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# Manifold reconstruction using tangential Delaunay complexes

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## Abstract

We give a provably correct algorithm to reconstruct a  $k$ -dimensional smooth manifold embedded in  $d$ -dimensional Euclidean space. The input to our algorithm is a point sample coming from an unknown manifold. Our approach is based on two main ideas : the notion of tangential Delaunay complex defined in [BF04, Flö03, Fre02], and the technique of sliver removal by weighting the sample points [CDE<sup>+</sup>00]. Differently from previous methods, we do not construct any subdivision of the  $d$ -dimensional ambient space. As a result, the running time of our algorithm depends only linearly on the extrinsic dimension  $d$  while it depends quadratically on the size of the input sample, and exponentially on the intrinsic dimension  $k$ . To the best of our knowledge, this is the first certified algorithm for manifold reconstruction whose complexity depends linearly on the ambient dimension. We also prove that for a dense enough sample the output of our algorithm is isotopic to the manifold and a close geometric approximation of the manifold. A preliminary version of this paper appeared in the proceedings of the Symposium on Computational Geometry, SoCG 2010.

*Keywords:* Tangential Delaunay complex, weighted Delaunay triangulation, manifold reconstruction, manifold learning, sampling conditions, sliver exudation.

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# 1 Introduction

Manifold reconstruction consists of computing a piecewise linear approximation of an unknown manifold  $\mathcal{M} \subset \mathbb{R}^d$  from a finite sample of unorganized points  $\mathcal{P}$  lying on  $\mathcal{M}$  or close to  $\mathcal{M}$ . When the manifold is a two-dimensional surface embedded in  $\mathbb{R}^3$ , the problem is known as the surface reconstruction problem. Surface reconstruction is a problem of major practical interest which has been extensively studied in the fields of Computational Geometry, Computer Graphics and Computer Vision. In the last decade, solid foundations have been established and the problem is now pretty well understood. Refer to Dey's book [Dey06], and the survey by Cazals and Giesen in [CG06] for recent results. The output of those methods is a triangulated surface that approximates  $\mathcal{M}$ . This triangulated surface is usually extracted from a 3-dimensional subdivision of the ambient space (typically a grid or a triangulation). Although rather inoffensive in 3-dimensional space, such data structures depend exponentially on the dimension of the ambient space, and all attempts to extend those geometric approaches to more general manifolds have led to algorithms whose complexities depend exponentially on  $d$  [BGO09, CL08, CDR05, NSW08].

The problem in higher dimensions is also of great practical interest in data analysis and machine learning. In those fields, the general assumption is that, even if the data are represented as points in a very high dimensional space  $\mathbb{R}^d$ , they in fact live on a manifold of much smaller intrinsic dimension [SL00]. If the manifold is linear, well-known global techniques like principal component analysis (PCA) or multi-dimensional scaling (MDS) can be efficiently applied. When the manifold is highly nonlinear, several more local techniques have attracted much attention in visual perception and

many other areas of science. Among the prominent algorithms are Isomap [TdSL00], LLE [RS00], Laplacian eigenmaps [BN02], Hessian eigenmaps [DG03], diffusion maps [LL06, NLCK05], principal manifolds [ZZ04]. Most of those methods reduce to computing an eigendecomposition of some connection matrix. In all cases, the output is a mapping of the original data points into  $\mathbb{R}^k$  where  $k$  is the estimated intrinsic dimension of  $\mathcal{M}$ . Those methods come with no or very limited guarantees. For example, Isomap provides a correct embedding only if  $\mathcal{M}$  is isometric to a convex open set of  $\mathbb{R}^k$  and LLE can only reconstruct topological balls. To be able to better approximate the sampled manifold, another route is to extend the work on surface reconstruction and to construct a piecewise linear approximation of  $\mathcal{M}$  from the sample in such a way that, under appropriate sampling conditions, the quality of the approximation can be guaranteed. First breakthrough, along this line, was the work of Cheng, Dey and Ramos [CDR05] and this was followed by the work of Boissonnat, Guibas and Oudot [BGO09]. In both cases, however, the complexity of the algorithms is exponential in the ambient dimension  $d$ , which highly reduces their practical relevance.

In this paper, we extend the geometric techniques developed in small dimensions and propose an algorithm that can reconstruct smooth manifolds of arbitrary topology while avoiding the computation of data structures in the ambient space. We assume that  $\mathcal{M}$  is a smooth manifold of known dimension  $k$  and that we can compute the tangent space to  $\mathcal{M}$  at any sample point. Under those conditions, we propose a provably correct algorithm that constructs a simplicial complex of dimension  $k$  that approximates  $\mathcal{M}$ . The complexity of the algorithm is linear in  $d$ , quadratic in the size  $n$  of the sample, and exponential in  $k$ . Our work builds on [BGO09] and [CDR05] but dramatically reduces the dependence on  $d$ . To the best of our knowledge, this is the first certified algorithm for manifold reconstruction whose complexity depends only linearly on the ambient dimension. In the same spirit, Chazal and Oudot [CO08] have devised an algorithm of intrinsic complexity to solve the easier problem of computing the homology of a manifold from a sample.

Our approach is based on two main ideas : the notion of *tangential Delaunay complex* introduced in [BF04, Flö03, Fre02], and the technique of sliver removal by weighting the sample points [CDE<sup>+</sup>00]. The tangential complex is obtained by gluing local (Delaunay) triangulations around each sample point. The tangential complex is a subcomplex of the  $d$ -dimensional Delaunay triangulation of the sample points but it can be computed using mostly operations in the  $k$ -dimensional tangent spaces at the sample points. Hence the dependence on  $k$  rather than  $d$  in the complexity. However, due to the presence of so-called inconsistencies, the local triangulations may not form a triangulated manifold. Although this problem has already been reported [Fre02], no solution was known except for the case of curves ( $k = 1$ ) [Flö03]. The idea of removing inconsistencies among local triangulations that have been computed independently has already been used for maintaining dynamic meshes [She05] and generating anisotropic meshes [BWY08]. Our approach is close in spirit to the one in [BWY08]. We show that, under appropriate sample conditions, we can remove inconsistencies by weighting the sample points. We can then prove that the approximation returned by our algorithm is ambient isotopic to  $\mathcal{M}$ , and a close geometric approximation of  $\mathcal{M}$ .

Our algorithm can be seen as a *local* version of the cocone algorithm of Cheng et al. [CDR05]. By local, we mean that we do not compute any  $d$ -dimensional data structure like a grid or a triangulation of the ambient space. Still, the tangential complex is a subcomplex of the weighted  $d$ -dimensional Delaunay triangulation of the (weighted) data points and therefore implicitly relies on a global partition of the ambient space. This is a key to our analysis and distinguishes our method from the other local algorithms that have been proposed in the surface reconstruction literature [CSD04, GKS00].

## Organisation of the paper

In Section 2, we introduce the basic concepts used in this paper. We recall the notion of weighted Voronoi (or power) diagrams and Delaunay triangulations in Section 2.1 and define sampling conditions in Section 2.2. We introduce various quantities to measure the shape of simplices in Section 2.3 and, in particular, the central notion of fatness. In Section 2.4, we define the two main notions of this paper: the tangential complex and inconsistent configurations.

The algorithmic part of the paper is given in Section 3.

The main structural results are given in Section 4. Under some sampling condition, we bound the shape measure of the simplices of the tangential complex in Section 4.2 and of inconsistent configurations in Section 4.3. A crucial fact is that inconsistent configurations cannot be fat. We also bound the number of simplices and inconsistent configurations that can be incident on a point in Section 4.4. In Sections 4.5 and 4.6, we prove the correctness of the algorithm, and space and time complexity respectively. In Section 5, we prove that the simplicial complex output by the algorithm is indeed a good approximation of the sampled manifold. Finally, in Section 6, we conclude with some possible extensions.

The list of main notations has been added at the end of the paper as a reference for the readers.

## 2 Definitions and preliminaries

The standard Euclidean distance between two points  $x, y \in \mathbb{R}^d$  will be denoted by  $\|x - y\|$ . *Distance* between a point  $x \in \mathbb{R}^d$  and a set  $X \subset \mathbb{R}^d$ , is defined as

$$\text{dist}(x, X) = \inf_{p \in X} \|x - p\|.$$

We will denote the topological boundary and interior of a set  $X \subseteq \mathbb{R}^d$  by  $\partial X$  and  $\text{int } X$  respectively.

For a map  $f : X \rightarrow Y$  and  $X_1 \subseteq X$ ,  $f|_{X_1}$  denotes the restriction of map  $f$  to the subset  $X_1$ .

A  $j$ -simplex is the convex hull of  $j + 1$  affinely independent points. If  $\tau$  is a  $j$ -simplex with vertices  $\{x_0, \dots, x_j\}$ , we will also denote  $\tau$  as  $[x_0, \dots, x_j]$ . For convenience, we often identify a simplex and the set of its vertices. Hence, if  $\tau$  is a simplex,  $p \in \tau$  means that  $p$  is a vertex of  $\tau$ . If  $\tau$  is a  $j$ -simplex,  $\text{aff}(\tau)$  denotes the  $j$ -dimensional affine hull of  $\tau$  and  $N_\tau$  denotes the  $(d - j)$ -dimensional normal space of  $\text{aff}(\tau)$ .

For a given simplicial complex  $\mathcal{K}$  and a simplex  $\sigma$  in  $\mathcal{K}$ , the *star* and *link* of a simplex  $\sigma$  in  $\mathcal{K}$   $\sigma \in \mathcal{K}$  denotes the subcomplexes

$$\text{st}(\sigma, \mathcal{K}) = \{\tau : \text{for some } \tau_1 \in \mathcal{K}, \sigma, \tau \subseteq \tau_1\}$$

and

$$\text{lk}(\sigma, \mathcal{K}) = \{\tau \in \text{st}(\sigma, \mathcal{K}) : \tau \cap \sigma = \emptyset\}$$

respectively. The *open star* of  $\sigma$  in  $\mathcal{K}$ ,  $\mathring{\text{st}}(\sigma, \mathcal{K}) = \text{st}(\sigma, \mathcal{K}) \setminus \text{lk}(\sigma, \mathcal{K})$ .

For all  $x, y$  in  $\mathbb{R}^d$ ,  $[xy]$  denotes the line segment joining the points  $x$  and  $y$ .

In this paper,  $\mathcal{M}$  denotes a differentiable manifold of dimension  $k$  embedded in  $\mathbb{R}^d$  and  $\mathcal{P} = \{p_1, \dots, p_n\}$  a finite sample of points from  $\mathcal{M}$ . We will further assume that  $\mathcal{M}$  has positive reach (see Section 2.2). We denote by  $T_p$  and  $N_p$  the  $k$ -dimensional tangent space and  $(d - k)$ -dimensional normal space at point  $p \in \mathcal{M}$  respectively.

For a given  $p \in \mathbb{R}^d$  and  $r \geq 0$ ,  $B(p, r)$  ( $\bar{B}(p, r)$ ) denotes the  $d$ -dimensional Euclidean open (close) ball centered at  $p$  of radius  $r$ , and  $B_{\mathcal{M}}(p, r)$  ( $\bar{B}_{\mathcal{M}}(p, r)$ ) denotes  $B(p, r) \cap \mathcal{M}$  ( $\bar{B}(p, r) \cap \mathcal{M}$ ).

For a given  $p \in \mathcal{P}$ ,  $\mathfrak{m}(p)$  denotes the distance of  $p$  to its nearest neighbor in  $\mathcal{P} \setminus \{p\}$ , i.e.,

$$\mathfrak{m}(p) = \min_{x \in \mathcal{P}, x \neq p} \|x - p\|.$$

If  $U$  and  $V$  are vector subspaces of  $\mathbb{R}^d$ , with  $\dim(U) \leq \dim(V)$ , the *angle* between them is defined by

$$\angle(U, V) = \max_{u \in U} \min_{v \in V} \angle(u, v),$$

where  $u$  and  $v$  are vectors in  $U$  and  $V$  respectively. This is the largest *principal angle* between  $U$  and  $V$ . In case of affine spaces, the angle between two of them is defined as the angle between the corresponding parallel vector subspaces.

The following lemma is a direct consequence of the above definition. We have included the proof for completeness.

**Lemma 2.1** *Let  $U$  and  $V$  be affine spaces of  $\mathbb{R}^d$  with  $\dim(U) \leq \dim(V)$ , and let  $U^\perp$  and  $V^\perp$  be affine spaces of  $\mathbb{R}^d$  with  $\dim(U^\perp) = d - \dim(U)$  and  $\dim(V^\perp) = d - \dim(V)$ .*

1. *If  $U^\perp$  and  $V^\perp$  are the orthogonal complements of  $U$  and  $V$  in  $\mathbb{R}^d$ , then  $\angle(U, V) = \angle(V^\perp, U^\perp)$ .*
2. *If  $\dim(U) = \dim(V)$  then  $\angle(U, V) = \angle(V, U)$ .*

**Proof.** Without loss of generality we assume that the affine spaces  $U$ ,  $V$ ,  $U^\perp$  and  $V^\perp$  are vector subspaces of  $\mathbb{R}^d$ , i.e. they all pass through the origin.

1. Suppose  $\angle(U, V) = \alpha$ . Let  $v_* \in V^\perp$  be a unit vector. There are unit vectors  $u \in U$ , and  $u_* \in U^\perp$  such that  $v_* = au + bu_*$ . We will show that  $\angle(v_*, u_*) \leq \alpha$ . First note that this angle is complementary to  $\angle(v_*, u)$ , i.e.,

$$\angle(v_*, u_*) = \frac{\pi}{2} - \angle(v_*, u). \quad (1)$$

There is a unit vector  $v \in V$  such that  $\angle(u, v) = \alpha_0 \leq \alpha$ . Viewing angles between unit vectors as distances on the unit sphere, we exploit the triangle inequality:  $\angle(v_*, v) \leq \angle(v_*, u) + \angle(u, v)$ , from whence

$$\angle(v_*, u) \geq \frac{\pi}{2} - \alpha_0.$$

Using this expression in Eq. (1), we find

$$\angle(v_*, u_*) \leq \alpha_0 \leq \alpha,$$

which implies, since  $v_*$  was chosen arbitrarily, that  $\angle(V^\perp, U^\perp) \leq \angle(U, V)$ .

Since  $\dim V^\perp \leq \dim U^\perp$ , and the orthogonal complement is a symmetric relation on subspaces, the same argument yields the reverse inequality.

2. Let  $\angle(U, V) = \alpha$ , and let  $P : U \rightarrow V$  denotes the projection map of the vector space  $U$  on  $V$ .

Case a.  $\alpha \neq \pi/2$ . Since  $\alpha \neq \pi/2$  and  $\dim(U) = \dim(V)$ , the projection map  $P$  is an isomorphism between vector spaces  $U$  and  $V$ . Therefore, for any unit vector  $v \in V$  there exist a vector  $u \in U$  such that  $P(u) = v$ . From the definition of angle between affine space and the linear map  $P$ , we

have  $\angle(u, v) (= \angle(v, u))$ . This implies,  $\angle(V, U) \leq \angle(U, V) = \alpha$ . Using the same arguments, we can show that  $\angle(U, V) \leq \angle(V, U)$  hence  $\angle(U, V) = \angle(V, U)$ .

Case b.  $\alpha = \pi/2$ . We have  $\angle(V, U) = \pi/2$ , otherwise using the same arguments as in Case 1 we can show that  $\alpha = \angle(U, V) \leq \angle(V, U) < \pi/2$ .  $\square$

## 2.1 Weighted Delaunay triangulation

### 2.1.1 Weighted points

A weighted point is a pair consisting of a point  $p$  of  $\mathbb{R}^d$ , called the *center* of the weighted point, and a non-negative real number  $\omega(p)$ , called the *weight* of the weighted point. It might be convenient to identify a weighted point  $(p, \omega(p))$  and the hypersphere (we will simply say sphere in the sequel) centered at  $p$  of radius  $\omega(p)$ .

Two weighted points (or spheres)  $(p, \omega(p))$  and  $(q, \omega(q))$  are called *orthogonal* when  $\|p - q\|^2 = \omega(p)^2 + \omega(q)^2$ , *further than orthogonal* when  $\|p - q\|^2 > \omega(p)^2 + \omega(q)^2$ , and *closer than orthogonal* when  $\|p - q\|^2 < \omega(p)^2 + \omega(q)^2$ .

Given a point set  $\mathcal{P} = \{p_1, \dots, p_n\} \subseteq \mathbb{R}^d$ , a *weight function* on  $\mathcal{P}$  is a function  $\omega$  that assigns to each point  $p_i \in \mathcal{P}$  a non-negative real weight  $\omega(p_i)$ :  $\omega(\mathcal{P}) = (\omega(p_1), \dots, \omega(p_n))$ . We write  $p_i^\omega = (p_i, \omega(p_i))$  and  $\mathcal{P}^\omega = \{p_1^\omega, \dots, p_n^\omega\}$ .

We define the *relative amplitude* of  $\omega$  as

$$\tilde{\omega} = \max_{p \in \mathcal{P}, q \in \mathcal{P} \setminus \{p\}} \frac{\omega(p)}{\|p - q\|}. \quad (2)$$

In the paper, we will assume that  $\tilde{\omega} \leq \omega_0$ , for some constant  $\omega_0 \in [0, 1/2)$  (see Hypothesis 3.2). Under this hypothesis, all the balls bounded by weighted spheres are disjoint.

Given a subset  $\tau$  of  $d + 1$  weighted points whose centers are affinely independent, there exists a unique sphere orthogonal to the weighted points of  $\tau$ . The sphere is called the *orthosphere* of  $\tau$  and its center  $o_\tau$  and radius  $\Phi_\tau$  are called the *orthocenter* and the *orthoradius* of  $\tau$ . If the weights of the vertices of  $\tau$  are 0 (or all equal), then the orthosphere is simply the *circumscribing sphere* of  $\tau$  whose center and radius are respectively called *circumcenter* and *circumradius*. If  $\tau$  is a  $j$ -simplex,  $j < d$ , the orthosphere of  $\tau$  is the smallest sphere that is orthogonal to the (weighted) vertices of  $\tau$ . Its center  $o_\tau$  lies in  $\text{aff}(\tau)$ . Note that a simplex  $\tau$  may have an orthoradius which is imaginary, i.e.  $\Phi_\tau^2 < 0$ . This situation however cannot happen if the relative amplitude of the weight function is  $< 1/2$ .

A finite set of weighted points  $\mathcal{P}^\omega$  is said to be in *general position* if there exists no sphere orthogonal to  $d + 2$  weighted points of  $\mathcal{P}^\omega$ .

### 2.1.2 Weighted Voronoi diagram and Delaunay triangulation

Let  $\omega$  be a weight function defined on  $\mathcal{P}$ . We define the weighted Voronoi cell of  $p \in \mathcal{P}$  as

$$\text{Vor}^\omega(p) = \{x \in \mathbb{R}^d : \|p - x\|^2 - \omega(p)^2 \leq \|q - x\|^2 - \omega(q)^2, \forall q \in \mathcal{P}\}. \quad (3)$$

The weighted Voronoi cells and their  $k$ -dimensional faces,  $0 \leq k \leq d$ , form a cell complex that decomposes  $\mathbb{R}^d$  into convex polyhedral cells. This cell complex is called the weighted Voronoi diagram or power diagram of  $\mathcal{P}$  [Aur87].

Let  $\tau$  be a subset of points of  $\mathcal{P}$  and write

$$\text{Vor}^\omega(\tau) = \bigcap_{x \in \tau} \text{Vor}^\omega(x).$$

We say  $\mathcal{P}^\omega$  is in *general position*<sup>1</sup> if  $\text{Vor}^\omega(\tau) = \emptyset$  when  $|\tau| > d + 1$ . The collection of all simplices  $\text{conv}(\tau)$  such that  $\text{Vor}^\omega(\tau) \neq \emptyset$  constitutes a triangulation of  $\text{conv}(\mathcal{P})$  if  $\dim \text{aff}(\mathcal{P}) = d$ , and this triangulation is called the weighted Delaunay triangulation  $\text{Del}^\omega(\mathcal{P})$ . The mapping that associates to the face  $\text{Vor}^\omega(\tau)$  of  $\text{Vor}^\omega(\mathcal{P})$  the face  $\text{conv}(\tau)$  of  $\text{Del}^\omega(\mathcal{P})$  is a *duality*, i.e. a bijection that reverses the inclusion relation. See [Del34, Aur87, BDG12].

Alternatively, a  $d$ -simplex  $\tau$  is in  $\text{Del}^\omega(\mathcal{P})$  if the orthosphere  $o_\tau$  of  $\tau$  is further than orthogonal from all weighted points in  $\mathcal{P}^\omega \setminus \tau$ .

