Numerical computation of the cut locus via a variational approximation of the distance function

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Abstract

We propose a new method for the numerical computation of the cut locus of a compact submanifold of \( \mathbb{R}^3 \) without boundary. This method is based on a convex variational problem with conic constraints, with proven convergence. We illustrate the versatility of our approach by the approximation of Voronoi cells on embedded surfaces of \( \mathbb{R}^3 \).

1 Introduction

Let \( S \) be a real analytic surface without boundary embedded in \( \mathbb{R}^3 \), and \( b \in S \) any point of \( S \) (that can be thought of as a base point).

Definition 1.1. The cut locus of \( b \) in \( S \) can be defined as the closure of the set of points \( p \in S \) such that there exists at least two minimizing geodesics between \( p \) and \( b \). We will denote it by \( \text{Cut}_b(S) \). Equivalently, it is also the set of points around which the distance function to the point \( b \) - denoted by \( d_b \) - is not smooth.

The cut locus is a fundamental object in Riemannian geometry, and it is a natural problem to try and find ways to compute it numerically. In this paper, we propose a numerical approximation of \( \text{Cut}_b(S) \), based on a convex variational problem on \( S \), with proven convergence. It is not trivial to compute \( \text{Cut}_b(S) \) because it is not stable with respect to \( C^1 \)-small variations of \( S \). See for instance [1, Example 2]. For instance, one can’t approximate the cut locus of \( S \) with the cut locus of a piecewise linear approximation of \( S \).

Related works. Let us review the techniques used in the past by different authors to approximate the cut locus. We may divide them in two categories.

Geodesic approximation on parametrized surfaces. This approach was used in [16] and [13]. In [16], on genus 1 parametrized surfaces, the authors computed a degree 4 polynomial approximation of the exponential map using the geodesic equation, and deduced an approximation of the cut locus from there. In [13], the authors used the deformable simplicial complexes (DSC) method and finite differences techniques for geodesic computations, to compute geodesic circles of
increasing radius and their self intersection, *i.e.* the cut locus. They apply the method to genus 1 surfaces. These papers contain no proof of convergence of the computed cut locus.

**Exact geodesic computation on discretized surfaces.** This approach was used in [11] and [7]. In [11], the authors computed the geodesics on a convex triangulated surface. They deduced an approximation of the cut locus of the triangulated surface, and filtered it according to the angle formed by the geodesics meeting at a point of the approximated cut locus, to make their approximation stable. They applied the method to ellipsoids. There is no proof of convergence. In [7], the authors computed shortest curves on a graph obtained from a sufficiently dense sample of points of a surface. From there they deduced an approximation of the cut locus, and filtered it according to the maximal distance (called *spread*) between the geodesics meeting at a point of the approximated cut locus. They proved that the set they compute converges to the cut locus (see [7, Theorem 4.1]).

We may also mention [3], where the authors use some more geometric tools to compute (numerically) the cut locus of an ellipsoid, or a sphere with some particular metric with singularities.

**Our method.** The strategy we use is quite different. Given $m > 0$ a constant, let $u_m$ be the minimizer of the following variational problem

$$
\min_{u \in H^1(S)} \int_S \left( |\nabla_S u|^2 - mu \right),
$$

where $\nabla_S$ denotes the gradient operator on the surface $S$. For $\lambda > 0$ to be chosen small, we will use the set $E_{m,\lambda} := \left\{ |\nabla_S u_m|^2 \leq 1 - \frac{\lambda^2}{u_m^2} \right\}$ as an approximation of $\text{Cut}_b(S)$. This is justified by some theoretical results regarding problem (1.1) obtained in [10], which will be summarized in section 4. Now the set $E_{m,\lambda}$ can be well approximated using finite elements on a triangulation of the surface $S$.

The rest of the paper is organized as follows. In section 2 we recall the notion of $\lambda$-medial axis that was introduced in [5], and summarize some of its properties. In section 3 following the strategy of the $\lambda$-medial axis, we define a "$\lambda$-cut locus" $\text{Cut}_b(S)_\lambda$ and show that it can be used as an approximation of the complete cut locus. In section 4 we recall the result from [10] which states that the set $E_{m,\lambda}$ defined above is a good approximation of $\text{Cut}_b(S)_\lambda$ if $m$ is big enough. In section 5 we discretize problem (1.1) using finite elements, to find a discrete minimizer $v_h$, where $h > 0$ is the step of the discretization. We show that the set

$$
E_{m,\lambda,h} := \left\{ x \in S \setminus \{b\} : |\nabla_S u_{m,h}^i(x)|^2 \leq 1 - \frac{\lambda^2}{(u_{m,h}^i)^2(x)} \right\},
$$

is a good approximation of $E_{m,\lambda}$ as $h \to 0$. In section 6 we present the results of some numerical experiments.
2 \( \lambda \)-Medial axis

