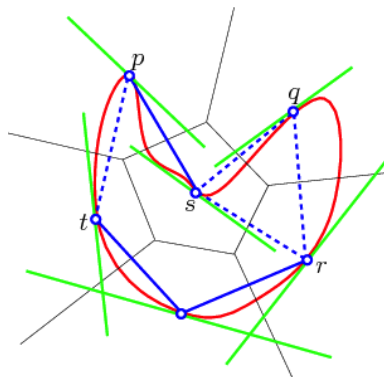
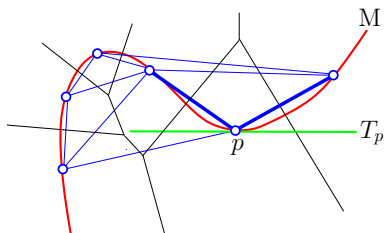
If σ is a j -simplex, $j \leq k$, $\text{vert}(\sigma) \subset \mathbb{M}$, $\Delta(\sigma) \leq \delta \text{rch}(\mathbb{M})$

$$\forall p \in \sigma, \quad \sin \angle(\text{aff}(\sigma), T_p) \leq \frac{\delta}{\Theta(\sigma)}$$

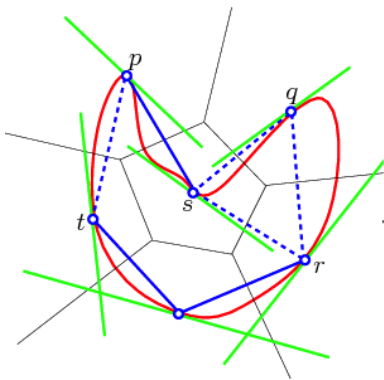
($\eta \leq \frac{\Delta(\sigma)^2}{2 \text{rch}(\mathbb{M})}$ by the Chord Lemma)

The assumptions

- \mathbb{M} is a differentiable submanifold of **positive reach** of \mathbb{R}^d
- The dimension k of \mathbb{M} is **small**
- \mathbf{P} is an **ε -net** of \mathbb{M} , i.e.
 - ▶ $\forall x \in \mathbb{M}, \exists p \in \mathbf{P}, \|x - p\| \leq \varepsilon \operatorname{rch}(\mathbb{M})$
 - ▶ $\forall p, q \in \mathbf{P}, \|p - q\| \geq \bar{\eta} \varepsilon$
- ε is small enough



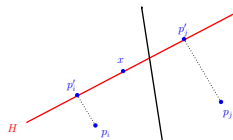
- 1 Construct the star of $p \in P$ in the Delaunay triangulation $\text{Del}_{T_p}(P)$ of P restricted to T_p
- 2 $\text{Del}_{T_M}(P) = \bigcup_{p \in P} \text{star}(p)$



- + $\text{Del}_{TM}(\mathbf{P}) \subset \text{Del}(\mathbf{P})$
- + $\text{star}(p)$, $\text{Del}_{T_p}(\mathbf{P})$ and therefore $\text{Del}_{TM}(\mathbf{P})$ can be computed without computing $\text{Del}(\mathbf{P})$
- $\text{Del}_{TM}(\mathbf{P})$ is **not** necessarily a triangulated manifold

Construction of $\text{Del}_{T_p}(\mathbf{P})$

Given a d -flat $H \subset \mathbb{R}^d$, $\text{Vor}(\mathbf{P}) \cap H$ is a **weighted** Voronoi diagram in H



$$\|x - p_i\|^2 \leq \|x - p_j\|^2$$

$$\Leftrightarrow \|x - p'_i\|^2 + \|p_i - p'_i\|^2 \leq \|x - p'_j\|^2 + \|p_j - p'_j\|^2$$

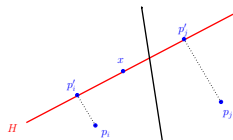
Corollary: construction of Del_{T_p}

$$\psi_p(p_i) = (p'_i, -\|p_i - p'_i\|^2) \quad (\text{weighted point})$$

- 1 project \mathbf{P} onto T_p which requires $O(Dn)$ time
- 2 construct $\text{star}(\psi_p(p_i))$ in $\text{Del}(\psi_p(p_i)) \subset T_p$
- 3 $\text{star}(p_i) \approx \text{star}(\psi_p(p_i))$ (isomorphic)

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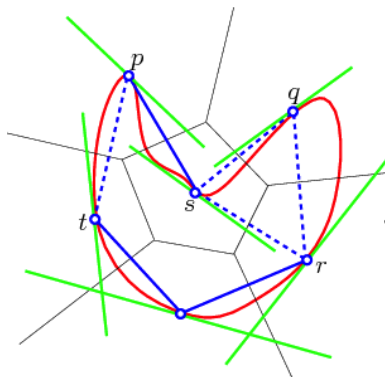
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Inconsistencies in the tangential complex