Observe that the definition of weighted Voronoi diagrams makes sense if, for some  $p \in \mathcal{P}$ ,  $\omega(p)^2 < 0$ , i.e. some of the weights are imaginary. In fact, since adding a same positive quantity to all  $\omega(p)^2$  does not change the diagram, handling imaginary weights is as easy as handling real weights. In the sequel, we will only consider positive weight functions with relative amplitude  $< 1/2$ .

The weighted Delaunay triangulation of a set of weighted points can be computed efficiently in small dimensions and has found many applications, see e.g. [Aur87]. In this paper, we use weighted Delaunay triangulations for two main reasons. The first one is that the restriction of a  $d$ -dimensional weighted Voronoi diagram to an affine space of dimension  $k$  is a  $k$ -dimensional weighted Voronoi diagram that can be computed without computing the  $d$ -dimensional diagram (see Lemma 2.2). The other main reason is that some flat simplices named slivers can be removed from a Delaunay triangulation by weighting the vertices (see [BGO09, CDE<sup>+</sup>00, CDR05] and Section 3).

**Lemma 2.2** *Let  $H$  be a  $k$ -dimensional affine space of  $\mathbb{R}^d$ . The restriction of the weighted Voronoi diagram of  $\mathcal{P} = \{p_0, \dots, p_m\}$  to  $H$  is identical to the  $k$ -dimensional weighted Voronoi diagram of  $\mathcal{P}' = \{p'_0, \dots, p'_m\}$  in  $H$ , where  $p'_i$  denotes the orthogonal projection of  $p_i$  onto  $H$  and the squared weight of  $p'_i$  is*

$$\omega(p_i)^2 - \|p_i - p'_i\|^2 + \lambda^2$$

where  $\lambda = \max_{p_j \in \mathcal{P}} \|p_j - p'_j\|$  is used to have all weights non-negative. In other words,

$$\text{Vor}^\omega(p_i) \cap H = \text{Vor}^\xi(p'_i)$$

where  $\text{Vor}^\xi(p'_i) = \{x \in H : \|x - p'_i\|^2 - \xi(p'_i)^2 \leq \|x - p'_j\|^2 - \xi(p'_j)^2, \forall p'_j \in \mathcal{P}'\}$ .

**Proof.** By Pythagoras theorem, we have  $\forall x \in H \cap \text{Vor}^\omega(p_i)$ ,  $\|x - p_i\|^2 - \omega(p_i)^2 \leq \|x - p_j\|^2 - \omega(p_j)^2 \Leftrightarrow \|x - p'_i\|^2 + \|p_i - p'_i\|^2 - \omega(p_i)^2 \leq \|x - p'_j\|^2 + \|p_j - p'_j\|^2 - \omega(p_j)^2$ , where  $p'_i$  denotes the orthogonal projection of  $p_i \in \mathcal{P}$  onto  $H$ . See Figure 1. Hence the restriction of  $\text{Vor}^\omega(\mathcal{P})$  to  $H$  is the weighted Voronoi diagram  $\text{Vor}^\xi(\mathcal{P}')$  of the points  $\mathcal{P}'$  with the weight function:  $\xi : \mathcal{P}' \rightarrow [0, \infty)$ , with  $\xi(p'_i)^2 = -\|p_i - p'_i\|^2 + \omega(p_i)^2 + \lambda^2$  where  $\lambda = \max_{p_j \in \mathcal{P}} \|p_j - p'_j\|$ .  $\square$

---

<sup>1</sup>This is an extension of the general position notion introduced by Delaunay [Del34] for Delaunay triangulations. See also [BDG12].



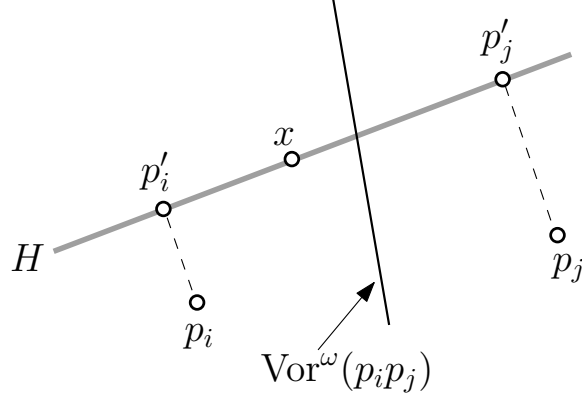


Figure 1: Refer to Lemma 2.2. The grey line denotes the  $k$ -dimensional plane  $H$  and the black line denotes  $\text{Vor}^\omega(p_i p_j)$ .

## 2.2 Sampling conditions

### 2.2.1 Local feature size

The *medial axis* of  $\mathcal{M}$  is the closure of the set of points of  $\mathbb{R}^d$  that have more than one nearest neighbor on  $\mathcal{M}$ . The *local feature size* of  $x \in \mathcal{M}$ ,  $\text{lfs}(x)$ , is the distance of  $x$  to the medial axis of  $\mathcal{M}$  [AB99]. As is well known and can be easily proved,  $\text{lfs}$  is 1-*Lipschitz*, i.e.,

$$\text{lfs}(x) \leq \text{lfs}(y) + \|x - y\|.$$

The infimum of  $\text{lfs}$  over  $\mathcal{M}$  is called the *reach* of  $\mathcal{M}$ . In this paper, we assume that  $\mathcal{M}$  is a compact submanifold of  $\mathbb{R}^d$  of (strictly) positive reach. Note that the class of submanifolds with positive reach includes  $C^2$ -smooth submanifolds [Fed59].

### 2.2.2 Sampling parameters

The point sample  $\mathcal{P}$  is said to be an  $(\varepsilon, \delta)$ -sample (where  $0 < \delta < \varepsilon < 1$ ) if (1) for any point  $x \in \mathcal{M}$ , there exists a point  $p \in \mathcal{P}$  such that  $\|x - p\| \leq \varepsilon \text{lfs}(x)$ , and (2) for any two distinct points  $p, q \in \mathcal{P}$ ,  $\|p - q\| \geq \delta \text{lfs}(p)$ . The parameter  $\varepsilon$  is called the *sampling rate*,  $\delta$  the *sparsity*, and  $\varepsilon/\delta$  the *sampling ratio* of the sample  $\mathcal{P}$ .

The following lemma, proved in [GW04], states basic properties of manifold samples. As before, we write  $\mathbf{m}(p)$  for the distance between  $p \in \mathcal{P}$  and its nearest neighbor in  $\mathcal{P} \setminus \{p\}$ .

**Lemma 2.3** *Given an  $(\varepsilon, \delta)$ -sample  $\mathcal{P}$  of  $\mathcal{M}$ , we have*

1. *For all  $p \in \mathcal{P}$ , we have*

$$\delta \text{lfs}(p) \leq \mathbf{m}(p) \leq \frac{2\varepsilon}{1 - \varepsilon} \text{lfs}(p).$$

2. *For any two points  $p, q \in \mathcal{M}$  such that  $\|p - q\| = t \text{lfs}(p)$ ,  $0 < t < 1$ ,*

$$\sin \angle(pq, T_p) \leq t/2.$$

3. Let  $p$  be a point in  $\mathcal{M}$ . Let  $x$  be a point in  $T_p$  such that  $\|p - x\| \leq t \text{lfs}(p)$  for some  $0 < t \leq 1/4$ . Let  $x'$  be the point on  $\mathcal{M}$  closest to  $x$ . Then

$$\|x - x'\| \leq 2t^2 \text{lfs}(p).$$

### 2.3 Fat simplices

Consider a  $j$ -simplex  $\tau$ , where  $1 \leq j \leq k + 1$ . We denote by  $R_\tau, \Delta_\tau, L_\tau, V_\tau$  and  $\Gamma_\tau = \Delta_\tau / L_\tau$  the circumradius, the longest edge length, the shortest edge length, the  $j$ -dimensional volume, and *spread* i.e. the longest edge to shortest edge ratio of  $\tau$  respectively.

We define the *fatness* of a  $j$ -dimensional simplex  $\tau$  as

$$\Theta_\tau = \begin{cases} 1 & \text{if } j = 0 \\ V_\tau^{1/j} / \Delta_\tau & \text{if } j > 0 \end{cases} \quad (4)$$

The following important lemma is due to Whitney [Whi57, Chap. II].

**Lemma 2.4** *Let  $\tau = [p_0, \dots, p_j]$  be a  $j$ -dimensional simplex and let  $H$  be an affine flat such that  $\tau$  is contained in the offset of  $H$  by  $\eta$  (i.e. any point of  $\tau$  is at distance at most  $\eta$  from  $H$ ). If  $u$  is a unit vector in  $\text{aff}(\tau)$ , then there exists a unit vector  $u_H$  in  $H$  such that*

$$\sin \angle(u, u_H) \leq \frac{2\eta}{(j-1)! \Theta_\tau^j L_\tau}.$$

We deduce from the above lemma the following corollary. See also Lemma 1 in [Fu93] and Lemma 16 in [CDR05].

**Corollary 2.5 (Tangent space approximation)** *Let  $\tau$  be a  $j$ -simplex,  $j \leq k$ , with vertices on  $\mathcal{M}$ , and let  $p$  be vertex of  $\tau$ . Assuming that  $\Delta_\tau < \text{lfs}(p)$ , we have*

$$\sin \angle(\text{aff}(\tau), T_p) \leq \frac{\Delta_\tau^2}{\Theta_\tau^k L_\tau \text{lfs}(p)} = \frac{\Gamma_\tau \Delta_\tau}{\Theta_\tau^k \text{lfs}(p)}.$$

**Proof.** It suffices to take  $H = T_p$  and to use  $\eta = \Delta_\tau^2 / 2 \text{lfs}(p)$  (from Lemma 2.3 (2)) and  $\Gamma_\tau = \Delta_\tau / L_\tau$ . Hence

$$\sin \angle(\text{aff}(\tau), T_p) \leq \frac{2\eta}{(j-1)! \Theta_\tau^j L_\tau} \leq \frac{2\eta}{\Theta_\tau^j L_\tau} \leq \frac{\Delta_\tau^2}{\Theta_\tau^j L_\tau \text{lfs}(p)} = \frac{\Gamma_\tau \Delta_\tau}{\Theta_\tau^j \text{lfs}(p)}$$

□

A sliver is a special type of flat simplex. The property of being a sliver is defined in terms of a parameter  $\Theta_0$ , to be fixed later in Section 4.

The following definition is a variant of a definition given in [Li03].

**Definition 2.6 ( $\Theta_0$ -fat simplices and  $\Theta_0$ -slivers)** *Given a positive parameter  $\Theta_0$ , a simplex  $\tau$  is said to be  $\Theta_0$ -fat if the fatness of  $\tau$  and of all its subsimplices is at least  $\Theta_0$ .*

*A simplex of dimension at least 2 which is not  $\Theta_0$ -fat but whose subsimplices are all  $\Theta_0$ -fat is called a  $\Theta_0$ -sliver.*

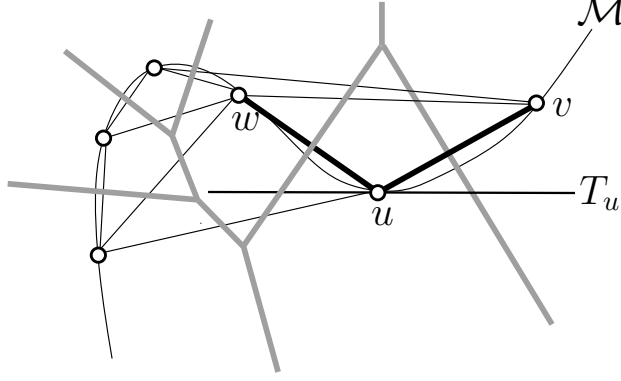


Figure 2:  $\mathcal{M}$  is the black curve. The sample  $\mathcal{P}$  is the set of small circles. The tangent space at  $u$  is denoted by  $T_u$ . The Voronoi diagram of the sample is in grey. The edges of the Delaunay triangulation  $\text{Del}(\mathcal{P})$  are the line segments between small circles. In bold,  $\text{st}(u) = \{uv, uw\}$ .

## 2.4 Tangential Delaunay complex and inconsistent configurations

Let  $\mathcal{P}$  be a point sample of  $\mathcal{M}$  and  $u$  be a point of  $\mathcal{P}$ . We denote by  $\text{Del}_u^\omega(\mathcal{P})$  the weighted Delaunay triangulation of  $\mathcal{P}$  restricted to the tangent space  $T_u$ . Equivalently, the simplices of  $\text{Del}_u^\omega(\mathcal{P})$  are the simplices of  $\text{Del}^\omega(\mathcal{P})$  whose Voronoi dual faces intersect  $T_u$ , i.e.,  $\tau \in \text{Del}_u^\omega(\mathcal{P})$  iff  $\text{Vor}^\omega(\tau) \cap T_u \neq \emptyset$ . Observe that, from Lemma 2.2,  $\text{Del}_u^\omega(\mathcal{P})$  is in general isomorphic to a weighted Delaunay triangulation of a  $k$ -dimensional convex hull. Since this situation can always be ensured by applying some infinitesimal perturbation on  $\mathcal{P}$  or of the weights, in addition to the general position assumption of  $\mathcal{P}^\omega$  we will also assume, in the rest of the paper, that all  $\text{Del}_u^\omega(\mathcal{P})$  are isomorphic to a weighted Delaunay triangulation of a  $k$ -dimensional convex hull. We will discuss this in more details at the beginning of Section 5. Finally, write  $\text{st}(u)$  for the *star* of  $u$  in  $\text{Del}_u^\omega(\mathcal{P})$ , i.e.,  $\text{st}(u) = \text{st}(u, \text{Del}_u^\omega(\mathcal{P}))$  (see Figure 2).

We denote by *tangential Delaunay complex* or *tangential complex* for short, the simplicial complex  $\{\tau : \tau \in \text{st}(u), u \in \mathcal{P}\}$ . We denote it by  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$ . By our assumption above,  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  is a  $k$ -dimensional subcomplex of  $\text{Del}^\omega(\mathcal{P})$ . Note that  $\text{st}(u) = \text{st}(u, \text{Del}_u^\omega(\mathcal{P}))$  and  $\text{st}(u, \text{Del}_{T\mathcal{M}}^\omega(\mathcal{P}))$ , the star of  $u$  in the complex  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$ , are in general different.

By duality, computing  $\text{st}(u)$  is equivalent to computing the restriction to  $T_u$  of the (weighted) Voronoi cell of  $u$ , which, by Lemma 2.2, reduces to computing a cell in a  $k$ -dimensional weighted Voronoi diagram embedded in  $T_u$ . To compute such a cell, we need to compute the intersection of  $|\mathcal{P}| - 1$  halfspaces of  $T_u$  where  $|\mathcal{P}|$  is the cardinality of  $\mathcal{P}$ . Each halfspace is bounded by the bisector consisting of the points of  $T_u$  that are at equal weighted distance from  $u^\omega$  and some other point in  $\mathcal{P}^\omega$ . This can be done in optimal time [Cha93, CS89]. It follows that the tangential complex can be computed without constructing any data structure of dimension higher than  $k$ , the intrinsic dimension of  $\mathcal{M}$ .

The tangential Delaunay complex is *not* in general a triangulated manifold and therefore not a good approximation of  $\mathcal{M}$ . This is due to the presence of so-called inconsistencies. Consider a  $k$ -simplex  $\tau$  of  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  with two vertices  $u$  and  $v$  such that  $\tau$  is in  $\text{st}(u)$  but not in  $\text{st}(v)$  (refer to Figure 3). We write  $B_u(\tau)$  (and  $B_v(\tau)$ ) for the open ball centered on  $T_u$  (and  $T_v$ ) that is orthogonal to the (weighted) vertices of  $\tau$ , and denote by  $m_u(\tau)$  (and  $m_v(\tau)$ ), or  $m_u$  ( $m_v$ ) for short, its center.



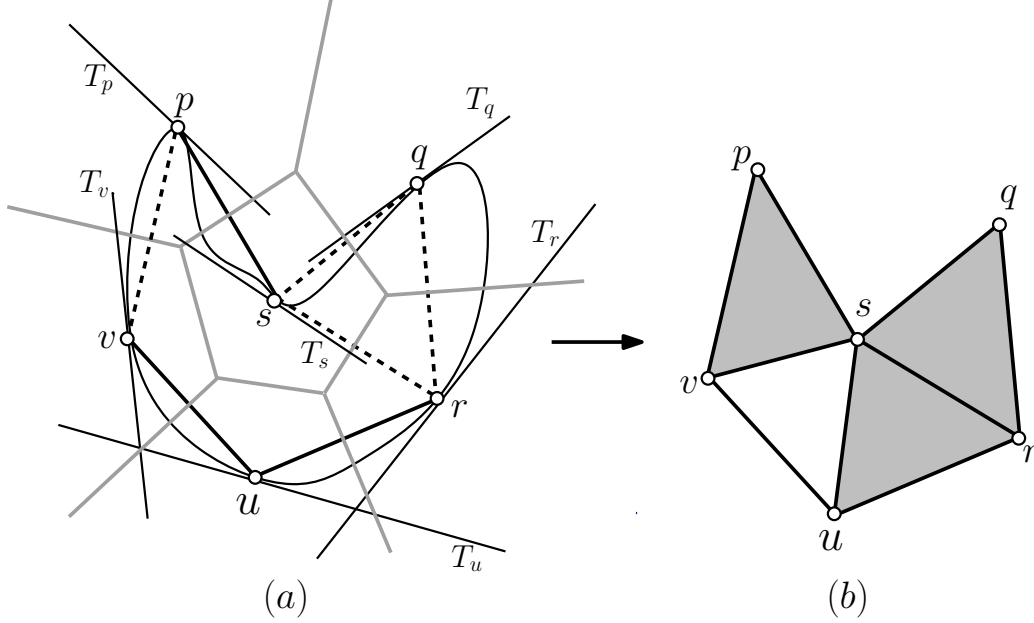


Figure 4: In Figure (a),  $\mathcal{M}$  is the black curve, the sample  $\mathcal{P}$  is the set of small circles, the tangent space at a point  $x \in \mathcal{P}$  is denoted by  $T_x$  and the Voronoi diagram of the sample is in grey and  $\text{Del}_{T,\mathcal{M}}(\mathcal{P})$  is the line segments between the sample points. In dashed lines, are the inconsistent simplices in  $\text{Del}_{T,\mathcal{M}}(\mathcal{P})$ . In Figure (b), the grey triangles denote the inconsistent configurations corresponding to the inconsistent simplices in Figure (a).

An inconsistent configuration is therefore a  $(k+1)$ -simplex of  $\text{Del}^\omega(\mathcal{P})$ . However, an inconsistent configuration does not belong to  $\text{Del}_{T,\mathcal{M}}^\omega(\mathcal{P})$  since  $\text{Del}_{T,\mathcal{M}}^\omega(\mathcal{P})$  has no  $(k+1)$ -simplex under our general position assumption. Moreover, the lower dimensional faces of an inconsistent configuration do not necessarily belong to  $\text{Del}_{T,\mathcal{M}}^\omega(\mathcal{P})$ .

Since the inconsistent configurations are  $k+1$ -dimensional simplices, we will use the same notations for inconsistent configurations as for simplices, e.g.  $R_\phi$  and  $c_\phi$  for the circumradius and the circumcenter of  $\phi$ ,  $\rho_\phi$  and  $\Theta_\phi$  for its radius-edge ratio and fatness respectively.

We write  $\text{Inc}^\omega(\mathcal{P})$  for the subcomplex of  $\text{Del}^\omega(\mathcal{P})$  consisting of all the  $\Theta_0$ -inconsistent configurations of  $\text{Del}_{T,\mathcal{M}}^\omega(\mathcal{P})$  and their subfaces. We also define the *completed complex*  $C^\omega(\mathcal{P})$  as

$$C^\omega(\mathcal{P}) = \text{Del}_{T,\mathcal{M}}^\omega(\mathcal{P}) \cup \text{Inc}^\omega(\mathcal{P}).$$

Refer to Figure 4.

An important observation, stated as Lemma 4.9 in Section 4.3, is that, if  $\varepsilon$  is sufficiently small with respect to  $\Theta_0$ , then the fatness of  $\phi$  is less than  $\Theta_0$ . Hence, if the subfaces of  $\phi$  are  $\Theta_0$ -fat simplices,  $\phi$  will be a  $\Theta_0$ -sliver. This observation is at the core of our reconstruction algorithm.

### 3 Manifold reconstruction

The algorithm removes all  $\Theta_0$ -slivers from  $C^\omega(\mathcal{P})$  by weighting the points of  $\mathcal{P}$ . By the observation mentioned above, all inconsistencies in the tangential complex will then also be removed. All

simplices being consistent, the resulting weighted tangential Delaunay complex  $\hat{\mathcal{M}}$  output by the algorithm will be a simplicial  $k$ -manifold that approximates  $\mathcal{M}$  well, as will be shown in Section 5.