In this section, we recall briefly the notion of \( \lambda \)-medial axis introduced by Chazal and Lieutier in \cite{ChazalLieutier}. Given an open subset \( \Omega \) of \( \mathbb{R}^2 \), its medial axis \( \mathcal{M}(\Omega) \) is defined as the set of points of \( \Omega \) that have at least two closest points on the boundary \( \partial \Omega \) of \( \Omega \):

\[
\mathcal{M}(\Omega) := \{ x \in \Omega : \exists y, z \in \partial \Omega, y \neq z \text{ and } d_{\partial \Omega}(x) = |x - y| = |x - z| \},
\]

where for any \( x \in \Omega \), \( d_{\partial \Omega}(x) \) is the distance from \( x \) to the boundary \( \partial \Omega \),

\[
d_{\partial \Omega}(x) = \min \{ |x - y| : y \in \partial \Omega \}.
\]

The medial axis \( \mathcal{M}(\Omega) \) is unstable with respect to small non-smooth perturbations of the boundary of \( \Omega \). To deal with this issue, Chazal and Lieutier defined the so called \( \lambda \)-medial axis of \( \Omega \) by setting, for any \( \lambda > 0 \),

\[
\mathcal{M}_\lambda(\Omega) := \{ x \in \Omega : r(x) \geq \lambda \},
\]

where \( r(x) \) is the radius of the smallest ball containing the set of all closest points to \( x \) on \( \partial \Omega \), i.e. the set \( \{ z \in \partial \Omega : |x - z| = d_{\partial \Omega}(x) \} \). The map \( \lambda \mapsto \mathcal{M}_\lambda(\Omega) \) is non increasing, and

\[
\mathcal{M}(\Omega) = \bigcup_{\lambda > 0} \mathcal{M}_\lambda(\Omega).
\]

It is further proved in \cite{ChazalLieutier} section 3, theorem 2] that \( \mathcal{M}_\lambda(\Omega) \) has the same homotopy type as \( \mathcal{M}(\Omega) \), for \( \lambda \) small enough. These facts justify that \( \mathcal{M}_\lambda(\Omega) \) is a good approximation of \( \mathcal{M}(\Omega) \), for \( \lambda \) small enough. The crucial difference though is that \( \mathcal{M}_\lambda(\Omega) \) is stable with respect to small variations of \( \Omega \), whereas \( \mathcal{M}(\Omega) \) is not. We refer the reader to \cite{ChazalLieutier} section 4] for precise statements and proofs.

3 \( \lambda \)-Cut locus

We want to define a set similar to the \( \lambda \)-medial axis, in the case of the cut locus \( \text{Cut}_b(S) \). To this end, we note that, as it can be seen from \cite{ChazalLieutier} section 2.1], we have

\[
\mathcal{M}_\lambda(\Omega) = \left\{ x \in \Omega : |\nabla d_{\partial \Omega}(x)|^2 \leq 1 - \frac{\lambda^2}{d_{\partial \Omega}^2(x)} \right\},
\]

where \( \nabla d_{\partial \Omega} \) denotes the generalized gradient wherever \( d_{\partial \Omega} \) is not differentiable. Analogously, for \( \lambda > 0 \), we define the \( \lambda \)-cut locus as

\[
\text{Cut}_b(S)_\lambda := \left\{ x \in S \setminus \{ b \} : |\nabla_b d_b(x)|^2 \leq 1 - \frac{\lambda^2}{d_b^2(x)} \right\}.
\]

Note that, according to \cite{Langeron} Proposition 3.4], the function \( d_b \) is locally semiconcave on \( S \setminus \{ b \} \), so it has a generalized gradient everywhere on \( S \setminus \{ b \} \), whose norm is given by the following formula:

\[
|\nabla_b d_b|(x) = \max(0, \sup_{v \in T_x S, |v| = 1} \partial^+_v d_b(x)).
\]
We have the following proposition from [10, Proposition 2.9].

**Proposition 3.1.** The map \( \lambda \mapsto \text{Cut}_b(S)_\lambda \) is non-increasing, and

\[
\text{Cut}_b(S) = \bigcup_{\lambda > 0} \text{Cut}_b(S)_\lambda.
\]

In addition, the following proposition holds.

**Proposition 3.2.** If \( S \) is a real analytic surface, then for \( \lambda > 0 \) small enough, one of the connected component of \( \text{Cut}_b(S)_\lambda \) has the same homotopy type as \( \text{Cut}_b(S) \), while the other connected components, if any, are contractible.