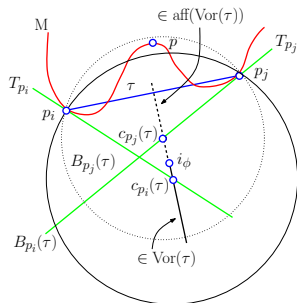
A simplex is **not** in the star of all its vertices



● $\tau \in \text{star}(p_i) \iff T_{p_i} \cap \text{Vor}(\tau) \neq \emptyset \iff B(c_{p_i}(\tau) \cap \mathbf{P} = \emptyset$

● $\tau \notin \text{star}(p_j) \iff T_{p_j} \cap \text{Vor}(\tau) = \emptyset \iff B(c_{p_j}(\tau) \cap \mathbf{P} \ni p$

Inconsistent thick simplices are not well-protected



If τ is small and thick, then

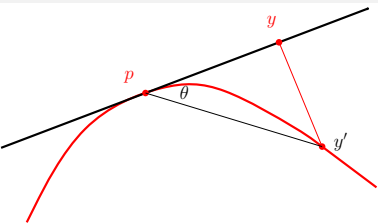
- $T_{p_i} \approx T_{p_j} \approx \text{aff}(\tau)$ \Leftarrow sample density
- $\|c_{p_i} - c_{p_j}\|$ small $\Rightarrow B_{ij} := B_{p_i}(\tau) \setminus B_{p_j}(\tau)$ small \Leftarrow thickness
- $\exists p \in \mathbf{P} \cap B_{ij}$

Inconsistent thick simplices are not well protected

Bound on $\Delta(\tau)$

- (i) $\text{Vor}(p) \cap T_p \subseteq B(p, \alpha \text{rch}(\mathbb{M}))$ where α is the smallest positive root of $\alpha (1 - \tan(\arcsin \frac{\alpha}{2})) = \varepsilon$ ($\alpha \approx \varepsilon$)
- (ii) $\forall \tau \in \text{star}(p), R_p(\tau) \leq \alpha \text{rch}(\mathbb{M})$
- (iii) $\forall \tau \in \text{Del}_{T\mathbb{M}}(\mathbf{P}), \Delta(\tau) \leq 2\alpha \text{rch}(\mathbb{M})$.

Proof of (i)



By contradiction:

$\exists x \in \text{Vor}(p) \cap T_p$ s.t. $\|p - x\| > \alpha \text{rch}(\mathbb{M})$

$y : y \in [xp]$ and $\|y - p\| = \alpha \text{rch}(\mathbb{M})$

by convexity, $y \in \text{intVor}(p) \cap T_p$.

$y' : y' \in \mathbb{M}$, whose closest point on T_p is y

$\theta := \angle(py', T_p)$

By the Chord Lemma, $\sin \theta \leq \frac{\|p - y'\|}{2\text{rch}(\mathbb{M})} = \frac{\|p - y\|}{2\text{rch}(\mathbb{M}) \cos \theta} \Rightarrow \sin 2\theta \leq \alpha$.

$$\|y - y'\| = \|p - y\| \tan \omega \leq \alpha \text{rch}(\mathbb{M}) \tan(\arcsin \frac{\alpha}{2})$$

Since \mathbf{P} is an ε -sample, $\exists t \in \mathbf{P}$, s.t. $\|y' - t\| \leq \varepsilon \text{rch}(M)$. Hence

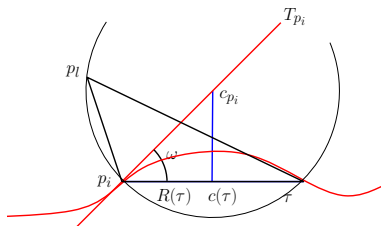
$$\begin{aligned} \|y - t\| &\leq \|y - y'\| + \|y' - t\| \leq (\alpha \tan(\arcsin \frac{\alpha}{2}) + \varepsilon) \text{rch}(\mathbb{M}) \\ &= \alpha \text{rch}(\mathbb{M}) = \|y - p\|. \end{aligned} \tag{1}$$

Hence $y \notin \text{intVor}(p)$, which contradicts our assumption and proves (i).