In this section, we describe the algorithm. Its analysis is deferred to Section 4.

We will make the two following hypotheses.

**Hypothesis 3.1**  $\mathcal{M}$  is a compact smooth submanifold of  $\mathbb{R}^d$  without boundary, and  $\mathcal{P}$  is an  $(\varepsilon, \delta)$ -sample of  $\mathcal{M}$  of sampling ratio  $\varepsilon/\delta \leq \eta_0$  for some positive constant  $\eta_0$ .

As shown in [CWW08, GW04], we can estimate the tangent space  $T_p$  at each sample point  $p$  and also the dimension  $k$  of the manifold from  $\mathcal{P}$  and  $\eta_0$ . We assume now that  $T_p$ , for any point  $p \in \mathcal{P}$ ,  $k$  and  $\eta_0$  are known.

The algorithm fixes a bound  $\omega_0$  on the relative amplitude of the weight assignment :

**Hypothesis 3.2**  $\tilde{\omega} \leq \omega_0$ , for some constant  $\omega_0 \in [0, 1/2)$

Observe that, under this hypothesis, all the balls bounded by weighted spheres are disjoint.

The algorithm also fixes  $\Theta_0$  to a constant defined in Theorem 4.15, that depends on  $k$ ,  $\omega_0$  and  $\eta_0$ .

We define the *local neighborhood* of  $p \in \mathcal{P}$  as

$$LN(p) = \{q \in \mathcal{P} : |B(p, \|p - q\|) \cap \mathcal{P}| \leq N\}. \quad (5)$$

where  $N$  is a constant that depends on  $k$ ,  $\omega_0$  and  $\eta_0$  to be defined in Section 4.4. We will show in Lemma 4.12, that  $LN(p)$  includes all the points of  $\mathcal{P}$  that can form an edge with  $p$  in  $C^\omega(\mathcal{P})$ . In fact, the algorithm can use instead of  $LN(p)$  any subset of  $\mathcal{P}$  that contains  $LN(p)$ . This will only affect the complexity of the algorithm, not the output.

**Outline of the algorithm.** Initially, all the sample points in  $\mathcal{P}$  are assigned zero weights, and the completed complex  $C^\omega(\mathcal{P})$  is built for this zero weight assignment. Then the algorithm processes each point  $p_i \in \mathcal{P} = \{p_1, \dots, p_n\}$  in turn, and assigns a new weight to  $p_i$ . The new weight is chosen so that all the simplices of all dimensions in  $C^\omega(\mathcal{P})$  are  $\Theta_0$ -fat. See Algorithm 1.

We now give the details of the functions used in the manifold reconstruction algorithm. The function **update\_completed\_complex**( $Q, \omega$ ) is described as Algorithm 2. It makes use of two functions, **build\_star**( $p$ ) and **build\_inconsistent\_configurations**( $p, \tau$ ).

The function **build\_star**( $p$ ) calculates the weighted Voronoi cell of  $p$ , which reduces to computing the intersection of the halfspaces of  $T_p$  bounded by the (weighted) bisectors between  $p$  and other points in  $LN(p)$ .

The function **build\_inconsistent\_configurations**( $u, \tau$ ) adds to  $C^\omega(\mathcal{P})$  all the inconsistent configurations of the form  $\phi = \tau \cup \{w\}$  where  $\tau$  is an inconsistent simplex of  $\text{st}(u)$ . More precisely, for each vertex  $v \neq u$  of  $\tau$  such that  $\tau \not\subset \text{st}(v)$ , we calculate the points  $w \in LN(p)$ , such that  $(u, v, w)$  witnesses the inconsistent configuration  $\phi = \tau \cup \{w\}$ . Specifically, we compute the restriction of the Voronoi diagram of the points in  $LN(u)$  to the line segment  $[m_u m_v]$ , where  $m_u = T_u \cap \text{aff}(\text{Vor}^\omega(\tau))$  and  $m_v = T_v \cap \text{aff}(\text{Vor}^\omega(\tau))$ . According to the definition of an inconsistent configuration,  $w$  is one of the sites whose (restricted) Voronoi cell is the first to be intersected by the line segment  $[m_u m_v]$ , oriented from  $m_u$  to  $m_v$ . We add inconsistent configuration  $\phi = \tau \cup \{w\}$  to the completed complex.

We now give the details of function **weight**( $p, \omega$ ) that computes  $\omega(p)$ , keeping the other weights fixed (see Algorithm 3). This function extends a similar subroutine introduced in [CDE<sup>+</sup>00] for

---

**Algorithm 1** **Manifold\_reconstruction**( $\mathcal{P} = \{p_0, \dots, p_n\}, \eta_0$ )

---

```
// Initialization
for  $i = 1$  to  $n$  do
  calculate the local neighborhood  $LN(p_i)$ 
end for
for  $i = 1$  to  $n$  do
   $\omega(p_i) \leftarrow 0$ 
end for
// Build the full unweighted completed complex  $C^\omega(\mathcal{P})$ 
 $C^\omega(\mathcal{P}) \leftarrow \text{update\_completed\_complex}(\mathcal{P}, \omega)$ 
// Weight assignment to remove inconsistencies
for  $i = 1$  to  $n$  do
   $\omega(p_i) \leftarrow \text{weight}(p_i, \omega)$ 
   $\text{update\_completed\_complex}(LN(p_i), \omega)$ 
end for
// Output
output :  $\hat{\mathcal{M}} \leftarrow \text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$ 
```

---

---

**Algorithm 2** Function **update\_completed\_complex**( $Q, \omega$ )

---

```
for each point  $q \in Q$  do
  build_star( $q$ )
end for
for each  $q \in Q$  do
  for each  $k$ -simplex  $\tau$  in  $\text{st}(q)$  do
    if  $\tau$  is  $\Theta_0$ -fat and  $\exists v \in \tau, \tau \not\subset \text{st}(v)$  then
      //  $\tau$  is inconsistent
      build_inconsistent_configurations( $q, \tau$ )
    end if
  end for
end for
```

---

removing slivers in  $\mathbb{R}^3$ . We need first to define candidate simplices. A *candidate simplex* of  $p$  is defined as a simplex of  $C^\omega(\mathcal{P})$  that becomes incident to  $p$  when the weight of  $p$  is varied from 0 to  $\omega_0 \mathfrak{m}(p)$ , keeping the weights of all the points in  $\mathcal{P} \setminus \{p\}$  fixed. Note that a candidate simplex of  $p$  is incident to  $p$  for some weight  $\omega(p)$  but does not necessarily belong to  $\text{st}(p)$ .

Let  $\tau$  be a candidate simplex of  $p$  that is a  $\Theta_0$ -sliver. We associate to  $\tau$  a *forbidden interval*  $W(p, \tau)$  that consists of all squared weights  $\omega(p)^2$  for which  $\tau$  appears as a simplex in  $C^\omega(\mathcal{P})$  (the weights of the other points remaining fixed).

---

**Algorithm 3** Function **weight**( $p, \omega$ )

---

```

 $S(p) \leftarrow \text{candidate\_slivers}(p, \omega)$ 
//  $J(p)$  is the set of squared weights of  $p$  such that  $C^\omega(\mathcal{P})$  contains
// no  $\Theta_0$ -sliver incident to  $p$ 
 $J(p) \leftarrow [0, \omega_0^2 \mathfrak{m}(p)^2] \setminus \bigcup_{\tau \in S(p)} W(p, \tau)$ 
 $\omega(p)^2 \leftarrow$  a squared weight from  $J(p)$ 
return  $\omega(p)$ 

```

---

The function **candidate\_slivers**( $p, \omega$ ) varies the weight of  $p$  and computes all the candidate slivers of  $p$  and their corresponding weight intervals  $W(p, \tau)$ . More precisely, this function follows the following steps.

1. We first detect all candidate  $j$ -simplices for all  $2 \leq j \leq k+1$ . This is done in the following way. We vary the weight of  $p$  from 0 to  $\omega_0 \mathfrak{m}(p)$ , keeping the weights of the other points fixed. For each new weight assignment to  $p$ , we modify the stars and inconsistent configurations of the points in  $LN(p)$  and detect the new  $j$ -simplices incident to  $p$  that have not been detected so far. The weight of point  $p$  changes only in a finite number of instances  $0 = P_0 < P_1 < \dots < P_{n-1} < P_n = \omega_0 \mathfrak{m}(p)$ .

2. We determine the next weight assignment of  $p$  in the following way. For each new simplex  $\tau$  currently incident to  $p$ , we keep it in a priority queue ordered by the weight of  $p$  at which  $\tau$  will disappear for the first time. Hence the minimum weight in the priority queue gives the next weight assignment for  $p$ . Since the number of points in  $LN(p)$  is bounded, the number of simplices incident to  $p$  is also bounded, as well as the number of times we have to change the weight of  $p$ .

3. For each candidate sliver  $\tau$  of  $p$  which is detected, we compute  $W(p, \tau)$  on the fly. The candidate slivers are stored in a list with their corresponding forbidden intervals.

## 4 Analysis of the algorithm

The analysis of the algorithm relies on structural results that will be proved in Sections 4.2, 4.3 and 4.4. We will then prove that the algorithm is correct and analyze its complexity in Sections 4.5 and 4.6. In Section 5, we will show that the output  $\hat{\mathcal{M}}$  of the reconstruction algorithm is a good approximation of  $\mathcal{M}$ .

The bounds to be given in the lemmas of this section will depend on the dimension  $k$  of  $\mathcal{M}$ , the bound  $\eta_0$  on the sampling ratio, and on a positive scalar  $\Theta_0$  that bounds the fatness and will be used to define slivers, fat simplices and inconsistent configurations.



## 4.1 First lemmas

### Circumradius and orthoradius

The following lemma states some basic facts about weighted Voronoi diagrams when the relative amplitude of the weighting function is bounded. Similar results were proved in [CDE<sup>+</sup>00].

**Lemma 4.1** *Assume that Hypothesis 3.2 is satisfied. If  $\tau$  is a simplex of  $\text{Del}^\omega(\mathcal{P})$  and  $p$  and  $q$  are any two vertices of  $\tau$ , then*

1.  $\forall z \in \text{aff}(\text{Vor}^\omega(\tau)), \|q - z\| \leq \frac{\|p - z\|}{\sqrt{1 - 4\omega_0^2}}.$
2.  $\forall z \in \text{aff}(\text{Vor}^\omega(\tau)), \sqrt{\|z - p\|^2 - \omega^2(p)} \geq \Phi_\tau.$
3.  $\forall \sigma \subseteq \tau, \Phi_\sigma \leq \Phi_\tau.$

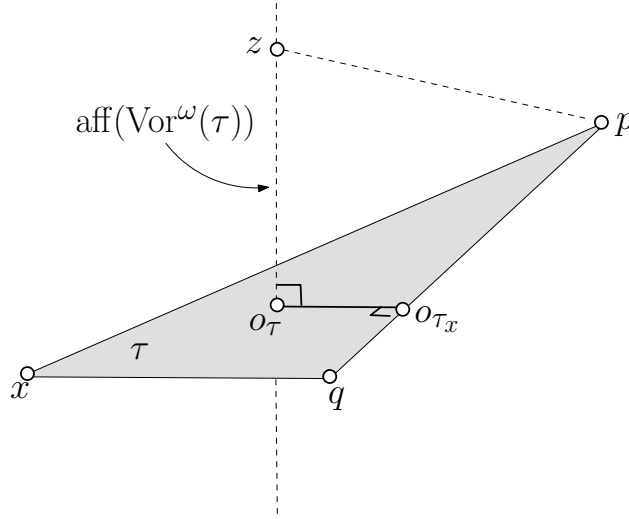


Figure 5: For the proofs of Lemma 4.1.

**Proof.** Refer to Figure 5.

1. If  $\|z - q\| \leq \|z - p\|$ , then the lemma is proved since  $0 < \sqrt{1 - 4\omega_0^2} \leq 1$ . Hence assume that  $\|z - q\| > \|z - p\|$ . Since  $z \in \text{aff}(\text{Vor}^\omega(\tau))$

$$\begin{aligned}
 \|z - p\|^2 &= \|z - q\|^2 + \omega(p)^2 - \omega(q)^2 \\
 &\geq \|z - q\|^2 - \omega(q)^2 \\
 &\geq \|z - q\|^2 - \omega_0^2 \|p - q\|^2 \\
 &\geq \|z - q\|^2 - \omega_0^2 (\|z - p\| + \|z - q\|)^2 \\
 &> \|z - q\|^2 - 4\omega_0^2 \|z - q\|^2 = (1 - 4\omega_0^2) \|z - q\|^2.
 \end{aligned}$$

2. We know that  $o_\tau = \text{aff}(\text{Vor}^\omega(\tau)) \cap \text{aff}(\tau)$ . Therefore, using Pythagoras theorem,

$$\begin{aligned}
 \|z - p\|^2 - \omega(p)^2 &= \|p - o_\tau\|^2 + \|o_\tau - z\|^2 - \omega(p)^2 \\
 &= \Phi_\tau^2 + \|o_\tau - z\|^2 \geq \Phi_\tau^2.
 \end{aligned}$$

3. The result directly follows from part 2 and the fact that  $\text{aff}(\text{Vor}^\omega(\tau)) \subseteq \text{aff}(\text{Vor}^\omega(\sigma))$  (since  $\sigma \subseteq \tau$ ).  $\square$

### Altitude and fatness

If  $p \in \tau$ , we define  $\tau_p = \tau \setminus \{p\}$  to be the  $(j-1)$ -face of  $\tau$  opposite to  $p$ . We also write  $D_\tau(p)$  for the distance from  $p$  to the affine hull of  $\tau_p$ .  $D_\tau(p)$  will be called the *altitude* of  $p$  in  $\tau$ .

From the definition of fatness, we easily derive the following lemma.

**Lemma 4.2** *Let  $\tau = [p_0, \dots, p_j]$  be a  $j$ -dimensional simplex and  $p$  be a vertex of  $\tau$ .*

1.  $\Theta_\tau^j \leq \frac{1}{j!}$
2.  $j! \Theta_\tau^j \leq \frac{D_\tau(p)}{\Delta_\tau} \leq j \Gamma_\tau^{j-1} \times \frac{\Theta_\tau^j}{\Theta_{\tau_p}^{j-1}}$

**Proof.** 1. Without loss of generality we assume that  $\tau = [p_0, \dots, p_j]$  is embedded in  $\mathbb{R}^j$ . From the definition of fatness, we have

$$\Delta_\tau^j \Theta_\tau^j = V_\tau = \frac{|\det(p_1 - p_0 \dots p_j - p_0)|}{j!} \leq \frac{\Delta_\tau^j}{j!}.$$

2. Using the above inequality and the definition of fatness, we get

$$D_\tau(p) = \frac{j V_\tau}{V_{\tau_p}} \geq \frac{j \Theta_\tau^j \Delta_\tau^j}{\frac{\Delta_\tau^{j-1}}{(j-1)!}} \geq j! \Theta_\tau^j \Delta_\tau.$$

Moreover, using  $\Delta_{\tau_p} \geq L_{\tau_p} \geq L_\tau$  and  $\Gamma_\tau = \frac{\Delta_\tau}{L_\tau}$ ,

$$\frac{D_\tau(p)}{\Delta_\tau} = \frac{j V_\tau}{\Delta_\tau V_{\tau_p}} = j \frac{\Theta_\tau^j \Delta_\tau^{j-1}}{\Theta_{\tau_p}^{j-1} \Delta_{\tau_p}^{j-1}} \leq j \frac{\Theta_\tau^j \Delta_\tau^{j-1}}{\Theta_{\tau_p}^{j-1} L_\tau^{j-1}} = j \Gamma_\tau^{j-1} \times \frac{\Theta_\tau^j}{\Theta_{\tau_p}^{j-1}}.$$

□

### Excentricity

Let  $\tau$  be a simplex and  $p$  be a vertex of  $\tau$ . We define the *excentricity*  $H_\tau(p, \omega(p))$  of  $\tau$  with respect to  $p$  as the signed distance from  $o_\tau$  to  $\text{aff}(\tau_p)$ . Hence,  $H_\tau(p, \omega(p))$  is positive if  $o_\tau$  and  $p$  lie on the same side of  $\text{aff}(\tau_p)$  and negative if they lie on different sides of  $\text{aff}(\tau_p)$ .

The following lemma is a generalization of Claim 13 from [CDE<sup>+</sup>00]. It bounds the excentricity of a simplex  $\tau$  as a function of  $\omega(p)$  where  $p$  is a vertex of  $\tau$ . The proof is included for completeness even though it is exactly the same as the one given in [CDE<sup>+</sup>00].

**Lemma 4.3** *Let  $\tau$  be a simplex of  $\text{Del}^\omega(\mathcal{P})$  and let  $p$  be any vertex of  $\tau$ . We have*

$$H_\tau(p, \omega(p)) = H_\tau(p, 0) - \frac{\omega(p)^2}{2 D_\tau(p)}.$$

**Proof.** Refer to Figure 6. For convenience, we write  $R = \Phi_\tau$  for the orthoradius of  $\tau$  and  $r = \Phi_{\tau_p}$  for the orthoradius of  $\tau_p$ . The orthocenter  $o_{\tau_p}$  of  $\tau_p$  is the projection of  $o_\tau$  onto  $\tau_p$ . When the weight  $\omega(p)$  varies while the weights of other points remain fixed,  $o_\tau$  moves on a (fixed) line  $L$  that passes through  $o_{\tau_p}$ . Now, let  $p'$  and  $p''$  be the projections of  $p$  onto  $L$  and  $\text{aff}(\tau_p)$  respectively.



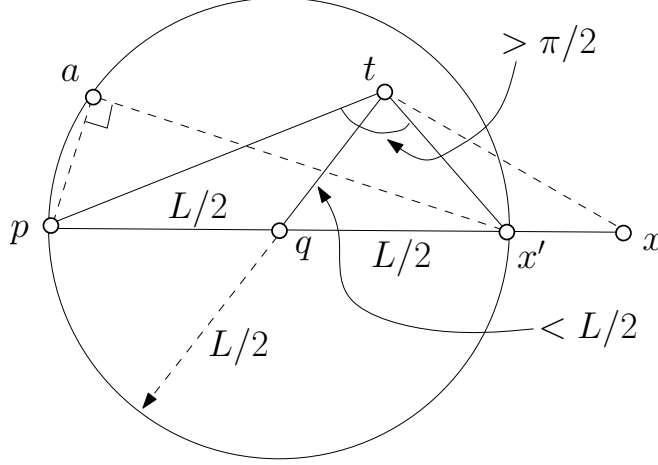


Figure 7: Refer to Lemma 4.4.  $x'$  is a point on the line segment such that  $\|p - x'\| = C_1 \varepsilon \text{lfs}(p)$ ,  $L = C_1 \varepsilon \text{lfs}(p)$ ,  $\angle pax' = \pi/2$  and  $\angle ptx \geq \angle ptx' > \pi/2$ .

Hence,

$$\|p - q'\| \leq \|p - q\| + \|q - q'\| < \frac{C_1}{2} \varepsilon (1 + C_1 \varepsilon) \text{lfs}(p)$$

Since  $\mathcal{P}$  is an  $\varepsilon$ -sample, there exists a point  $t \in \mathcal{P}$  s.t.  $\|q' - t\| \leq \varepsilon \text{lfs}(q')$ . Using the fact that  $\text{lfs}$  is 1-Lipschitz and Eq. (6),

$$\text{lfs}(q') \leq \text{lfs}(p) + \|p - q'\| < (1 + \frac{C_1}{2} \varepsilon (1 + C_1 \varepsilon)) \text{lfs}(p) < \frac{C_1}{2} (1 - C_1 \varepsilon) \text{lfs}(p),$$

which yields  $\|q' - t\| < \frac{C_1}{2} \varepsilon (1 - C_1 \varepsilon) \text{lfs}(p)$ . We thus have

$$\|q - t\| \leq \|q - q'\| + \|q' - t\| < \frac{C_1}{2} \varepsilon \text{lfs}(p).$$

It follows (see Figure 7), that  $\angle ptx > \pi/2$ , which implies that

$$\|x - p\|^2 - \|x - t\|^2 - \|p - t\|^2 > 0.$$

Hence,

$$\begin{aligned} \|x - p\|^2 - \|x - t\|^2 - \omega^2(p) + \omega^2(t) &\geq \|p - t\|^2 - \omega^2(p) \\ &\geq \|p - t\|^2 - \omega_0^2 \|p - t\|^2 \\ &> 0 \quad (\text{since } \omega_0 < \tfrac{1}{2}) \end{aligned}$$

This implies  $x \notin \text{Vor}^\omega(p)$ , which contradicts our initial assumption. We conclude that  $\text{Vor}^\omega(p) \cap T_p \subseteq B(p, C_1 \varepsilon \text{lfs}(p))$  if Eq. (6) is satisfied, which is true for  $C_1 \stackrel{\text{def}}{=} 3 + \sqrt{2} \approx 4.41$  and  $\varepsilon \leq \varepsilon_0 < 0.09$ .  $\square$

The following lemma states that, under Hypotheses 3.2 and 3.1, the simplices of  $\text{Del}_p^\omega(\mathcal{P})$  are small, have a good radius-edge ratio and a small excentricity.