These two propositions justify that \( \text{Cut}_b(S)_\lambda \) is a good approximation of \( \text{Cut}_b(S) \), for \( \lambda > 0 \) small enough. Before proving proposition 3.2, we prove the following lemma.

**Lemma 3.3.** Let \( x \in \text{Cut}_b(S) \) be such that there exists two unit speed minimizing geodesics \( \gamma_1, \gamma_2 : [0, d_b(x)] \to S \) such that \( \gamma_i(0) = b \) and \( \gamma_i(d_b(x)) = x \). Let \( \theta \in (0, \pi) \) be the angle between \( \gamma_1 \) and \( \gamma_2 \) at \( x \). Then, we have

\[
|\nabla_S d_b|(x) \leq \cos(\theta/2).
\]

**Proof.** For \( i = 1, 2 \), let us set \( v_i = -\dot{\gamma}_i(d_b(x)) \). Let \( v \in T_x S \). Let us denote by \( \exp_x \) the Riemannian exponential map at the point \( x \). Let \( t_0 \in (0, d_b(x)) \) and \( x_i = \exp_{x_i} t_0 \). Note that we have \( x \notin \text{Cut}_b(S) \), so the function \( d_{x_i} \) is smooth at \( x \), and its gradient is \( -v_i \). Given \( v \in T_x S \) such that \( |v| = 1 \), we have

\[
\partial^+_v d_b(x) = \lim_{t \to 0^+} \frac{d_b(\exp_x(vt)) - d_b(x)}{t} \\
\leq \lim_{t \to 0^+} \frac{d_{x_i}(\exp_x(vt)) + d_b(x_i) - (d_{x_i}(x) + d_b(x_i))}{t} \\
= \lim_{t \to 0^+} \frac{d_{x_i}(\exp_x(vt)) - d_{x_i}(x)}{t} \\
= -v \cdot v_i
\]

Given that the angle between \( v_1 \) and \( v_2 \) is \( \theta \), there exists \( i \in \{1, 2\} \), such that the angle between \( v \) and \( v_i \) is at most \( \pi - \theta/2 \). Thus the last inequality gives \( \partial^+_v d_b(x) \leq \cos(\theta/2) \). This concludes the proof.

Using lemma 3.3, Proposition 3.2 will mainly be a consequence of [7, Proposition 3.4] and the proof of [7, Proposition 3.5]. Following [7], we will use the following terminology. A point \( x \) of a finite graph \( G \) is called a tree point if \( G \setminus \{x\} \) has a connected component whose closure is a tree. Otherwise, \( x \) is called a cycle point. It is a consequence of the proof of [7, Proposition 3.5] that any closed connected subset of \( G \) that contains all cycle points is a deformation retract of \( G \).
Proof of proposition 3.2. As $S$ is real analytic, the cut locus $\text{Cut}_b(S)$ is a finite graph (see [14] in dimension 2, and [4] for the generalization to arbitrary dimensions). According to the lemma 3.3, given any $\theta > 0$, if $\lambda$ has been taken small enough, then for any point $x \in \text{Cut}_b(S) \setminus \text{Cut}_b(S)_\lambda$, the angle between the minimizing geodesics from $b$ to $x$ is smaller than $\theta$. Given two unit speed minimizing geodesics $\gamma_1$ and $\gamma_2$, following [7], the spread between $\gamma_1$ and $\gamma_2$ is defined as

$$\text{spd}(\gamma_1, \gamma_2) = \sup_t d(\gamma_1(t), \gamma_2(t)).$$

As geodesics verify a second order differential equation, if their angle at their common starting point is small, then their spread is also small. Therefore, applying [7, Proposition 3.4], we deduce that if $\lambda$ has been taken small enough, then any point $x \in \text{Cut}_b(S) \setminus \text{Cut}_b(S)_\lambda$ is a tree point of $\text{Cut}_b(S)$.

It remains to show that $\text{Cut}_b(S)_\lambda$ is closed to conclude that it is a deformation retract of $\text{Cut}_b(S)$ and conclude the proof. But this is a consequence of the fact that $d_b$ is semiconcave, and the upper semicontinuity of the generalized gradient of convex functions. 

Therefore, we will use $\text{Cut}_b(S)_\lambda$ as an approximation of $\text{Cut}_b(S)$ for $\lambda$ small enough.