Inconsistent thick simplices are not well protected

If τ is an inconsistent k -simplex and $\omega = \angle(\text{aff}(\tau), T_{p_i})$, then

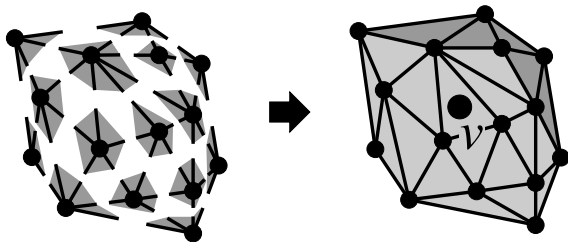
$$\sin \omega \leq \frac{\Delta(\tau)}{\Theta(\tau) \text{rch}(M)} \Rightarrow \|c_{p_i} - c_{p_j}\| \leq 2R(\tau) \tan \omega \approx \frac{4\varepsilon^2 \text{rch}(M)}{\Theta(\tau)}$$



$$p_l \in B(c_{p_i}, R_{p_i}(\tau) + \delta) \setminus B(c_{p_i}, R_{p_i}(\tau)) \text{ where } \delta = \frac{4\varepsilon^2 \text{rch}(M)}{\Theta(\tau)}$$

Reconstruction of smooth submanifolds

- 1 For each vertex v , compute the star $\text{star}(p)$ of p in $\text{Del}_p(P)$
- 2 Remove inconsistencies among the stars by perturbing either the points or by weighting the points
- 3 Stitch the stars to obtain a triangulation of P



Algorithm hypotheses

Known quantities in red

- \mathbb{M} = a differentiable submanifold of positive reach of dim. $k \subset \mathbb{R}^d$
- \mathbf{P} = an (ε, δ) -sample of \mathbb{M}
- $\varepsilon \leq \varepsilon_0$
- $\varepsilon/\delta \leq \eta_0$
- we can estimate the tangent space T_p at any $p \in \mathbf{P}$

Manifold reconstruction algorithm via perturbation

Picking regions : pick p' in $B(p, \rho)$

Sampling parameters of a perturbed point set

If P is an $(\varepsilon, \bar{\eta})$ -net, P' is an $(\varepsilon', \bar{\eta}')$ -net, where

$$\varepsilon' = \varepsilon(1 + \bar{\rho}) \quad \text{and} \quad \bar{\eta}' = \frac{\bar{\eta} - 2\bar{\rho}}{1 + \bar{\rho}}$$

Notation : $\bar{x} = \frac{x}{\varepsilon}$

The LLL framework

Random variables : P' a set of random points $\{p', p' \in B(p, \rho), p \in P\}$

Events: Type 1 : σ' s.t. $\Theta(\sigma') < \Theta_0$

Type 2 : $\phi' = (\sigma', p', q', l')$ s.t. **(Bad configuration)**

1. σ' is an inconsistent k -simplex
2. $p', q' \in \sigma'$
3. $\sigma' \in \text{star}(p')$
4. $\sigma' \notin \text{star}(q')$
5. $l' \in P' \setminus \sigma' \wedge l' \in B_q(\sigma')$ (the ball centered on T_q that cc σ')

Algorithm

input: P, ρ, Θ_0

while an event ϕ' occurs **do**

 resample the points of ϕ'

 update $\text{Del}(P')$

output: P' and $\text{Del}_{TM}(P')$

Summary

• Termination

- ▶ If $\frac{\bar{\eta}}{2} \geq \bar{\rho} \geq f(\Theta_0)$, the algorithm terminates and returns a complex $\hat{\mathbb{M}}$ that has no inconsistent configurations

• Complexity

- ▶ No d -dimensional data structure \Rightarrow linear in d
- ▶ exponential in k

• Approximation

- ▶ $\hat{\mathbb{M}}$ is a PL simplicial k -manifold
- ▶ $\hat{\mathbb{M}} \subset \text{tub}(\mathbb{M}, \varepsilon)$
- ▶ $\hat{\mathbb{M}}$ is homeomorphic to \mathbb{M}

$\hat{\mathbb{M}}$ is a PL simplicial k -manifold

Lemma Let P be an ε -sample of a manifold \mathbb{M} and let $p \in P$. The link of any vertex p in $\hat{\mathbb{M}}$ is a topological $(k - 1)$ -sphere

Proof :

1. Since $\hat{\mathbb{M}}$ contains no inconsistencies, the star of any vertex p in $\hat{\mathbb{M}}$ is identical to $\text{star}(p)$, the star of p in $\text{Del}_p(P)$
2. $\text{Del}_p(P) \subset \mathbb{R}^d \approx \text{Del}(\psi_p(P)) \subset T_p \Rightarrow \text{star}(p) \approx \text{star}_p(p)$
3. $\text{star}_p(p)$ is a k -dimensional triangulated topological ball (general position)
4. p cannot belong to the boundary of $\text{star}_p(p)$
(the Voronoi cell of $p = \psi_p(p)$ in $\text{Vor}(\psi_p(P))$ is bounded)

Applications and extensions

ERC Advanced Grant GUDHI

- Anisotropic mesh generation
- Discrete metric sets (see the previous lecture on the witness complex)
- Stratified manifolds
- Non euclidean embedding space (e.g. statistical manifolds)