**Lemma 4.5** Assume that Hypotheses 3.1 and 3.2 are satisfied. There exists positive constants  $C_2$ ,  $C_3$  and  $C_4$  that depend on  $\omega_0$  and  $\eta_0$  such that, if  $\varepsilon < \frac{1}{2C_2}$ , the following holds.

1. If  $pq$  is an edge of  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$ , then  $\|p - q\| < C_2\varepsilon\text{lfs}(p)$ .
2. If  $\tau$  is a simplex of  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$ , then  $\Phi_\tau \leq C_3L_\tau$  and  $\Gamma_\tau = \Delta_\tau/L_\tau \leq C_3$ .
3. If  $\tau$  is a simplex of  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  and  $p$  a vertex of  $\tau$ , the eccentricity  $|H_\tau(p, \omega(p))|$  is at most  $C_4\varepsilon\text{lfs}(p)$ .

**Proof.** 1. We will consider the following two cases:

- (a) Consider first the case where  $pq$  is an edge of  $\text{Del}_p^\omega(\mathcal{P})$ . Then  $T_p \cap \text{Vor}^\omega(pq) \neq \emptyset$ . Let  $x \in T_p \cap \text{Vor}^\omega(pq)$ . From Lemma 4.4, we have  $\|p - x\| \leq C_1\varepsilon\text{lfs}(p)$ . By Lemma 4.1,

$$\|q - x\| \leq \frac{\|p - x\|}{\sqrt{1 - 4\omega_0^2}} \leq \frac{C_1\varepsilon\text{lfs}(p)}{\sqrt{1 - 4\omega_0^2}}.$$

Hence,  $\|p - q\| \leq C'_1\varepsilon\text{lfs}(p)$  where  $C'_1 \stackrel{\text{def}}{=} C_1(1 + 1/\sqrt{1 - 4\omega_0^2})$ .

- (b) From the definition of  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$ , there exists a vertex  $r$  of  $\tau$  such that  $[pq] \in \text{st}(r)$ . From (a),  $\|r - p\|$  and  $\|r - q\|$  are at most  $C'_1\varepsilon\text{lfs}(r)$ . Using the fact that  $\text{lfs}$  is 1-Lipschitz,  $\text{lfs}(p) \geq \text{lfs}(r) - \|p - r\| \geq (1 - C'_1\varepsilon)\text{lfs}(r)$  (from part 1 (a)), which yields  $\text{lfs}(r) \leq \frac{\text{lfs}(p)}{1 - C'_1\varepsilon}$ . It follows that

$$\|p - q\| \leq \|p - r\| + \|r - q\| \leq \frac{2C'_1\varepsilon\text{lfs}(p)}{1 - C'_1\varepsilon}.$$

The first part of the lemma is proved by taking  $C_2 \stackrel{\text{def}}{=} \frac{5C'_1}{2}$  and using  $2C_2\varepsilon < 1$ .

2. Without loss of generality, let  $\tau_1 \in \text{st}(p)$  be a simplex incident to  $p$  and  $\tau \subseteq \tau_1$ . Let  $z \in \text{Vor}^\omega(\tau_1) \cap T_p$ , and  $r_z = \sqrt{\|z - p\|^2 - \omega^2(p)}$ . The ball centered at  $z$  with radius  $r_z$  is orthogonal to the weighted vertices of  $\tau_1$ . From Lemma 4.1 (2) and (3), we have  $r_z \geq \Phi_{\tau_1}$  and  $\Phi_\tau \leq \Phi_{\tau_1}$ . Hence it suffices to prove that there exists a constant  $C_3$  such that  $r_z \leq C_3L_\tau$ . Since  $z \in \text{Vor}^\omega(\tau) \cap T_p$ , we deduce from Lemma 4.4 that  $\|z - p\| \leq C_1\varepsilon\text{lfs}(p)$ . Therefore

$$r_z = \sqrt{\|z - p\|^2 - \omega^2(p)} \leq \|z - p\| \leq C_1\varepsilon\text{lfs}(p).$$

For any vertex  $q$  of  $\tau$ , we have  $\|p - q\| \leq C_2\varepsilon\text{lfs}(p)$  (By part 1). Using  $2C_2\varepsilon < 1$  and the fact that  $\text{lfs}$  is 1-Lipschitz,  $\text{lfs}(p) \leq 2\text{lfs}(q)$ . Therefore, taking for  $q$  a vertex of the shortest edge of  $\tau$ , we have, using Lemma 2.3 and Hypothesis 1,

$$r_z \leq C_1\varepsilon\text{lfs}(p) \leq C_1 \left( \frac{\varepsilon}{\delta} \times \delta \right) \times 2\text{lfs}(q) \leq 2C_1\eta_0 L_\tau.$$

From part (1) of the lemma we have  $\Delta_{\tau_1} \leq 2C_2\varepsilon\text{lfs}(p)$ . Therefore

$$\Gamma_\tau = \frac{\Delta_\tau}{L_\tau} \leq \frac{\Delta_{\tau_1}}{L_\tau} \leq \frac{2C_2\varepsilon\text{lfs}(p)}{\delta\text{lfs}(q)} \leq 4C_2\eta_0 \stackrel{\text{def}}{=} C_3.$$

The last inequality follows from the fact that  $\text{lfs}(p) \leq 2\text{lfs}(q)$  and  $\varepsilon/\delta \leq \eta_0$ .

3. From the definition of  $H_\tau(p, \omega(p))$ , we have for all vertices  $q \in \tau_p$

$$|H_\tau(p, \omega(p))| = \text{dist}(o_\tau, \text{aff}(\tau_p)) = \|o_\tau - o_{\tau_p}\| \leq \|o_\tau - q\|$$

Using the facts that  $\Phi_\tau \leq C_3 L_\tau$  (from part 2),  $L_\tau \leq \|p - q\|$ ,  $\omega(q) \leq \omega_0 \|p - q\|$ , and  $\|p - q\| \leq C_2 \varepsilon \text{lfs}(p)$  (from part 1), we have

$$\begin{aligned} \|o_\tau - q\| &= \sqrt{\Phi_\tau^2 + \omega(q)^2} \\ &\leq \sqrt{C_3^2 L_\tau^2 + \omega_0^2 \|p - q\|^2} \\ &\leq \|p - q\| \sqrt{C_3^2 + \omega_0^2} \\ &\leq C_2 \sqrt{C_3^2 + \omega_0^2} \times \varepsilon \text{lfs}(p) \\ &\stackrel{\text{def}}{=} C_4 \varepsilon \text{lfs}(p) \end{aligned}$$

□

### 4.3 Properties of inconsistent configurations

We now give lemmas on inconsistent configurations which are central to the proof of correctness of the reconstruction algorithm given later in the paper. The first lemma is the analog of Lemma 4.5 applied to inconsistent configurations. Differently from Lemma 4.5, we need to use Corollary 2.5 to control the orientation of the facets of  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  and require the following additional hypothesis relating the sampling rate  $\varepsilon$  and the fatness bound  $\Theta_0$ .

**Hypothesis 4.6**  $2A\varepsilon < 1$  where  $A \stackrel{\text{def}}{=} 2C_2C_3/\Theta_0^k$ , and  $C_2$  and  $C_3$  are the constants defined in Lemma 4.5.

**Lemma 4.7 (Tangent approximation)** *Let  $\tau$  be a  $k$ -simplex of  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  that is  $\Theta_0$ -fat. For all vertices  $p$  of  $\tau$ , we have  $\sin \angle(\text{aff}(\tau), T_p) = \sin \angle(T_p, \text{aff}(\tau)) \leq A\varepsilon$ . The constant  $A$ , defined in Hypothesis 4.6, depends on  $k$ ,  $\omega_0$ ,  $\eta_0$  and  $\Theta_0$ .*

**Proof.** Since  $\dim(\tau) = \dim(T_p) (= k)$ , we have from Lemma 2.1,  $\angle(\text{aff}(\tau), T_p) = \angle(T_p, \text{aff}(\tau))$ . From Lemma 4.5 (1) and the Triangle inequality, we get  $\Delta_\tau \leq 2C_2\varepsilon \text{lfs}(p)$ , and, from Lemma 4.5 (2), we get  $\Gamma_\tau \leq C_3$ . Using Corollary 2.5, we obtain

$$\sin \angle(\text{aff}(\tau), T_p) \leq \frac{\Gamma_\tau \Delta_\tau}{\Theta_\tau^k \text{lfs}(p)} \leq \frac{2C_2C_3\varepsilon}{\Theta_0^k} = A\varepsilon$$

□

**Lemma 4.8** *Assume that Hypotheses 3.1, 3.2 and 4.6 are satisfied. Let  $\phi \in \text{Inc}^\omega(\mathcal{P})$  be an inconsistent configuration witnessed by  $(u, v, w)$ . There exist positive constants  $C'_2 > C_2$ ,  $C'_3 > C_3$  and  $C'_4 > C_4$  that depend on  $\omega_0$  and  $\eta_0$  s.t., if  $\varepsilon < 1/C'_2$ , then*

1.  $\|p - i_\phi\| \leq \frac{C'_2}{2} \varepsilon \text{lfs}(p)$  for all vertices  $p$  of  $\phi$ .
2. If  $pq$  is an edge of  $\phi$  then  $\|p - q\| \leq C'_2 \varepsilon \text{lfs}(p)$ .

3. If  $\sigma \subseteq \phi$ , then  $\Phi_\sigma \leq C'_3 L_\sigma$  and  $\Gamma_\sigma = \Delta_\sigma / L_\sigma \leq C'_3$ .

4. If  $\sigma \subseteq \phi$  and  $p$  is any vertex of  $\sigma$ ,  $|H_\sigma(p, \omega(p))|$  of  $\sigma$  is at most  $C'_4 \varepsilon \text{lfs}(p)$ .

**Proof.** From the definition of inconsistent configurations, the  $k$ -dimensional simplex  $\tau = \phi \setminus \{w\}$  belongs to  $\text{Del}_u^\omega(\mathcal{P})$ . We first bound  $\text{dist}(i_\phi, \text{aff}(\tau)) = \|o_\tau - i_\phi\|$  where  $o_\tau$  is the orthocenter of  $\tau$ . Let  $m_u (\in \text{Vor}^\omega(\tau) \cap T_u)$  denote, as in Section 2.4, the point of  $T_u$  that is the center of the ball orthogonal to the weighted vertices of  $\tau$ . By definition,  $m_u$  is further than orthogonal to all other weighted points of  $\mathcal{P} \setminus \tau$ . Observe that  $\|u - o_\tau\| \leq \|u - m_u\|$ , since  $o_\tau$  belongs to  $\text{aff}(\tau)$  and therefore is the closest point to  $u$  in  $\text{aff}(\text{Vor}^\omega(\tau))$ . Moreover, by Lemma 4.4,  $\|u - m_u\| \leq C_1 \varepsilon \text{lfs}(u)$ . Then, by Lemma 4.1, we have for all vertices  $p \in \tau$

$$\|p - o_\tau\| \leq \frac{\|u - o_\tau\|}{\sqrt{1 - 4\omega_0^2}} \leq \frac{\|u - m_u\|}{\sqrt{1 - 4\omega_0^2}} \leq \frac{C_1 \varepsilon \text{lfs}(u)}{\sqrt{1 - 4\omega_0^2}}. \quad (7)$$

Since  $\tau$  is a  $\Theta_0$ -fat simplex (by definition of an inconsistent configuration), we get from Lemma 4.7 that

$$\sin \angle(\text{aff}(\tau), T_p) \leq A \varepsilon \quad (8)$$

for all vertices  $p$  of  $\tau$ . This implies, together with  $2A\varepsilon < 1$  (Hypothesis 4.6),

$$\tan^2 \angle(\text{aff}(\tau), T_u) \leq \frac{A^2 \varepsilon^2}{1 - A^2 \varepsilon^2} < 4A^2 \varepsilon^2, \quad (9)$$

Observing again that  $\|u - m_u\| \leq C_1 \varepsilon \text{lfs}(u)$  (from Lemma 4.4) and Eq. 8, we deduce

$$\|m_u - o_\tau\| \leq \|m_u - u\| \sin \angle(\text{aff}(\tau), T_u) \leq A C_1 \varepsilon^2 \text{lfs}(u). \quad (10)$$

We also have,  $\|v - o_\tau\| \leq \|v - m_u\|$  as  $o_\tau$  is the closest point to  $v$  in  $\text{aff}(\text{Vor}^\omega(\tau))$ . Hence we have, using Eqs (7) and (9),

$$\|m_v - o_\tau\| \leq \|v - o_\tau\| \tan \angle(\text{aff}(\tau), T_v) < \frac{2 A C_1 \varepsilon^2}{\sqrt{1 - 4\omega_0^2}} \text{lfs}(u) \quad (11)$$

Let  $i_\phi$  denote, as in Section 2.4, the first point of the line segment  $[m_u m_v]$  that is in  $\text{Vor}^\omega(\phi)$ . We get from Eqs (10) and (11) that

$$\|o_\tau - i_\phi\| \leq \frac{2 A C_1 \varepsilon^2}{\sqrt{1 - 4\omega_0^2}} \text{lfs}(u).$$

1. Using Lemma 4.4, and the facts that  $2A\varepsilon < 1$  and  $\|u - o_\tau\| \leq \|u - m_u\|$ , we get

$$\begin{aligned} \|u - i_\phi\| &\leq \|u - o_\tau\| + \|o_\tau - i_\phi\| \\ &\leq \|u - m_u\| + \|o_\tau - i_\phi\| \\ &\leq \left( C_1 \varepsilon + \frac{2 A C_1 \varepsilon^2}{\sqrt{1 - 4\omega_0^2}} \right) \text{lfs}(u) \\ &\leq \left( C_1 \varepsilon + \frac{C_1 \varepsilon}{\sqrt{1 - 4\omega_0^2}} \right) \text{lfs}(u) \\ &\leq \frac{C_2}{4} \varepsilon \text{lfs}(u) \end{aligned} \quad (12)$$

where  $C_2$  is the constant introduced in Lemma 4.5. Eq. (12), together with Lemma 4.1 and  $C'_2 \stackrel{\text{def}}{=} C_2/\sqrt{1-4\omega_0^2} > C_2$ , yields

$$\|p - i_\phi\| \leq \frac{\|u - i_\phi\|}{\sqrt{1-4\omega_0^2}} \leq \frac{C'_2}{4} \varepsilon \text{lfs}(u) \quad (13)$$

for all vertices  $p$  of  $\phi$ . We deduce  $\|p - u\| \leq \|p - i_\phi\| + \|u - i_\phi\| \leq \frac{C'_2}{2} \varepsilon \text{lfs}(u)$ .

We now express  $\text{lfs}(u)$  in terms of  $\text{lfs}(p)$  using the fact that  $\text{lfs}$  is 1-Lipschitz and using  $C'_2 \varepsilon < 1$ :

$$|\text{lfs}(p) - \text{lfs}(u)| \leq \|p - u\| \leq \frac{C'_2}{2} \varepsilon \text{lfs}(u) \leq \frac{1}{2} \text{lfs}(u). \quad (14)$$

We deduce that  $\text{lfs}(u) \leq 2\text{lfs}(p)$  and  $\|p - i_\phi\| \leq \frac{C'_2}{2} \varepsilon \text{lfs}(p)$

2. Using Eq. (13) and (14), from part 1 of this lemma, we have

$$\|p - q\| \leq \|p - i_\phi\| + \|q - i_\phi\| \leq \frac{C'_2}{2} \varepsilon \text{lfs}(u) \leq C'_2 \varepsilon \text{lfs}(p).$$

3. If  $\sigma$  belongs to  $\text{Del}_{\mathcal{T}\mathcal{M}}^\omega(\mathcal{P})$ , the result has been proved in Lemma 4.5 (3) with  $C'_3 = C_3$ . Let  $r_\phi = \sqrt{\|i_\phi - u\|^2 - \omega(u)^2}$ . Since  $i_\phi \in \text{Vor}^\omega(\sigma)$ , the sphere centered at  $i_\phi$  with radius  $r_\phi$  is orthogonal to the weighted vertices of  $\sigma$ . From Lemma 4.1 (2) and (3), we have  $r_\phi \geq \Phi_\phi$  and  $\Phi_\phi \geq \Phi_\sigma$  respectively. Hence it suffices to show that there exists a constant  $C'_3$  such that  $r_\phi \leq C'_3 L_\sigma$ . Using Eq. (12), we get

$$r_\phi = \sqrt{\|i_\phi - u\|^2 - \omega(u)^2} \leq \|i_\phi - u\| \leq \frac{C_2}{4} \varepsilon \text{lfs}(u).$$

Let  $q$  be a vertex of a shortest edge of  $\sigma$ . We have, from part 2, of this lemma,  $\|u - q\| \leq C'_2 \varepsilon \text{lfs}(q) < \text{lfs}(q)$ . From which we deduce that  $\text{lfs}(u) \leq 2\text{lfs}(q)$ . Therefore, using Hypothesis 3.1,

$$\Phi_\sigma \leq r_\phi \leq \frac{C_2}{2} \varepsilon \text{lfs}(q) = \frac{C_2}{2} \left( \frac{\varepsilon}{\delta} \times \delta \right) \times \text{lfs}(q) \leq \frac{C_2}{2} \eta_0 L_\sigma.$$

From part 2 of the lemma, we have  $\Delta_\phi \leq 2C'_2 \varepsilon \text{lfs}(q)$ , and using the facts that  $L_\sigma \geq \delta \text{lfs}(q)$  and  $\varepsilon/\delta \leq \eta_0$ , this implies

$$\Gamma_\sigma = \frac{\Delta_\sigma}{L_\sigma} \leq \frac{\Delta_\phi}{L_\sigma} \leq 2C'_2 \eta_0$$

We get the result by taking  $C'_3 \stackrel{\text{def}}{=} \max\{4C_2\eta_0, 2C'_2\eta_0\}$ .

4. Using the same arguments as in the proof of Lemma 4.5 (3) we have for all vertices  $q$  of  $\sigma_p$

$$|H_\sigma(p, \omega(p))| \leq \|o_\sigma - p\|$$

Using the facts that  $\Phi_\sigma \leq C'_3 L_\sigma$  (from part 3),  $L_\sigma \leq \|p - q\|$ ,  $\omega(q) \leq \omega_0 \|p - q\|$ , and  $\|p - q\| \leq 2C'_2 \varepsilon \text{lfs}(p)$  (from part 2), we have

$$\begin{aligned} \|o_\sigma - q\| &= \sqrt{\Phi_\sigma^2 + \omega(q)^2} \\ &\leq \sqrt{C'_3{}^2 L_\sigma^2 + \omega_0^2 \|p - q\|^2} \\ &\leq \|p - q\| \sqrt{C'_3{}^2 + \omega_0^2} \\ &\leq C'_2 \sqrt{C'_3{}^2 + \omega_0^2} \times \varepsilon \text{lfs}(p) \\ &\stackrel{\text{def}}{=} C'_4 \varepsilon \text{lfs}(p) \end{aligned}$$



□

The next crucial lemma bounds the fatness of inconsistent configurations.

**Lemma 4.9** *Assume Hypotheses 3.1, 3.2 and 4.6, and  $\varepsilon < 1/2C'_2$ . The fatness  $\Theta_\phi$  of an inconsistent configuration  $\phi$  is at most*

$$\left( \frac{C'_2 \varepsilon}{(k+1)!} \left( 1 + \frac{2C_3}{\Theta_0^k} \right) \right)^{\frac{1}{k+1}}$$

**Proof.** Let  $\phi$  be witnessed by  $(u, v, w)$ . From the definition of inconsistent configurations, the  $k$ -dimensional simplex  $\tau = \phi \setminus \{w\}$  belongs to  $\text{st}(u)$  and  $\tau$  is a  $\Theta_0$ -fat simplex. Using Lemma 4.8 (2), we have  $\Delta_\phi \leq 2C'_2 \varepsilon \text{lfs}(u)$ . As in the proof of Lemma 4.8, we have  $\sin \angle(T_u, \text{aff}(\tau)) \leq \frac{\Gamma_\tau \Delta_\tau}{\Theta_\tau^k \text{lfs}(u)}$  (refer to Eq. (8)) by Corollary 2.5 and the fact that  $\Delta_\phi < \text{lfs}(u)$ . Also, using the fact that  $\|u - w\| < C'_2 \varepsilon \text{lfs}(u) < \text{lfs}(u)$  (from Lemma 4.5 (2) and  $\varepsilon < 1/2C'_2$ ) and Lemma 2.3 (2), we have

$$\sin \angle(uw, T_u) \leq \frac{\|u - w\|}{2 \text{lfs}(u)} \leq \frac{\Delta_\phi}{2 \text{lfs}(u)}.$$

We can bound the altitude  $D_\phi(w)$  of  $w$  in  $\phi$

$$\begin{aligned} D_\phi(w) &= \text{dist}(w, \text{aff}(\tau)) \\ &= \sin \angle(uw, \text{aff}(\tau)) \times \|u - w\| \\ &\leq (\sin \angle(uw, T_u) + \sin \angle(\text{aff}(\tau), T_u)) \times \Delta_\phi \\ &\leq \left( \frac{\Delta_\phi}{2 \text{lfs}(u)} + \frac{\Gamma_\tau \Delta_\tau}{\Theta_\tau^k \text{lfs}(u)} \right) \Delta_\phi \\ &\leq \frac{\Delta_\phi^2}{2 \text{lfs}(u)} \left( 1 + \frac{2\Gamma_\tau}{\Theta_\tau^k} \right). \end{aligned} \tag{15}$$

From the definition of fatness of a simplex and Lemma 4.2 (1), we get

$$V_\tau = \Theta_\tau^k \Delta_\tau^k \leq \frac{\Delta_\tau^k}{k!}. \tag{16}$$

We deduce

$$\begin{aligned} \Theta_\phi^{k+1} &= \frac{V_\phi}{\Delta_\phi^{k+1}} \\ &= \frac{D_\phi(w) V_\tau}{(k+1)} \times \frac{1}{\Delta_\phi^{k+1}} \\ &\leq \frac{\Delta_\phi^2}{2 \text{lfs}(u)} \left( 1 + \frac{2\Gamma_\tau}{\Theta_\tau^k} \right) \times \frac{\Delta_\tau^k}{(k+1)! \Delta_\phi^{k+1}} \quad \text{using Eq. (15) and (16)} \\ &\leq \frac{C'_2 \varepsilon}{(k+1)!} \left( 1 + \frac{2C_3}{\Theta_0^k} \right). \end{aligned}$$

The last inequality comes from the facts that  $\sigma$  is  $\Theta_0$ -fat,  $\Delta_\phi \leq 2C'_2 \varepsilon \text{lfs}(u)$  (from Lemma 4.8 (2)) and  $\Gamma_\tau \leq C_3$  (from Lemma 4.5 (3)). □

A consequence of the lemma is that, if the subfaces of  $\phi$  are  $\Theta_0$ -fat simplices and if the following hypothesis

**Hypothesis 4.10**  $\frac{C'_2 \varepsilon}{(k+1)!} \left(1 + \frac{2C_3}{\Theta_0^k}\right) < \Theta_0^{k+1}$

is satisfied, then  $\phi$  is a  $\Theta_0$ -sliver. Hence, techniques to remove slivers can be used to remove inconsistent configurations.

In the above lemmas, we assumed that  $\varepsilon$  is small enough. Specifically in addition to Hypotheses 3.1, 3.2 and 4.6, we assumed that  $2C_2\varepsilon < 1$  in Lemma 4.5,  $C'_2\varepsilon < 1$  in Lemma 4.8 and  $2C'_2\varepsilon < 1$  in Lemma 4.9. We will make another hypothesis that subsumes these two previous conditions.

**Hypothesis 4.11**  $C'_2(1 + C'_2\eta_0)\varepsilon < 1/2$ .

Observe that this hypothesis implies  $C'_2(1 + C'_2)\varepsilon < 1/2$  since  $\eta_0 > 1$ . Observe also that Hypotheses 4.6, 4.10 and 4.11 are satisfied for fixed  $\omega_0$ ,  $\eta_0$ ,  $\Theta_0$  and a sufficiently small  $\varepsilon$ .

#### 4.4 Number of local neighbors

We will use the result from this section for the analysis of the algorithm, and also for calculating its time and space complexity.

Let  $N \stackrel{\text{def}}{=} (4C'_2\eta_0 + 6)^k$ , where the constant  $C'_2$  is defined in Lemma 4.8.

**Lemma 4.12** *Assume Hypotheses 3.1, 3.2, 4.6 and 4.11. The set*

$$LN(p) = \{q \in \mathcal{P} : |B(p, \|p - q\|) \cap \mathcal{P}| \leq N\},$$

where  $N = 2^{O(k)}$  and the constant in big-O depends on  $\omega_0$  and  $\eta_0$ , includes all the points of  $\mathcal{P}$  that can form an edge with  $p$  in  $C^\omega(\mathcal{P})$ .

**Proof.** Lemmas 4.5 and 4.8 show that, in order to construct  $\text{st}(p)$  and search for inconsistencies involving  $p$ , it is enough to consider the points of  $\mathcal{P}$  that lie in ball  $B_p = B(p, C'_2\varepsilon \text{lfs}(p))$ . Therefore it is enough to count the number of points in  $B_p \cap \mathcal{P}$ .

Let  $x$  and  $y$  be two points of  $B_p \cap \mathcal{P}$ . Since  $\text{lfs}()$  is a 1-Lipschitz function, we have

$$\text{lfs}(p)(1 - C'_2\varepsilon) \leq \text{lfs}(x), \text{lfs}(y) \leq \text{lfs}(p)(1 + C'_2\varepsilon). \quad (17)$$

By definition of an  $(\varepsilon, \delta)$ -sample of  $\mathcal{M}$ , the two balls  $B_x = B(x, r_x)$  and  $B_y = B(y, r_y)$ , where  $r_x = \delta \text{lfs}(x)/2$  and  $r_y = \delta \text{lfs}(y)/2$ , are disjoint. Moreover, both balls are contained in the ball  $B_p^+ = B(p, r^+)$ , where

$$r^+ = C'_2\varepsilon \text{lfs}(p) + (1 + C'_2\varepsilon)\delta \text{lfs}(p).$$

Let  $\bar{B}_x = B_x \cap T_p$ ,  $\bar{B}_y = B_y \cap T_p$  and  $\bar{B}_p^+ = B_p^+ \cap T_p$ . From Lemma 2.3(2), the distance from  $x$  to  $T_p$  is

$$\text{dist}(x, T_p) = \|p - x\| \times \sin(\angle px, T_p) \leq C'^2_2 \varepsilon^2 \text{lfs}(p)/2. \quad (18)$$

Using Eq.s (17), (18) and the fact that  $\varepsilon/\delta \leq \eta_0$ , we see that  $\bar{B}_x$  is a  $k$ -dimensional ball of squared radius

$$\begin{aligned} \delta^2 \text{lfs}^2(x)/4 - \text{dist}(x, T_p)^2 &\geq \delta^2 \text{lfs}^2(p)(1 - C'_2\varepsilon)^2/4 - C'^4_2 \varepsilon^4 \text{lfs}(p)/4 \\ &\geq \delta^2 \text{lfs}^2(p)/4 \times \left((1 - C'_2\varepsilon)^2 - C'^4_2 \eta_0^2 \varepsilon^2\right) \\ &\stackrel{\text{def}}{=} (r^-)^2. \end{aligned}$$

We can now use a packing argument. Since the balls  $\overline{B}_x$ ,  $x$  in  $B_p \cap \mathcal{P}$ , are disjoint and all contained in  $B_p^+$ , the number of points of  $B_p \cap \mathcal{P}$  is at most

$$\begin{aligned} \left(\frac{r^+}{r^-}\right)^k &= \left(\frac{(C'_2\varepsilon + (1 + C'_2\varepsilon)\delta)^2}{\delta^2/4 \times ((1 - C'_2\varepsilon)^2 - C_2'^4\eta_0^2\varepsilon^2)}\right)^{k/2} \\ &\leq \left(\frac{4(C'_2\eta_0 + (1 + C'_2\varepsilon))^2}{(1 - C'_2\varepsilon)^2 - C_2'^4\eta_0^2\varepsilon^2}\right)^{k/2} \\ &= \left(\frac{4(C'_2\eta_0 + (1 + C'_2\varepsilon))^2}{(1 - C'_2\varepsilon - C_2'^2\eta_0\varepsilon)(1 - C'_2\varepsilon + C_2'^2\eta_0\varepsilon)}\right)^{k/2} \\ &\leq (4C'_2\eta_0 + 6)^k \stackrel{\text{def}}{=} N \quad (\text{using Hypothesis 4.11}) \end{aligned}$$

And the result follows.  $\square$

#### 4.5 Correctness of the algorithm

**Definition 4.13 (Sliverity range)** Let  $\omega$  be a weight assignment satisfying Hypothesis 3.2. The weight of all the points in  $\mathcal{P} \setminus \{p\}$  are fixed and the weight  $\omega(p)$  of  $p$  is varying. The sliverity range  $\Sigma(p)$  of a point  $p \in \mathcal{P}$  is the measure of the set of all squared weights  $\omega(p)^2$  for which  $p$  is a vertex of a  $\Theta_0$ -sliver in  $C^\omega(\mathcal{P})$ .

**Lemma 4.14** Under Hypotheses 3.1, 3.2, 4.6, 4.10 and 4.11, the sliverity range satisfies

$$\Sigma(p) < 2N^{k+1}C_5\Theta_0\mathfrak{m}(p)^2$$

for some constant  $C_5$  that depends on  $k$ ,  $\omega_0$  and  $\eta_0$  but not on  $\Theta_0$ .

**Proof.** Let  $\tau$  be a  $j$ -dimensional simplex of  $C^\omega(\mathcal{P})$  incident on  $p$  (with  $2 \leq j \leq k+1$ ). assume that  $\tau$  is a  $\Theta_0$ -sliver. If  $\omega(p)$  is the weight of  $p$ , we write  $H_\tau(p, \omega(p))$  for the excentricity of  $\tau$  with respect to  $p$  and parameterized by  $\omega(p)$ . From Lemma 4.8(4), we have

$$|H_\tau(p, \omega(p))| \leq C'_4\varepsilon\text{fs}(p) \stackrel{\text{def}}{=} D \quad (19)$$

Using Lemma 4.2 (2), we have

$$D_p(\tau) \leq j\Gamma_\tau^{j-1}\Delta_\tau \times \frac{\Theta_\tau^j}{\Theta_{\tau_p}^{j-1}} \leq (k+1)C_3'^k\Theta_0\Delta_\tau \stackrel{\text{def}}{=} E \quad (20)$$

The last inequality follows from the facts that  $j \leq k+1$ ,  $\Gamma_\tau \leq C_3'$  (from Lemmas 4.5 (2) and 4.8 (3)) and  $\tau$  is a  $\Theta_0$ -sliver. Moreover, from Lemma 4.3,

$$H_\tau(p, \omega(p)) = H_\tau(p, 0) - \frac{\omega(p)^2}{2D_p(\tau)}. \quad (21)$$

It then follows from Eq. (19), (20) and (21) that the set of squared weights of  $p$  for which  $\tau$  belongs to  $C^\omega(\mathcal{P})$  is a subset of the following interval

$$[2D_p(\tau)H_\tau(p, 0) - \beta, 2D_p(\tau)H_\tau(p, 0) + \beta],$$

where  $\beta = 2DE$ . Therefore, from Eq. (20) and (21), the measure of the set of weights for which  $\tau$  belongs to  $C^\omega(\mathcal{P})$  is at most

$$2\beta = 4DE = 4(k+1)C_3'^k \Theta_0 \Delta_\tau C_4' \varepsilon \text{lfs}(p).$$

Let  $q_1$  and  $q_2$  be two vertices of  $\tau$  such that  $\Delta_\tau = \|q_1 - q_2\|$ . Using Lemmas 4.5 (1) and 4.8 (2), we get

$$\Delta_\tau \leq \|p - q_1\| + \|p - q_2\| \leq 2C_2' \varepsilon \text{lfs}(p).$$

Using this inequality,  $\text{lfs}(p) \leq \mathfrak{m}(p)/\delta$  (Lemma 2.3) and  $\varepsilon/\delta \leq \eta_0$  (Hypothesis 3.1), the sliverity range of  $\tau$  is at most  $8(k+1)C_3'^k C_2' C_4' \Theta_0 \eta_0^2 \mathfrak{m}(p)^2 = C_5 \Theta_0 \mathfrak{m}(p)^2$  with  $C_5 \stackrel{\text{def}}{=} 8(k+1)C_3'^k C_2' C_4' \eta_0^2$ . By Lemma 4.12, the number of  $j$ -simplices that are incident to  $p$  is at most  $N^j$ . Hence, the sliverity range of  $p$  is at most

$$\Sigma(p) \leq \sum_{j=3}^{k+1} N^j C_5 \Theta_0 \mathfrak{m}(p)^2 < 2N^{k+1} C_5 \Theta_0 \mathfrak{m}(p)^2.$$

The last inequality follows from the fact that  $\sum_{j=3}^{k+1} N^j < 2N^{k+1}$  (as  $N = (4C_2' \eta_0 + 6)^k > 2$ ).  $\square$

**Theorem 4.15** *Take  $\Theta_0 = \frac{\omega_0^2}{2N^{k+1}C_5}$  and assume that Hypotheses 3.1, 3.2, 4.6, 4.10 and 4.11 are satisfied. Then, the simplicial complex  $\hat{\mathcal{M}}$  output by Algorithm 1 has no inconsistencies and its simplices are all  $\Theta_0$ -fat.*

**Proof.** The sliverity range  $\Sigma(p)$  of  $p$  is at most  $2N^{k+1}C_5\Theta_0\mathfrak{m}(p)^2$  from Lemma 4.14. Since  $\Theta_0 = \frac{\omega_0^2}{2N^{k+1}C_5}$ ,  $\Sigma(p)$  is less than the total range of possible squared weights  $\omega_0^2 \mathfrak{m}(p)^2$ . Hence, Function **weight** ( $p, \omega$ ) will always find a weight for any point  $p \in \mathcal{P}$  and any weight assignment of relative amplitude at most  $\omega_0$  for the points of  $\mathcal{P} \setminus \{p\}$ .

Since the algorithm removes all the simplices of  $C^\omega(\mathcal{P})$  that are not  $\Theta_0$ -fat, all the simplices of  $\hat{\mathcal{M}}$  are  $\Theta_0$ -fat.

By Lemma 4.9 and Hypothesis 4.10, all the inconsistent configurations in  $C^\omega(\mathcal{P})$  are either  $\Theta_0$ -slivers or contain a subfaces which are  $\Theta_0$ -slivers. It follows that  $\hat{\mathcal{M}}$  has no inconsistency since, when the algorithm terminates, all simplices of  $C^\omega(\mathcal{P})$  are  $\Theta_0$ -fat.  $\square$

## 4.6 Time and space complexity

**Theorem 4.16** *Assume that Hypotheses 3.1, 3.2, 4.6, 4.10 and 4.11 are satisfied. Then the space complexity of the algorithm is*

$$(O(d) + 2^{O(k^2)})|\mathcal{P}|$$

*and the time complexity is*

$$O(d) |\mathcal{P}|^2 + d 2^{O(k^2)} |\mathcal{P}|,$$

*where  $|\mathcal{P}|$  denotes the cardinality of  $\mathcal{P}$ .*

**Proof.** **Space Complexity:** For each point  $p \in \mathcal{P}$  we maintain  $LN(p)$ . The total space complexity for storing  $LN(p)$  for each point  $p \in \mathcal{P}$  is thus  $O(N|\mathcal{P}|)$  by definition of  $LN(p)$ .

Rather than maintaining  $C^\omega(\mathcal{P})$  as one data structure, for all  $p \in \mathcal{P}$ , we maintain  $\text{st}(p)$  and  $\text{cosph}(p)$ , where

$$\begin{aligned} \text{cosph}(p) = & \left\{ \text{inconsistent configuration } \phi \mid \right. \\ & \left. \phi \text{ is witnessed by } (p, q, r), \text{ for some } q, r \in \mathcal{P} \setminus p \right\} \end{aligned}$$

Note that

$$C^\omega(\mathcal{P}) = \bigcup_{p \in \mathcal{P}} (\text{st}(p) \cup \text{cosph}(p)).$$

By Lemma 2.2, each  $\text{st}(p)$ ,  $p \in \mathcal{P}$ , has the same combinatorial complexity as a Voronoi cell in the  $k$ -dimensional flat  $T_p$ . Since the sites needed to compute this Voronoi cell all belong to  $LN(p)$ , their number is at most  $N$  by Lemma 4.12. From the Upper Bound Theorem of convex geometry, see e.g. [BY98], the combinatorial complexity of each star is therefore  $O(N^{\lfloor k/2 \rfloor})$ . Hence the total space complexity of the tangential Delaunay complex is  $O(kN^{\lfloor k/2 \rfloor})|\mathcal{P}|$ .

For a given inconsistent  $\Theta_0$ -fat  $k$ -simplex in  $\text{st}(p)$ , we can have from Lemmas 4.8 (2) and 4.12, at most  $k|LN(p)| \leq kN$  different inconsistent configurations. Hence, the number of inconsistent configurations to be stored in the completed complex  $C^\omega(\mathcal{P})$  is at most  $O(k^2N^{\lfloor k/2 \rfloor+1})|\mathcal{P}|$ .

The main loop of the algorithm processes each point  $p \in \mathcal{P}$  in turn, and finds a weight assignment  $\omega(p)$  using the function **weight**(, ). In the function **weight**, we maintain a list of candidate simplices of  $p$  with their corresponding forbidden intervals and a priority queue of candidate simplices of  $p$ . By Lemmas 4.5 (1), 4.8 (2) and 4.12, the space complexities of both these data structures are bounded by  $O(kN^{k+1})$ .

With  $N = 2^{O(k)}$  (refer to Lemma 4.12), we conclude that the total space complexity of the algorithm is

$$O(k^2N^{\lfloor k/2 \rfloor+1}|\mathcal{P}| + d|\mathcal{P}| + kN^{k+1}) = (O(d) + 2^{O(k^2)})|\mathcal{P}|.$$

**Time complexity:** In the initialization phase, the algorithm computes  $LN(p)$  for all  $p \in \mathcal{P}$  and initializes the weights to 0. This can easily be done in time  $O(d)|\mathcal{P}|^2$ .

Then the algorithm builds  $C^\omega(\mathcal{P})$  for the zero weight assignment. The time to compute  $\text{st}(p)$  is dominated by the time to compute the cell of  $p$  in the weighted  $k$ -dimensional Voronoi diagram of the projected points of  $LN(p)$  onto  $T_p$ . Since, by definition,  $|LN(p)| \leq N$ , the time for building the star of  $p$  is the same as the time to compute the intersection of  $N$  halfspaces in  $\mathbb{R}^k$ ,

$$O(kdN + k^3(N \log N + N^{\lfloor k/2 \rfloor})),$$

see e.g. [Cha93, BY98]. The factor  $O(kd)$  appears in the first term because calculating the projection of a point in  $\mathbb{R}^d$  on a  $k$ -flat requires to compute  $k$  inner products. The  $O(k^3)$  factor comes from the fact that the basic operation we need to perform is to decide whether a point lies in the ball orthogonal to a  $k$ -simplex. This operation reduces to the evaluation of the sign of the determinant of a  $(k+2) \times (k+2)$  matrix. The  $N^{\lfloor k/2 \rfloor}$  term bounds the combinatorial complexity of a cell in the Voronoi diagram of  $N$  sites in a  $k$ -flat. Therefore the time needed to build the stars of all the points  $p$  in  $\mathcal{P}$  is

$$O(kdN + k^3(N \log N + N^{\lfloor k/2 \rfloor}))|\mathcal{P}|.$$

Let  $\tau = [p_0, \dots, p_k]$  be a  $\Theta_0$ -fat  $k$ -simplex in  $\text{st}(u)$ . For each vertex  $v (\neq u)$  of  $\tau$  with  $\tau \notin \text{st}(v)$ , we need to compute the inconsistent configurations of the form  $\phi = [p_0, \dots, p_k, w]$  witnessed by

$(u, v, w)$  where  $w \in LN(p) \setminus \tau$ . The number of such inconsistent configurations is therefore less than  $|LN(p)| \leq N$ . The time complexity to compute all the inconsistent configurations of the form  $\phi = [p_0, \dots, p_k, w]$  witnessed by the triplet  $(u, v, w)$  is  $O(dN)$ . Since the number of choices of  $v$  is at most  $k$ , hence the time complexity for building all the inconsistent configurations of the form  $\phi = [p_0, \dots, p_k, w]$  witnessed by  $(u, v, w)$  with  $v (\neq u)$  being a vertex of  $\tau$  and  $w$  a point in  $LN(u) \setminus \tau$  is

$$O(dkN) \tag{22}$$

The time complexity to build all the inconsistent configurations corresponding to  $\text{st}(u)$  is  $O(dkN^{\lfloor k/2 \rfloor + 1})$  since the number of  $k$ -simplices in the star of a point  $p$  is  $O(N^{\lfloor k/2 \rfloor})$ .

Hence, the time complexity for building the inconsistent configurations of  $C^\omega(\mathcal{P})$  is  $O(dN + kN^{\lfloor k/2 \rfloor + 1}) |\mathcal{P}|$ . Therefore the total time complexity of the initialization phase is

$$O(dkN + k^3 N \log N + (dk + k^3) N^{\lfloor k/2 \rfloor + 1}) |\mathcal{P}|$$

Consider now the main loop of the algorithm. The time complexity of function **weight** $(p, \omega)$  is  $O((d + k^3) N^{k+1})$  since we need to sweep over at most all  $(k + 1)$ -simplices incident on  $p$  with vertices in  $LN(p)$ . The number of such simplices is at most  $N^{k+1}$ . We easily deduce from the above discussion that the time complexity of Function **update\_complete\_complex** $(LN(p), \omega)$  is

$$O(dkN + k^3 N \log N + (dk + k^3) N^{\lfloor k/2 \rfloor + 1}) N.$$

Since functions **weight** $(p, \Theta_0, \omega)$  and **update\_complete\_complex** $(C^\omega(\mathcal{P}), p, \omega)$  are called  $|\mathcal{P}|$  times, we conclude that the time complexity of the main loop of the algorithm is  $O(dk N^2 + (k^3 + dk) N^{k+1}) |\mathcal{P}|$ .

Combining the time complexities for all the steps of the algorithm and using  $N = O2^{O(k)}$  (refer to Lemma 4.12), we get the total time complexity of the algorithm

$$O(d) |\mathcal{P}|^2 + O(dk N^2 + (dk + k^3) N^{k+1}) |\mathcal{P}| = O(d) |\mathcal{P}|^2 + d 2^{O(k^2)} |\mathcal{P}|$$

□

Observe that, since  $\mathcal{P}$  is an  $(\varepsilon, \delta)$ -sample of  $\mathcal{M}$  with  $\varepsilon/\delta \leq \eta_0$ ,  $|\mathcal{P}| = O(\varepsilon^k)$ .

## 5 Topological and geometric guarantees

In this section we give conditions under which  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  is isotopic to and a close geometric approximation of the manifold  $\mathcal{M}$ . The results in this section will be used to prove the theoretical guarantees on the quality of the simplicial complex  $\hat{\mathcal{M}}$  output by the Algorithm 1 in Section 3.

### General position assumption

We assume that the weighted points in  $\mathcal{P}^\omega$  are in general position in  $\mathbb{R}^d$  and this will imply  $\text{Del}^\omega(\mathcal{P})$  is a simplicial complex, see Section 2.1.2. Therefore  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  is a simplicial complex.

We know from Lemma 2.2 that  $\text{Vor}^\omega(\mathcal{P}) \cap T_p$  is identical to the weighted Voronoi diagram  $\text{Vor}^{\xi_p}(\mathcal{P}_p)$  of  $\mathcal{P}_p$  in  $T_p$ , where  $\mathcal{P}_p$  is obtained by projecting  $\mathcal{P}$  orthogonally onto  $T_p$  and  $\xi_p$  is the weight assignment defined in Lemma 2.2.

Under Hypotheses 3.1 and 3.2,  $\text{Vor}^{\xi_p}(p)$  is bounded (Lemma 4.4) and therefore  $\dim \text{aff}(\mathcal{P}_p) = k$ . In addition to the general position assumption of  $\mathcal{P}^\omega$ , we will also assume that the weighted points  $\mathcal{P}_p^{\xi_p}$  are in *general position* on  $T_p$ , meaning that there exists no orthosphere centered on  $T_p$  that is orthogonal to  $k + 2$  weighted points of  $\mathcal{P}_p^{\xi_p}$ . See Sections 2.1.1 and 2.1.2. As we have already mentioned in Section 2.4, the general position assumption can be fulfilled by applying an infinitesimal perturbation to the point sample or to the weight assignment  $\omega$ .

Under the general position assumption above,  $\text{Del}^{\xi_p}(\mathcal{P}_p)$  is a triangulation of  $\text{conv}(\mathcal{P}_p)$  [Aur87, AE84, BDG12]. Moreover, since  $\text{Vor}^{\xi_p}(p)$  is bounded (from Lemma 4.4),  $p$  is an interior vertex of  $\text{Del}^{\xi_p}(\mathcal{P}_p)$ . This implies the following result.

**Lemma 5.1** *For all  $p \in \mathcal{P}$ ,  $\text{st}(p)$  is isomorphic to the star of an interior vertex of a compact triangulated  $k$ -dimensional polytope.*

### Medial axis and the projection map

Let  $\mathcal{O}$  denote the medial axis of  $\mathcal{M}$ , and let

$$\pi : \mathbb{R}^d \setminus \mathcal{O} \longrightarrow \mathcal{M}$$

denote the projection map that maps each point of  $\mathbb{R}^d \setminus \mathcal{O}$  to its closest point on  $\mathcal{M}$ . The following lemma is a standard result from Federer [Fed69].

**Lemma 5.2** *Let  $\mathcal{M}$  be a smooth submanifold of  $\mathbb{R}^d$  without boundary. Then, the projection map  $\pi : \mathbb{R}^d \setminus \mathcal{O} \rightarrow \mathcal{M}$  is a  $C^1$ -function in  $\mathbb{R}^d \setminus \mathcal{O}$ .*

1. *The map  $\pi$  is a  $C^1$ -function.*
2. *For all  $x \in \mathbb{R}^d \setminus \mathcal{O}$ , the kernel of the linear map  $d\pi(x) : \mathbb{R}^d \rightarrow T_{\pi(x)}$ , where  $d\pi(x)$  denotes the derivative of  $\pi$  at  $x$ , is parallel to  $N_{\pi(x)}$  and has dimension  $d - k$ .*

### Main result

The main result of this section is the following theorem.

**Theorem 5.3** *Assume that Hypotheses 3.1 and 3.2 are satisfied as well as the following conditions:*

**C1.**  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  *has no  $k$ -dimensional inconsistent simplices.*

**C2.** *All the simplices in  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$  are  $\Theta_0$ -fat.*

*Then there exists  $\varepsilon_0$  that depends only on  $k$ ,  $\omega_0$ ,  $\eta_0$ , and  $\Theta_0$  such that for all  $\varepsilon \leq \varepsilon_0$ , the simplicial complex  $\text{Del}_{T\mathcal{M}}^\omega(\mathcal{P})$ , also noted  $\hat{\mathcal{M}}$  for convenience, and the map  $\pi|_{\hat{\mathcal{M}}}$  satisfy the following properties:*

- P1.** *Tangent space approximation: Let  $\tau$  be a  $k$ -simplex in  $\hat{\mathcal{M}}$ . For all vertices  $p$  of  $\tau$ , we have  $\sin \angle(T_p, \text{aff}(\tau)) = \sin \angle(\text{aff}(\tau), T_p) = O(\varepsilon)$ . Note that this property directly follows from Lemma 4.7.*
- P2.** *PL  $k$ -manifold without boundary:  $\hat{\mathcal{M}}$  is a piecewise linear  $k$ -manifold without boundary;*

**P3. Homeomorphism:** The map  $\pi|_{\hat{\mathcal{M}}}$  provides a homeomorphism between  $\hat{\mathcal{M}}$  and  $\mathcal{M}$ ;

**P4. Pointwise approximation:**  $\forall x \in \mathcal{M}, \text{dist}(x, \pi|_{\hat{\mathcal{M}}}^{-1}(x)) = O(\varepsilon^2 \text{fs}(x))$ ;

**P5. Isotopy:** There exists an isotopy  $F : \mathbb{R}^d \times [0, 1] \rightarrow \mathbb{R}^d$  such that the map  $F(\cdot, 0)$  restricted to  $\hat{\mathcal{M}}$  is the identity map on  $\hat{\mathcal{M}}$  and  $F(\hat{\mathcal{M}}, 1) = \mathcal{M}$ .

The constants in the big- $O$  notations depend on  $k, \omega_0, \eta_0$  and  $\Theta_0$ .

Since Conditions **C1** and **C2** are satisfied when the conditions of Theorem 4.15 are satisfied, it follows that the simplicial complex  $\hat{\mathcal{M}}$  output by Algorithm 1 satisfies the above properties **P1-P5** provided that the additional condition on  $\varepsilon$  in the above theorem is satisfied.

The rest of the section will be devoted to the proof of Theorem 5.3.

## 5.1 Piecewise-linear $k$ -manifold

The following lemma is a direct consequence of condition **C1**.

**Lemma 5.4** *Under condition **C1** of Theorem 5.3 and for a sufficiently small  $\varepsilon$ ,  $\text{st}(p) = \text{st}(p, \hat{\mathcal{M}})$  for all  $p \in \mathcal{P}$ .*

We can now prove that  $\hat{\mathcal{M}}$  is a PL  $k$ -manifold without boundary (Property **P2**).

**Lemma 5.5 (PL  $k$ -manifold without boundary)** *Assume condition **C1** of Theorem 5.3. If  $\varepsilon$  is sufficiently small, then  $\hat{\mathcal{M}}$  is a PL  $k$ -manifold without boundary.*

**Proof.** Since  $\text{Del}^{\xi_p}(\mathcal{P}_p)$  is a triangulation of a  $k$ -dimensional polytope,  $\text{Del}^{\xi_p}(\mathcal{P}_p)$  is a PL  $k$ -manifold [Zee66, Lem. 8 of Chap. 3] and, since  $p$  is an interior vertex of  $\text{Del}^{\xi_p}(\mathcal{P}_p)$ ,  $\text{lk}(p, \text{Del}^{\xi_p}(\mathcal{P}_p))$  is a PL  $(k-1)$ -sphere [Zee66, Proof of Lem. 9 of Chap. 3].

From Lemma 5.4, we have  $\text{st}(p) = \text{st}(p, \hat{\mathcal{M}})$ . By Lemma 5.1,  $\text{st}(p, \hat{\mathcal{M}})$  is isomorphic to the star of an interior vertex of a  $k$ -dimensional triangulated convex domain. Hence,  $\text{lk}(p, \hat{\mathcal{M}})$  is a PL  $(k-1)$ -sphere. This directly implies that  $\hat{\mathcal{M}}$  is a PL  $k$ -manifold with no boundaries.  $\square$

Property **P2** follows since the output  $\hat{\mathcal{M}}$  of the Algorithm 1 has no inconsistency and therefore satisfies Condition **C1**.

## 5.2 Homeomorphism

Following lemma establishes Property **P3** of Theorem 5.3, and the rest of the section is devoted to its proof.

**Lemma 5.6 (Homeomorphism)** *Assume conditions **C1** and **C2** of Theorem 5.3. For  $\varepsilon$  sufficiently small, the map  $\pi$  restricted to  $\hat{\mathcal{M}}$  gives a homeomorphism between  $\hat{\mathcal{M}}$  and  $\mathcal{M}$ .*

In the broad outline, the proof of Lemma 5.6 is similar to the proof of homeomorphism given in [ACDL02] but the technical details are quite different in most places. The difficulties that arose in getting results analogous to the ones in [ACDL02] were handled using ideas from [Whi57, Chap. II].

In the rest of this paper, we assume that  $\omega_0, \eta_0, \Theta_0$  are fixed constants, and that  $\varepsilon$  is small enough. We use the asymptotic notations  $O()$  and  $\Omega()$  hiding the constants.

**Remark 5.7** *For the details of the missing proofs of Lemmas 5.8, 5.9, 5.10, and 5.11 refer to [Gho12, Chap. 4].*



### 5.2.1 Basic lemmas

For a point  $p \in \mathcal{P}$ , let

$$\pi_p : \mathbb{R}^d \rightarrow T_p$$

denotes the orthogonal projection of  $\mathbb{R}^d$  onto  $T_p$ , and

$$\tilde{\pi}_p : T_p \setminus \mathcal{O} \rightarrow \mathcal{M}$$

denotes the restriction of  $\pi$  to  $T_p \setminus \mathcal{O}$ .

We define in addition

$$\pi_p^* : \mathbb{R}^d \setminus \mathcal{O} \rightarrow T_p$$

as the map that maps each point  $x \in \mathbb{R}^d \setminus \mathcal{O}$  to the point of intersection of  $T_p$  and  $N_{\pi(x)}$ , the normal space of  $\mathcal{M}$  at  $\pi(x)$ .

The following lemma is a generalization of Proposition 6.2 in [NSW08] which bounds the variation of the angle between tangent spaces between two points on the manifold  $\mathcal{M}$ .

**Lemma 5.8 (Tangent variation)** *Let  $p, q \in \mathcal{M}$  and  $\|p - q\| = t \text{lfs}(p)$  with  $t \leq 1$ . Then,*

$$\sin \angle(T_q, T_p) = \sin \angle(T_p, T_q) < O(t).$$

Using Lemma 5.8, we can show that the derivative of the restriction  $\tilde{\pi}_p$  of  $\pi_p$  to a neighborhood of  $p$  is nonsingular and that  $\tilde{\pi}_p$  is injective. Using the Inverse function theorem this implies that  $\tilde{\pi}_p$  is a diffeomorphism.

**Lemma 5.9 (  $\tilde{\pi}_p$  is a  $C^1$ -diffeomorphism )** *There exists absolute constants  $r_1$  and  $r_2$  such that  $\tilde{\pi}_p$  restricted to  $\tilde{T}_p \stackrel{\text{def}}{=} B(p, r_1 \text{lfs}(p)) \cap T_p$  is a diffeomorphism onto  $\tilde{\pi}_p(\tilde{T}_p)$  and  $\tilde{\pi}_p(\tilde{T}_p) \supseteq \tilde{\mathcal{M}}_p \stackrel{\text{def}}{=} B_{\mathcal{M}}(p, r_2 \text{lfs}(p))$ .*

Note that  $\tilde{\mathcal{M}}_p \subseteq \tilde{\pi}_p(\tilde{T}_p)$  follows from Lemma 2.3 (3).

The next lemma follows directly from Lemmas 2.3 4.7, 5.8 and 5.9.

**Lemma 5.10** *Under Conditions C1 and C2 of Theorem 5.3, for any  $x \in \text{st}(p, \hat{\mathcal{M}})$ .*

$$\max\{ \|\pi_p(x) - x\|, \|\pi(x) - x\|, \|\pi_p^*(x) - x\| \} = O(\varepsilon^2) \text{lfs}(p).$$

We will need the following technical lemma bounding the fatness of a perturbed simplex. A similar inequality can be found in the proof of Lemma 14c in [Whi57, Chapter 4].

**Lemma 5.11** *Let  $\sigma = [p_0, \dots, p_j]$  be a  $j$ -dimensional simplex of  $\mathbb{R}^d$ , and let  $\sigma' = [p'_0, \dots, p'_j]$  where  $\|p_i - p'_i\| \leq \rho \Delta_\sigma$  with  $\rho < 1/2$ , for  $0 \leq i \leq j$ . Then,*

$$\Theta_{\sigma'}^{2j} \geq \frac{\Theta_\sigma^{2j}}{(1 + 2\rho)^{2j}} - \frac{2j\rho}{1 + 2\rho}$$

*This directly implies that, if  $\rho \leq \frac{\Theta_\sigma^{2j}}{j2^{j+1}} (1 - \frac{1}{2j})$ , then  $\Theta_{\sigma'} \geq \Theta_\sigma/2$ .*

We will also need the following standard lemma from convex geometry which bounds the distance between an interior point and a point on the boundary of a simplex. The proof can be found in [Whi57, Lem. 14b of Chap. 4].

**Lemma 5.12** *Let  $\sigma = [p_0, \dots, p_j]$  be a  $j$ -simplex and let  $p$  be a point of sigma. Writing  $p = \sum_{i=0}^j \mu_i p_i$  with  $\sum_{i=0}^j \mu_i = 1$  and  $\mu_i \geq 0$ , we have*

$$\text{dist}(p, \partial\sigma) \geq j! \Theta_\sigma^j \Delta_\sigma \times \min\{\mu_0, \dots, \mu_j\}$$

We will use the standard notion of *oriented simplices*, *positively oriented simplices* and *oriented PL manifolds* (see [Zee66, RS72] or [Whi57, Appendix II]).

Recall that the restriction of  $\pi_p$  to  $\text{st}(p)$  provides a simplicial isomorphism between  $\text{st}(p)$  and  $\text{st}(p, \text{Del}^{\xi_p}(\mathcal{P}_p))$  (Lemma 2.2), and that  $\text{st}(p) = \text{st}(p, \hat{\mathcal{M}})$  (by Lemma 5.4). We orient  $T_p$  and orient positively the  $k$ -dimensional simplices of  $\text{st}(p, \text{Del}^{\xi_p}(\mathcal{P}_p))$ . We then orient the  $k$ -dimensional simplices of  $\text{st}(p, \hat{\mathcal{M}})$  isomorphically, using the orientation of the corresponding  $k$ -dimensional simplices of  $\text{st}(p, \text{Del}^{\xi_p}(\mathcal{P}_p))$ .

Since  $\text{lk}(p, \hat{\mathcal{M}})$  is a PL  $(j-1)$ -sphere (see the proof of Lemma 5.5), we immediately get that  $\text{st}(p, \hat{\mathcal{M}})$  is an oriented PL  $k$ -ball.

**Definition 5.13 (Simplexwise positive map)** *Let  $\sigma$  be a positively oriented  $i$ -simplex of  $\mathbb{R}^i$ , and let  $f : \sigma \rightarrow \mathbb{R}^i$  be a  $C^1$ -function. The map  $f$  is called *simplexwise positive* if  $\det(J(f)) > 0$  for all  $x \in \sigma$ , where  $J(f)$  and  $\det(J(f))$  denote the Jacobian and the determinant of the Jacobian of the map  $f$  respectively.*

The following result is a special case of a standard result from PL-topology (see, e.g., Appendix II of [Whi57, Lemma 15a]).

**Lemma 5.14** *Let  $\mathcal{K}$  be a  $j$ -dimensional PL-manifold whose  $j$ -simplices are oriented, and let the continuous map  $f : \mathcal{K} \rightarrow \mathbb{R}^j$  be a simplexwise positive map. Then, for any connected open subset  $W$  of  $\mathbb{R}^j \setminus f(\partial\mathcal{K})$ , any two points of  $W$  not in  $f(\mathcal{K}^{-1})$  are covered the same number of times. If this number is 1, then  $f$ , restricted to the open subset  $W' = f^{-1}(W)$  of  $\mathcal{K}$ , is injective.*

### 5.2.2 Properties of $\pi_p^*$

Let  $\sigma = [p, p_1, \dots, p_k]$  be a  $k$ -simplex of  $\text{st}(p, \hat{\mathcal{M}})$ . Now, for all points  $q \in \sigma$ , let  $\pi_{p,t}(q) = (1-t)q + t\pi_p(q)$ , and  $\sigma_t = \pi_{p,t}(\sigma)$ . Since  $\pi_p$  is affine on each simplex,  $\pi_{p,t}$  is also affine. Therefore  $\sigma_t$  is a simplex with vertices  $p, p_{1t}, \dots, p_{kt}$  with  $p_{jt} = \pi_{p,t}(p_j)$ ,  $j \in \{1, \dots, k\}$ .

It directly follows from Lemma 5.10 that

$$\forall j \in \{1, \dots, k\}, \|p_j - p_{jt}\| = O(\varepsilon^2) \text{ lfs}(p). \quad (23)$$

Let  $\tau$  be an  $i$ -simplex, and let  $f : \tau \rightarrow \mathcal{M}$  be a  $C^1$ -function. The simplex  $\tau$  is said to be  *$C^1$ -embedded by  $f$  in  $\mathcal{M}$*  if  $f$  is an injective mapping and for all  $x \in \tau$ , the rank of the linear map  $df(x) : \mathbb{R}^i \rightarrow T_{f(x)}$  is  $i$ , where  $T_{f(x)}$  is the tangent space to  $\mathcal{M}$  at  $f(x)$ .

**Lemma 5.15 ( $\pi$   $C^1$ -embeds  $\sigma_t$ )** *Assume condition **C1** and **C2** of Theorem 5.3. If  $\varepsilon$  is sufficiently small, then the map  $\pi$   $C^1$ -embeds  $\sigma_t$  in  $\mathcal{M}$ , for all  $t \in [0, 1]$ .*

**Proof.** By Lemma 4.5 (1) and the definition of  $\sigma_t$ , we have  $\sigma_t \subset B(p, O(\varepsilon) \text{lfs}(p))$ . Therefore, for a sufficiently small  $\varepsilon$ ,  $\sigma_t$  cannot intersect the medial axis of  $\mathcal{M}$ , which implies, from Lemma 5.2 (1), that the restriction of the map  $\pi$  to  $\sigma_t$  is a  $C^1$ -function. The rest of the proof is devoted to showing that the restriction of  $\pi$  to  $\sigma_t$  is injective and that  $d\pi$  is nonsingular.

Let  $\rho \stackrel{\text{def}}{=} \max_{i \in \{1, \dots, k\}} (\|q_i - p_i\|) / \Delta_\sigma$ . Since  $\Delta_\sigma \geq \delta \text{lfs}(p)$  and  $\varepsilon/\delta \leq \eta_0$ ,  $\rho = O(\varepsilon^2) \text{lfs}(p) / \Delta_\sigma = O(\varepsilon)$ . It then follows from Lemma 5.11 and the fact that  $\Theta_\sigma \geq \Theta_0$ , that  $\Theta_{\sigma_t} \geq \Theta_0/2$  for  $\varepsilon$  sufficiently small. Using  $\Gamma_\sigma \leq C_3$  (Lemma 4.5 (2)), we get for sufficiently small  $\varepsilon$

$$\begin{aligned} L_{\sigma_t} &\geq L_\sigma - 2\rho\Delta_\sigma \\ &= L_\sigma(1 - 2\rho\Gamma_\sigma) \\ &\geq C_3\delta \text{lfs}(p)(1 - O(\varepsilon)) \\ &\geq C_3\delta \text{lfs}(p)/2 \end{aligned}$$

We then deduce from Corollary 2.5 and the bounds on  $\Theta_{\sigma_t}$  and  $L_{\sigma_t}$

$$\sin \angle(\text{aff}(\sigma), \text{aff}(\sigma_t)) = \frac{O(\varepsilon^2) \text{lfs}(p)}{\Theta_{\sigma_t}^k L_{\sigma_t}} = O(\varepsilon)$$

and from Lemma 4.7

$$\sin \angle(T_p, \text{aff}(\sigma_t)) \leq \sin \angle(T_p, \text{aff}(\sigma)) + \sin \angle(\text{aff}(\sigma), \text{aff}(\sigma_t)) = O(\varepsilon). \quad (24)$$

Let  $x$  be a point of  $\sigma$ . We have, from Lemmas 4.5 (1) and 5.10,  $\|p - \pi(x)\| = O(\varepsilon) \text{lfs}(p)$ . Therefore, from Lemma 5.8 and Eq. (24),

$$\sin \angle(T_{\pi(x)}, \text{aff}(\sigma_t)) \leq \sin \angle(T_{\pi(x)}, T_p) + \sin \angle(T_p, \text{aff}(\sigma_t)) = O(\varepsilon). \quad (25)$$

Eq. (25) implies that  $\pi$  restricted to  $\sigma_t$  is injective for  $\varepsilon$  sufficiently small. Indeed otherwise, there would exist  $x_1, x_2 \in \sigma_t$  such that  $\pi(x_1) = \pi(x_2)$ , and the line segment in  $\text{aff}(\sigma_t)$  joining the points  $x_1$  and  $x_2$  would be parallel to  $N_{\pi(x_1)}$ , contradicting Eq. (25).

We deduce also from Eq. (25) and Lemma 5.2 (2) that the derivative  $d\pi$  is non-singular when  $\pi$  is restricted to  $\sigma_t$ . This completes the proof of the lemma.  $\square$

**Lemma 5.16** ( $\pi^*$   $C^1$ -embeds  $\sigma_t$ ) *For  $\varepsilon$  sufficiently small,  $\pi_p^*$   $C^1$ -embeds  $\sigma_t$  in  $T_p$  for all  $t \in [0, 1]$ .*

**Proof.** From the definition of  $\sigma_t$ , for all  $x \in \sigma_t$ , we have  $\text{dist}(x, \sigma) = O(\varepsilon^2) \text{lfs}(p)$ . From Lemma 5.10, we have  $\|x - \pi(x)\| = O(\varepsilon^2) \text{lfs}(p)$  for all  $x \in \sigma$ . Combining these facts we get  $\|x - \pi(x)\| = O(\varepsilon^2) \text{lfs}(p)$  for all  $x \in \sigma_t$ . This implies, from Lemma 4.5 (1),

$$\pi(\sigma_t) \subset B_{\mathcal{M}}(p, O(\varepsilon) \text{lfs}(p)) \subset \widetilde{\mathcal{M}}_p$$

From Lemma 5.8 and the fact that  $\pi(\sigma_t) \subset B_{\mathcal{M}}(p, O(\varepsilon) \text{lfs}(p))$ , we have

$$\sin \angle(T_x, T_p) < 1, \quad \forall x \in \pi(\sigma_t) \quad (26)$$

Since for all  $x \in \pi(\sigma_t)$ ,  $\sin \angle(T_x, T_p) < 1$ , we get from Lemma 5.9,

$$N_x \cap T_p = \widetilde{\pi}_p^{-1}(x), \quad \forall x \in \pi(\sigma_t)$$

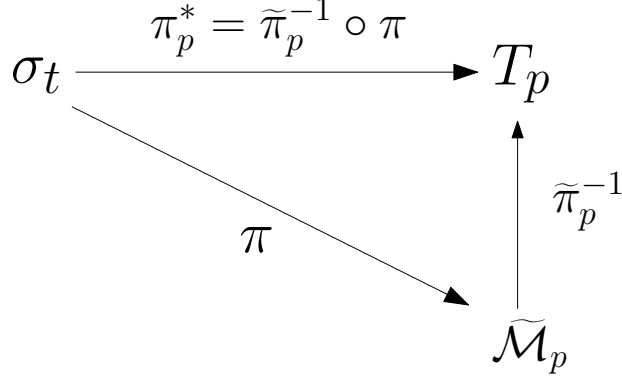


Figure 8: Refer to the proof of the Lemma 5.16.

Therefore, we deduce  $\pi_p^*|_{\sigma_t} = \pi_p^{-1} \circ \pi$ . Refer to Fig. 8.

Then, by Lemma 5.9,  $\pi_p^{-1}$  is a diffeomorphism from  $\pi_p(\widetilde{T}_p) \supseteq \widetilde{\mathcal{M}}_p$  to  $\widetilde{T}_p$ , and, by Lemma 5.15,  $\pi$   $C^1$ -embeds  $\sigma_t$  into  $\mathcal{M}$ . We conclude that  $\pi_p^*|_{\sigma_t} = \pi_p^{-1} \circ \pi$   $C^1$ -embeds  $\sigma_t$  into  $T_p$ .  $\square$

The proof of the following lemma is similar to Lemma 23a in Chapter 4 from [Whi57] and it will be used for proving that  $\pi_p^*$  restricted to open star  $\mathring{\text{st}}(p, \hat{\mathcal{M}})$  is injective in Lemma 5.18. We give a proof for completeness.

The proof of the following lemma is similar to Lemma 23a in Chapter 4 from [Whi57]. We give a proof for completeness.

**Lemma 5.17** *Assume conditions **C1** and **C2** of Theorem 5.3. For  $\varepsilon$  sufficiently small,  $\pi_p^*$  is a simplexwise positive mapping of  $\text{st}(p, \hat{\mathcal{M}})$  into  $T_p$ .*

**Proof.** We use the same notation  $\sigma_t$  as in the beginning of this section. We first observe that, since the simplex  $\sigma_1$  is in  $T_p$ ,  $\pi_p^*$  is the identity on  $\sigma_1$ . Therefore,  $\det(J(\pi_p^*)) > 0$  on  $\sigma_1$ . Moreover, since from Lemma 5.16,  $\det(J(\pi_p^*)) \neq 0$  in  $\sigma_t$  for all  $t \in [0, 1]$ , we also have  $\det(J(\pi_p^*)) > 0$  in  $\sigma_0$ . Which proves the lemma.  $\square$

We are now ready to prove that  $\pi_p^*$  restricted to the open star  $\mathring{\text{st}}(p, \hat{\mathcal{M}})$  is injective.

**Lemma 5.18 (Injectivity of  $\pi_p^*$  restricted to  $\mathring{\text{st}}(p, \hat{\mathcal{M}})$ )** *Assume conditions **C1** and **C2** of Theorem 5.3 and let  $\varepsilon$  be sufficiently small. For each point  $p \in \mathcal{P}$ , the map  $\pi_p^*$  is injective on the open star  $\mathring{\text{st}}(p, \hat{\mathcal{M}})$ .*

**Proof.** Note that, since  $\hat{\mathcal{M}}$  is a PL-manifold without boundary (Lemma 5.5),  $\text{lk}(p, \hat{\mathcal{M}})$  is a PL  $(k-1)$ -sphere,  $\text{st}(p, \hat{\mathcal{M}})$  is a PL  $k$ -ball, and  $\partial \text{st}(p, \hat{\mathcal{M}}) = \text{lk}(p, \hat{\mathcal{M}})$ .

For convenience, rewrite  $f = \pi_p^*$  and  $S = \text{st}(p, \hat{\mathcal{M}})$  and orient  $S$  as indicated in Section 5.2.1.  $S$  is thus an oriented PL  $k$ -ball, and, by Lemma 5.17,  $f = \pi_p^*$  is a simplexwise positive mapping from  $S$  to  $T_p$ . Let  $f(S^{k-1})$  be the image by  $f$  of the  $(k-1)$ -skeleton of  $S$  (i.e. the set of faces of  $S$  of dimension at most  $k-1$ ) and let  $W$  be any connected open subset of  $T_p \setminus f(\partial S)$ . From Lemma 5.14, any two points of  $W \setminus f(S^{k-1})$  are covered the same number of times. If this number is 1, then  $f$ , restricted to the open subset  $f^{-1}(W)$  of  $\text{st}(p)$ , is injective onto  $W$ .

To prove the lemma, it is therefore sufficient to prove that there exists a point  $g$  in  $S \setminus S^{k-1}$  whose image  $f(g)$  is not covered by any other point  $x$  of  $S \setminus S^{k-1}$ , i.e.  $f(g) \neq f(x)$  for all  $x \neq g$  and  $x \in S \setminus S^{k-1}$ . Let  $\sigma$  be a  $k$ -simplex  $[q_0, \dots, q_k]$  of  $S$  and let

$$g = \frac{1}{k+1} \sum_{i=0}^k q_i.$$

Using Lemma 5.12 and the facts that  $\Theta_\sigma \geq \Theta_0$  (since the simplices of  $\hat{\mathcal{M}}$  are  $\Theta_0$ -fat),  $\Delta_\sigma \geq \delta \text{lfs}(p)$ , and  $\varepsilon/\delta \leq \eta_0$  (Hypothesis 3.1), we have

$$\text{dist}(g, \partial\sigma) \geq \frac{k! \Theta_\sigma^k \Delta_\sigma}{k+1} \geq \frac{k! \Theta_0^k}{k+1} \times \delta \text{lfs}(p) \geq \frac{k! \Theta_0^k}{k+1} \times \frac{\varepsilon \text{lfs}(p)}{\eta_0} = \Omega(\varepsilon) \text{lfs}(p).$$

From the above equation and using the fact that  $\|\pi_p(x) - x\| = O(\varepsilon^2) \text{lfs}(p)$  for all  $x \in \sigma$  (Lemma 5.10), we have for  $\varepsilon$  sufficiently small

$$\text{dist}(\pi_p(g), \partial\pi_p(\sigma)) = \Omega(\varepsilon) \text{lfs}(p) - O(\varepsilon^2) \text{lfs}(p) = \Omega(\varepsilon) \text{lfs}(p). \quad (27)$$

Since  $\pi_p$  embeds  $S$  into  $T_p$  (from Lemma 2.2) with  $\pi_p(S) = \text{st}(p, \text{Del}^{\xi_p}(\mathcal{P}_p))$  and since  $\text{Del}^{\xi_p}(\mathcal{P}_p)$  is a triangulation of a  $k$ -dimensional convex hull (see the discussion on the general position assumption at the beginning of Section 5), Eq. (27) implies that, for all  $x \in S \setminus \text{int } \sigma$

$$\|\pi_p(g) - \pi_p(x)\| = \Omega(\varepsilon) \text{lfs}(p) - O(\varepsilon^2) \text{lfs}(p). \quad (28)$$

From Eq. (28), we have for all  $x \in S \setminus \text{int } \sigma$

$$\begin{aligned} \|\pi_p^*(x) - \pi_p^*(g)\| &\geq \|\pi_p(x) - \pi_p(g)\| - (\|\pi_p(x) - x\| + \|\pi_p^*(x) - x\|) \\ &\quad - (\|\pi_p(g) - g\| + \|\pi_p^*(g) - g\|) \\ &= \Omega(\varepsilon) \text{lfs}(p) - O(\varepsilon^2) \text{lfs}(p). \end{aligned} \quad (29)$$

which is positive for a sufficiently small  $\varepsilon$ .

Lemma 5.16 together with Eq. (29) shows that  $\pi_p^*(x) \neq \pi_p^*(g)$  for all  $x \in S \setminus \{g\}$ . Hence,  $g$  is not covered by any other point of  $S$ , and the lemma follows.  $\square$

### 5.2.3 Properties of $\pi$

**Lemma 5.19 (Injectivity of the restriction of  $\pi$  to  $\text{st}(p, \hat{\mathcal{M}})$ )** *Assume conditions **C1** and **C2** of Theorem 5.3. Let  $\varepsilon$  be sufficiently small. For all  $p$  in  $\mathcal{P}$ , the map  $\pi$  restricted to  $\text{st}(p, \hat{\mathcal{M}})$  is injective.*

**Proof.** To reach a contradiction, assume that there exist  $x_1, x_2$  ( $x_1 \neq x_2$ ) in  $\text{st}(p) \setminus \partial \text{st}(p)$  such that  $\pi(x_1) = \pi(x_2)$ . Then  $\pi_p^*(x_1) = \pi_p^*(x_2) = N_{\pi(x_1)} \cap T_p$ . Which contradicts the fact that  $\pi_p^*$  is injective when restricted to  $\text{st}(p, \hat{\mathcal{M}})$  (Lemma 5.18).  $\square$

**Lemma 5.20 (Surjectivity of  $\pi|_{\hat{\mathcal{M}}}$ )** *Assume conditions **C1** and **C2** of Theorem 5.3. If  $\varepsilon$  is sufficiently small, then  $\pi|_{\hat{\mathcal{M}}}$  is surjective on  $\mathcal{M}$ .*

**Proof.** Since  $\hat{\mathcal{M}}$  and  $\mathcal{M}$  are topological  $k$ -manifolds without boundaries, and  $\mathring{\text{st}}(p, \hat{\mathcal{M}})$ , which is an open set in  $\hat{\mathcal{M}}$  with  $\hat{\mathcal{M}} = \bigcup_{p \in \mathcal{P}} \mathring{\text{st}}(p, \hat{\mathcal{M}})$ , is mapped injectively by  $\pi$  into  $\mathcal{M}$ , we get from the Generalized Invariance of Domain Theorem,  $\pi$  is an open map and  $\pi(\hat{\mathcal{M}})$  is an open set in  $\mathcal{M}$ . Also, note that since  $\hat{\mathcal{M}}$  is compact and  $\pi$  is a continuous function,  $\pi(\hat{\mathcal{M}})$  is a compact subset of  $\mathcal{M}$  [Bre94, Thm. 7.6 of Chap. 1]. As  $\mathcal{M}$  is a Hausdorff space, this implies that  $\pi(\hat{\mathcal{M}})$  is a closed set of  $\mathcal{M}$  [Bre94, Thm. 7.5 of Chap. 1].

Now consider a connected component  $\mathcal{M}_\alpha$  of  $\mathcal{M}$  with  $\mathcal{V}_\alpha = \pi(\hat{\mathcal{M}}) \cap \mathcal{M}_\alpha \neq \emptyset$ . From the discussion in the previous paragraph we have  $\mathcal{V}_\alpha$  is both open and closed in  $\mathcal{M}_\alpha$ . Since  $\mathcal{M}_\alpha$  is connected, this implies  $\mathcal{V}_\alpha = \mathcal{M}_\alpha$  [Bre94, Def. 4.1 of Chap. 1]. To complete the proof of surjectivity, we only need to show that  $\mathcal{P}$  has non empty intersection with each connected components of  $\mathcal{M}$ , since  $\pi(p) = p$  for all  $p \in \mathcal{P}$ . By using the proof technique of Giesen and Wagner [GW04, Lem. 3], we can show that if  $\mathcal{P}$  is an  $(\varepsilon, \delta)$ -sample of  $\mathcal{M}$  with  $\varepsilon < 1$ , then  $\mathcal{P}$  intersects all the connected components of  $\mathcal{M}$ . See, [Gho12, Chap. 4] for details. This completes the proof.  $\square$

As discussed earlier,  $\pi$  is an open map  $\pi|_{\mathring{\text{st}}(p, \hat{\mathcal{M}})}$  is injective (Lemma 5.19) and  $\mathring{\text{st}}(p, \hat{\mathcal{M}})$  is an open set in  $\hat{\mathcal{M}}$ . The restriction of  $\pi$  to  $\mathring{\text{st}}(p, \hat{\mathcal{M}})$  is thus a homeomorphism onto its image. In addition, we know from Lemma 5.20 that  $\pi(\hat{\mathcal{M}}) = \mathcal{M}$ . Therefore, for all  $x \in \mathcal{M}$ , there exists open neighborhoods  $U_x$  of  $x$  in  $\mathcal{M}$  such that  $\pi^{-1}(U_x)$  is a disjoint union of open sets, each of which is contained in an open star of a vertex in  $\hat{\mathcal{M}}$  and mapped homeomorphically onto  $U_x$  by  $\pi$ . In other words,  $(\hat{\mathcal{M}}, \pi)$  is a covering space of  $\mathcal{M}$ .

A standard result in topology [Mas67, Lem. 3.4 of Chap. 5] asserts that  $\pi$  covers all the points of a connected component of its image  $\mathcal{M}$  the same number of times. Let us show that this number is exactly one. It cannot be zero by definition. Neither can it be more than one since, otherwise, some sample point  $p$  would be covered more than once, which would contradict Lemma A.1 from Appendix A. We have therefore proved that the restriction of  $\pi$  to  $\hat{\mathcal{M}}$  is injective.

**Lemma 5.21 (Injectivity of  $\pi|_{\hat{\mathcal{M}}}$ )** *Under the same conditions as in Lemma 5.21,  $\pi$  restricted to  $\hat{\mathcal{M}}$  is injective.*

Since  $\hat{\mathcal{M}}$  is compact and  $\mathcal{M}$  is a Hausdorff space, Lemmas 5.20 and 5.21, and a standard result from topology [Bre94, Thm. 7.8 of Chap. 1] implies that  $\pi : \hat{\mathcal{M}} \rightarrow \mathcal{M}$  is a homeomorphism.

This completes the proof of Lemma 5.6.

### 5.3 Pointwise approximation

Following lemma is a direct consequence is Lemmas 5.10 and 5.6, and the fact that  $\text{lfs}$  is 1-Lipschitz function. For the details of the proof, refer to [Gho12, Chap. 4].

**Lemma 5.22 (Pointwise approximation)** *Under conditions **C1** and **C2** of Theorem 5.3, and  $\varepsilon$  sufficiently, we have  $\text{dist}(x, \pi|_{\hat{\mathcal{M}}}^{-1}(x)) = O(\varepsilon^2) \text{lfs}(x)$ .*

### 5.4 Isotopy

We need the following lemma to show the isotopy property **P5** of Theorem 5.3. The proof of the lemma follows directly from [GW04, Lem. 5] and one can check the details in [Gho12, Chap. 4].

**Lemma 5.23** *Assume conditions **C1** and **C2** of Theorem 5.3. If  $\varepsilon$  is sufficiently small, for all points  $x, y$  ( $\neq p$ ) in  $\hat{\mathcal{M}}$ ,*

$$[x \pi(x)] \cap [y \pi(y)] = \emptyset$$

Consider the following map:

$$F : \hat{\mathcal{M}} \times [0, 1] \rightarrow \mathbb{R}^d, \text{ with } F(x, t) = x + t(\pi(x) - x).$$

We will denote by  $F_t = F(\cdot, t)$  and  $M_t = F_t(\hat{\mathcal{M}})$  for all  $t \in [0, 1]$ . It is easy to see that  $F(x, t)$  (and  $F_t$ ) is continuous,  $M_0 = \hat{\mathcal{M}}$ , and since  $\pi$  restricted to  $\hat{\mathcal{M}}$  gives a homeomorphism between  $\hat{\mathcal{M}}$  and  $\mathcal{M}$ ,  $M_1 = \mathcal{M}$ .

Using the facts that  $\hat{\mathcal{M}}$  is compact,  $M_t$  is a Hausdorff space (as  $M_t$  is a subspace with the subspace topology of  $\mathbb{R}^d$ ), and  $F_t$  is continuous, we get [Bre94, Thm. 7.8 of Chap. 1], the following result:

**Lemma 5.24** *Assume conditions **C1** and **C2** of Theorem 5.3. If  $\varepsilon$  is sufficiently small, then for all  $t \in [0, 1]$ ,  $F_t : \hat{\mathcal{M}} \rightarrow M_t$  is a homeomorphism.*

This implies:

**Lemma 5.25 (Isotopy)** *Assume conditions **C1** and **C2** of Theorem 5.3. If  $\varepsilon$  is sufficiently small then*

$$F : \hat{\mathcal{M}} \times [0, 1] \longrightarrow \mathbb{R}^d$$

*is an isotopy.*

## 6 Conclusion

We have given the first algorithm that is able to reconstruct a smooth closed manifold in a time that depends only linearly on the dimension of the ambient space. We believe that our algorithm is of practical interest when the dimension of the manifold is small, even if it is embedded in a space of high dimension. This situation is quite common in practical applications in machine learning. Unlike most surface reconstruction algorithms in  $\mathbb{R}^3$ , our algorithm does not need to orient normals (a critical issue in practical applications) and, in fact, works for non orientable manifolds.

The algorithm is simple. The basic ingredients we need are data structures for constructing weighted Delaunay triangulations in  $\mathbb{R}^k$ . We have assumed that the dimension of  $\mathcal{M}$  is known. If not, we can use algorithms given in [CWW08, GW04] to estimate the dimension of  $\mathcal{M}$  and the tangent space at each sample point. Moreover, our algorithm is easy to parallelize. One interesting feature of our approach is that it is robust and still works if we only have approximate tangent spaces at the sample points. We will report on experimental results in a forthcoming paper.

We have assumed that we know an upper bound on the sampling ratio  $\eta_0$  of the input sample  $\mathcal{P}$ . Ideas from [BGO09, FR02] may be useful to convert a sample to a subsample with a bounded sampling ratio.

We foresee other applications of the tangential complex and of our construction each time computations in the tangent space of a manifold are required, e.g. for dimensionality reduction and approximating the Laplace Beltrami operator [BSW08]. It easily follows from [BNN10] that our reconstruction algorithm can also be used in Bregman spaces where the Euclidean distance is replaced by any Bregman divergence, e.g. Kullback-Leibler divergence. This is of particular interest when considering statistical manifolds like, for example, spaces of images [CIdSZ08].

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## A Appendix to Section 5.2

This section is devoted to the proof of the following lemma.

**Lemma A.1** *Let  $\varepsilon$  be sufficiently small. For all  $p$  in  $\mathcal{P}$  and restricting the map  $\pi$  to  $\hat{\mathcal{M}}$ , we have  $\pi^{-1}(p) = \{p\}$ .*

**Proof.** To reach a contradiction, we assume that there exists a  $k$ -simplex  $\tau = [q_0, \dots, q_k]$  in  $\hat{\mathcal{M}}$  such that there exists  $x \in \tau$  with  $x \neq p$  and  $\pi(x) = p$ .

We will have to consider the following two cases.

**Case 1.**  $p$  is a vertex of  $\tau$ . This implies that the unit vector  $u \in \text{aff}(\tau)$  along the line joining the points  $p$  and  $x$  lies in  $N_p$ . But from Lemma 4.7 and  $\varepsilon$  sufficiently small, we have  $\sin \angle(\text{aff}(\tau), T_p) \leq A\varepsilon < 1$ .

**Case 2.**  $p$  is not a vertex of  $\tau$ . The outline of the proof for case 2 is the following: we will divide the  $k$ -simplex  $\tau$  into a union of  $k + 2$  convex cells and show that for each convex cell the distance of any point of the cell to  $p$  is  $\Omega(\varepsilon \text{lfs}(q))$ , where  $q$  is a vertex of  $\tau$ . For  $\varepsilon$  sufficiently small, we will reach a contradiction from Lemma 5.10.

Since  $\hat{\mathcal{M}}$  has no inconsistent configuration,  $\tau$  is in  $\text{st}(q_0)$ . Let  $c = \text{Vor}^\omega(\tau) \cap T_{q_0}$ . From Lemma 4.4, we have

$$R = \sqrt{\|c - q_0\|^2 - \omega(q_0)^2} \leq \|c - q_0\| \leq C_1 \varepsilon \text{lfs}(q_0). \quad (30)$$

Using the facts that for all vertices  $q_i, q_j$  of  $\tau$ ,  $\|q_i - q_j\| \leq C_2 \varepsilon \text{lfs}(q_i)$  (from Lemma 4.5 (1)),  $\text{lfs}$  is 1-Lipschitz and  $\varepsilon$  sufficiently small, we have

$$\text{lfs}(q_j) \geq \text{lfs}(q_i) - \|q_i - q_j\| \geq (1 - C_2 \varepsilon) \text{lfs}(q_i) \geq \frac{\text{lfs}(q_i)}{2} \quad (31)$$

Therefore using Eq. (30) and (31), we have for all vertices  $q_i$  of  $\tau$

$$R \leq C_1 \varepsilon \text{lfs}(q_0) \leq 2C_1 \varepsilon \text{lfs}(q_i). \quad (32)$$

Consider the edge  $q_i q_j$  of  $\tau$  and let  $c_{ij}$  be the projection of  $c$  onto the line segment  $[q_i q_j]$ . Observe that the ball of radius  $r_{ij} = \sqrt{\|c_{ij} - q_i\|^2 - \omega(q_i)^2} \leq R$  centered at  $c_{ij}$  is orthogonal to the balls  $B(q_i, \omega(q_i))$  and  $B(q_j, \omega(q_j))$ . Using the fact that  $\mathcal{P}$  is a  $(\varepsilon, \delta)$ -sample of  $\mathcal{M}$  and Lemma 4.1 (2), we have

$$r_{ij} \geq \frac{\sqrt{1 - 4\omega_0^2} \|q_i - q_j\|}{2} \geq b\delta \max\{\text{lfs}(q_i), \text{lfs}(q_j)\} \quad (33)$$

where  $b = \frac{\sqrt{1 - 4\omega_0^2}}{2}$ .



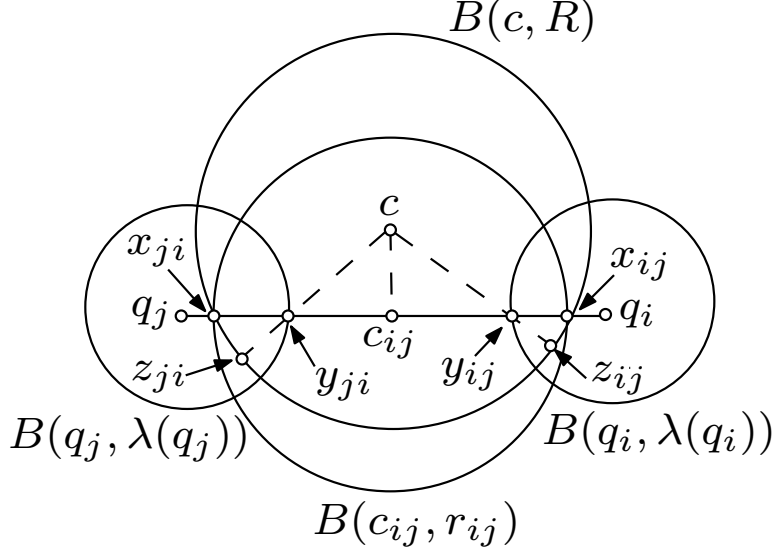


Figure 9: Refer to the proof of Lemma A.1, Case 2.

Let  $\lambda(q_i) = \max\{\omega(q_i), b\delta \text{lfs}(q_i)/4\}$  for  $q_i \in \{q_0, \dots, q_k\}$ . Observe that  $\lambda(q_i) < \mathfrak{m}(q_i)/2$ . For  $\varepsilon \leq 1/2$  and Lemma 2.3 (1),

$$\lambda(q_i) < \frac{\mathfrak{m}(q_i)}{2} \leq \frac{\varepsilon}{1-\varepsilon} \text{lfs}(q_i) \leq 2\varepsilon \text{lfs}(q_i). \quad (34)$$

Let  $x_{ij} = [c_{ij} q_i] \cap \partial B(c_{ij}, r_{ij})$  and  $y_{ij} = [c_{ij} q_i] \cap \partial B(q_i, \lambda(q_i))$ . Note that  $x_{ij} = \partial B(c, R) \cap [c_{ij} q_i]$ , see Figure 9. Therefore

$$\begin{aligned} \|x_{ij} - y_{ij}\| &= r_{ij} + \lambda(q_i) - \sqrt{r_{ij}^2 + \omega(q_i)^2} \\ &= \frac{2r_{ij}\lambda(q_i) + \lambda(q_i)^2 - \omega(q_i)^2}{r_{ij} + \lambda(q_i) + \sqrt{r_{ij}^2 + \omega(q_i)^2}} \\ &\geq \frac{2r_{ij}\lambda(q_i)}{R + \lambda(q_i) + \sqrt{R^2 + \lambda(q_i)^2}} && \text{since } \lambda(q_i) \geq \omega(q_i) \text{ and } r_{ij} \leq R \\ &\geq \frac{2r_{ij}\lambda(q_i)}{\left(2C_1\varepsilon + 2\varepsilon + \sqrt{4C_1^2\varepsilon^2 + 4\varepsilon^2}\right) \text{lfs}(q_i)} && \text{from Eq. (32) and (34)} \\ &\geq \frac{b^2\delta^2 \text{lfs}(q_i)}{4\varepsilon(C_1 + 1 + \sqrt{C_1^2 + 1})} && \text{using } \lambda(q_i) \geq \frac{b\delta \text{lfs}(q_i)}{4} \text{ and Eq. (33)} \\ &\geq \frac{b^2\delta \text{lfs}(q_i)}{4\eta_0(C_1 + 1 + \sqrt{C_1^2 + 1})} && \text{since } \varepsilon/\delta \leq \eta_0 \text{ from Hypothesis 3.1} \end{aligned}$$

Using the fact that  $r_{ij} \geq \|x_{ij} - y_{ij}\|$ , we have

$$A_{ij}^2 \stackrel{\text{def}}{=} (2r_{ij} - \|x_{ij} - y_{ij}\|) \times \|x_{ij} - y_{ij}\| \geq r_{ij} \times \|x_{ij} - y_{ij}\| \geq C_6^2 \delta^2 \text{lfs}(q_i)^2$$

where  $C_6$  is a constant that depends on  $\omega_0$  and  $\eta_0$ .

Let  $z_{ij}$  denote the point closest to  $y_{ij}$  on  $\partial B(c, R)$ .

From the Intersecting Chords Theorem (see Figure 10) for circles, we have :

$$(2R - \|z_{ij} - y_{ij}\|) \times \|z_{ij} - y_{ij}\| = A_{ij}^2.$$

By the definition of  $z_{ij}$ , the solution to the above quadratic equation in  $\|z_{ij} - y_{ij}\|$  is the smaller root:

$$\begin{aligned} \|z_{ij} - y_{ij}\| &= R - \sqrt{R^2 - A_{ij}^2} \\ &= \frac{A_{ij}^2}{R + \sqrt{R^2 - A_{ij}^2}} \geq \frac{A_{ij}^2}{2R} \geq \frac{C_6^2}{4C_1\eta_0^2} \varepsilon \text{ lfs}(q_i) \stackrel{\text{def}}{=} 2C_7\varepsilon \text{ lfs}(q_i) \end{aligned}$$

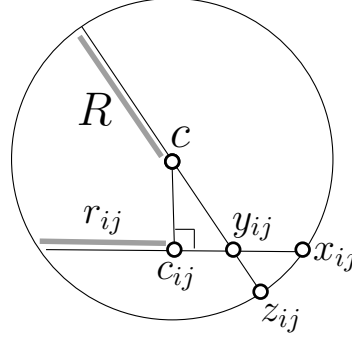


Figure 10: Intersecting Chords Theorem.

The last inequality follows from the facts that  $R \leq 2C_1\varepsilon \text{ lfs}(q_i)$  (from Eq. (32)) and  $\varepsilon/\delta \leq \eta_0$ .

Using the fact that  $\varepsilon$  is sufficiently small and Eq. (31), we have for all vertices  $q$  of  $\tau$

$$\|z_{ij} - y_{ij}\| \geq 2C_7\varepsilon \text{ lfs}(q_i) \geq C_7\varepsilon \text{ lfs}(q)$$

Let  $\text{conv}(S)$  denote the convex hull of the points  $y_{ij}$  :

$$S = \{y_{ij} \mid i, j (\neq i) \in \{0, \dots, k\}\}.$$

Using convexity, we have

$$\text{dist}(\text{conv}(S), \partial B(c, R)) = \min_{i, j (\neq i) \in \{0, \dots, k\}} \|z_{ij} - y_{ij}\| \geq C_7\varepsilon \text{ lfs}(q) \quad (35)$$

for all vertices  $q \in \{q_0, \dots, q_k\}$ .

Recall that we have assumed in the beginning of the proof that  $x \in \tau$  and  $\pi(x) = p$ . Eq. (35) implies that  $x \notin \text{conv}(S)$ . Indeed, if  $x \in \text{conv}(S)$ , then from Eq. (35) and the fact that the ball  $B(c, R)$  is empty, we will have

$$\|x - p\| \geq C_7\varepsilon \text{ lfs}(q)$$

for all vertices  $q$  of  $\tau$ . But from Lemma 5.10, we have

$$\|x - p\| = \|x - \pi(x)\| \leq C\varepsilon^2 \text{ lfs}(q)$$

for all vertices  $q$  of  $\tau$ . We have reached a contradiction for  $\varepsilon$  sufficiently small. Hence  $x \notin \text{conv}(S)$ .

Let  $S_i = \{y_{ij} \mid j \in \{0, \dots, k\} \setminus \{i\}\} \cup \{q_i\}$ . The convex hulls of  $S_i$ ,  $\text{conv}(S_i)$ , satisfies the following properties :

$$\begin{aligned} \text{conv}(S_i) &\subset B(q_i, \lambda(q_i)) \\ \tau &= \text{conv}(S) \cup \left( \bigcup_{i \in \{0, \dots, k\}} \text{conv}(S_i) \right) \end{aligned} \quad (36)$$

If  $x \in \text{conv}(S_i)$ , then from Lemma 5.10, we have for  $\varepsilon$  sufficiently small

$$\begin{aligned}
\|q_i - p\| &\leq \|q_i - x\| + C\varepsilon^2 \text{lfs}(q_i) \\
&\leq \lambda(q_i) + C\varepsilon^2 \text{lfs}(q_i) \\
&< \frac{\mathfrak{m}(q_i)}{2} + C\varepsilon^2 \text{lfs}(q_i) && \text{since } \lambda(q_i) < \mathfrak{m}(q_i)/2 \\
&< \mathfrak{m}(q_i)
\end{aligned}$$

The last inequality holds for  $\varepsilon < \frac{1}{2C\eta_0}$ . We have reached a contradiction.

Since  $x \notin \text{conv}(S_i)$  and  $x \notin \text{conv}(S)$ , it follows that  $x \notin \tau$ . We have therefore reached a contradiction, i.e. there does not exist  $x \in \tau$  such that  $\pi(x) = p$ .  $\square$

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## Main notations

### General notations (Section 2)

$B(c, r)$	is $\{x \in \mathbb{R}^d \mid \ c - x\  < r\}$
$\bar{B}(c, r)$	is $\{x \in \mathbb{R}^d \mid \ c - x\  \leq r\}$
$\dim(U)$	dimension of the affine space $U$
$\text{dist}(p, X)$	is $\min_{x \in X} \ p - x\ $
$\partial X$	topological boundary of $X$
$\text{int } X$	topological interior of $X$
$df(x)$	derivative of the function $f$ at $x$
$f _{X_1}$	map $f$ restricted to the subset $X_1$
$[xy]$	line segment connecting points $x$ and $y$
$\ x - y\ $	standard Euclidean distance between the points $x$ and $y$
$\angle(u, v)$	angle between the vectors $u$ and $v$
$\angle(U, V)$	angle between the affine spaces $U$ and $V$

## Submanifold (Section 2)

$B_{\mathcal{M}}(p, r)$	is $B(p, r) \cap \mathcal{M}$
$\bar{B}_{\mathcal{M}}(p, r)$	is $\bar{B}(p, r) \cap \mathcal{M}$
$d$	dimension of the ambient space
$k$	intrinsic dimension of $\mathcal{M}$
$\mathcal{M}$	manifold
$\mathcal{P}$	sample
$ \mathcal{P} $	cardinality of the sample $\mathcal{P}$ (Theorem 4.16, page 25)
$T_x$	tangent space at $x$
$N_x$	normal space at $x$

## Weights (Section 2.1.1)

$p^\omega$	weighted point $(p, \omega(p))$
$\omega$	weight assignment
$\tilde{\omega}$	relative amplitude
$\omega_0$	bound on the relative amplitude

## Sampling (Section 2.2)

$\varepsilon$	sampling rate
$\varepsilon/\delta$	sampling ratio
$\eta_0$	bound on the sampling ratio
$\delta$	sparsity
$\text{lfs}(p)$	local feature size at $p$
$\text{m}(p)$	distance of $p$ to its nearest neighbor
$N$	is $(4C_2''\eta_0 + 6)^k$ (Section 4.4, page 23)
$LN(p)$	local neighborhood of $p$ (Section 3)

### Shape measure of simplex $\tau$ (Section 2.3)

$c_\tau$	circumcenter
$D_\tau(p)$	altitude (Section 4.1, page 15)
$\text{aff}(\tau)$	affine hull of $\tau$
$m_u(\tau)$ or $m_u$	see Section 2.4, page 10
$i_\phi$	see Section 2.4, page 11
$N_\tau$	affine space orthogonal to $\text{aff}(\tau)$
$\Delta_\tau$	length of longest edge (diameter)
$H_\tau(p, \omega(p))$	excentricity (Section 4.1, page 15)
$L_\tau$	length of shortest edge
$o_\tau$	orthocenter
$\Phi_\tau$	orthoradius
$R_\tau$	circumradius
$\Gamma_\tau = \frac{\Delta_\tau}{L_\tau}$	longest edge to shortest-edge ratio
$\tau_p = \tau \setminus \{p\}$	face of $\tau$ opposite to $p$ (Section 4.1)
$\Theta_\tau$	fatness
$\Theta_0$	fatness bound
$V_\tau$	volume

### Complexes (Section 2.4)

$\mathcal{K}$	simplicial complex
$\text{st}(p, \mathcal{K})$	$\{\tau : \text{for some } \tau_1 \in \mathcal{K}, \sigma, \tau \subseteq \tau_1\}$
$\text{lk}(p, \mathcal{K})$	$\{\tau \in \text{st}(p, \mathcal{K}) : \tau \cap \sigma = \emptyset\}$
$\mathring{\text{st}}(p, \mathcal{K})$	$\text{st}(p, \mathcal{K}) \setminus \text{lk}(p, \mathcal{K})$
$\text{st}(p)$	$\text{st}(p, \text{Del}_p^\omega(\mathcal{P}))$ (Section 2.4)
$C^\omega(\mathcal{P})$	Complete complex (page 10)
$\text{Del}^\omega(\mathcal{P})$	Weighted Delaunay triangulation (Section 2.1.2, page 5)
$\text{Del}_p^\omega(\mathcal{P})$	restriction of $\text{Del}^\omega(\mathcal{P})$ to $T_p$
$\text{Del}_{T, \mathcal{M}}^\omega(\mathcal{P})$	Tangential Delaunay complex (page 8)
$\text{Inc}^\omega(\mathcal{P})$	contains $\Theta_0$ -inconsistent configurations of $\text{Del}_{T, \mathcal{M}}^\omega(\mathcal{P})$ (page 10)

### Mappings

$\pi$	projection onto $\mathcal{M}$
$\pi _{\hat{\mathcal{M}}}$	restriction of $\pi$ to $\hat{M}$
$\pi_p$	restriction of $\pi$ to $T_p$
$\pi_p^*$	mapping onto $T_p$