4 Approximation with a variational problem

For $m > 0$, recall that $u_m$ is the minimizer in (1.1). For $\lambda > 0$, let us define the set $E_{m,\lambda}$ by

$$E_{m,\lambda} := \left\{ x \in S \setminus \{b\} : |\nabla S u_m(x)|^2 \leq 1 - \frac{\lambda^2}{u_m^2(x)} \right\}.$$

We have the following theorem (see [10, Theorem 1.1 and Theorem 1.3]):

**Theorem 4.1.** For any $m > 0$, the function $u_m$ is locally $C^{1,1}$ on $S \setminus \{b\}$. For any $m > m' > m_0$,

$$\text{Cut}_b(S) \subset \{ |\nabla S u_m| < 1 \} \subset \{ |\nabla S u_{m'}| < 1 \}. \quad (4.1)$$

Moreover,

$$\{ |\nabla S u_m| < 1 \} \longrightarrow_{m \to +\infty} \text{Cut}_b(S) \text{ in the Hausdorff sense.} \quad (4.2)$$

Finally, for any $\varepsilon > 0$,

$$\sup_{x \in E_{m,\lambda}} d(x, \text{Cut}_b(S)_\lambda) \longrightarrow_{m \to +\infty} 0, \quad \text{and} \quad \sup_{x \in \text{Cut}_b(S)_\lambda + \varepsilon} d(x, E_{m,\lambda}) \longrightarrow_{m \to +\infty} 0. \quad (4.3)$$

Therefore, we can use $E_{m,\lambda}$ as an approximation of $\text{Cut}_b(S)_\lambda$. All in all, we will use $E_{m,\lambda}$ as an approximation of $\text{Cut}_b(S)$. 

5 Discretization

5.1 Finite elements of order $r$ on a surface approximation of order $k$

In this section we introduce a discretization framework adapted to variational problem (1.1) based on finite elements. We follow the notations of [6, 9].

Let $S$ be a compact oriented smooth two-dimensional surface embedded in $\mathbb{R}^3$. For $x \in S$, we denote by $\nu(x)$ the oriented normal vector field on $S$. Let $d : \mathbb{R}^3 \to \mathbb{R}$ be the signed distance associated to $S$ and $U_\eta = \{x \in \mathbb{R}^3, |d(x)| < \eta\}$ the tubular neighborhood of $S$ of width $\eta > 0$. It is well known that if $\eta$ is small enough (for instance $0 < \eta < \min_{i=1,2} \frac{1}{\kappa_i} L_\infty(S)$ where the $(\kappa_i)$ stand for the extremal sectional curvatures of $S$), then for every $x \in U_\eta$ it exists a unique $a(x) \in S$ such that

$$x = a(x) + d(x)\nu(a(x)) = a(x) + d(x)\nabla d(x). \quad (5.1)$$

We consider $S^1_h$ a triangular approximation of $S$ whose vertices lie on $S$ and whose faces are quasi-uniform and shape regular of diameter at most $h > 0$. Moreover, we will assume that $T_h$, the set of triangular faces of $S_h$, are contained in some tubular neighborhood $U_\eta$ such that the map $a$ defined by (5.1) is unique.

For $k \geq 2$ and for a triangle $T \in T_h$, we consider the $n_k$ Lagrange basis functions $\Phi^k_1, \ldots, \Phi^k_{n_k}$ of degree $k$ and define the discrete projection on $S_h$ by:

$$a_k(x) = \sum_{j=1}^{n_k} a(x_j)\Phi^k_j(x) \quad (5.2)$$

where $x_1, \ldots, x_{n_k}$ are the nodal points associated to the basis functions. Now we can define $S^k_h$ a polynomial approximation of order $k$ of $S$ associated to $T_h$

$$S^k_h = \{a_k(x), x \in S_h\}. \quad (5.3)$$

Observe that by definition the image by $a$ of the nodal points are both on $S$ and on $S^k_h$. Let us now introduce the finite element spaces on $S_h = S^1_h$ and $S^k_h$ for $k \geq 2$. For every integer $r \geq 1$, let

$$L^r_h = \{\chi \in C^0(S_h), \chi|_T \in \mathbb{P}_r, \forall T \in T_h\} \quad (5.4)$$

where $\mathbb{P}_r$ is the family of polynomials of degree at most $r$. Analogously, for $k \geq 2$ let

$$L^{r,k}_h = \{\hat{\chi} \in C^0(S^k_h), \hat{\chi} = \chi \circ a^{-1}_k, \text{for some } \chi \in L^r_h\}. \quad (5.5)$$

Analogously to (1.1), we define

$$\min_{u \in L^{r,k}_h} F^k_h(u) \quad (5.6)$$

where $|\nabla^k u| \leq 1, u(\hat{b}) = 0$. 
where \( F_k(u) = \int_{S_k} \left( \left| \nabla_{S_k} u \right|^2 - mu \right) \) and \( b \) some fixed nodal points of the mesh \( T_h \).

5.2 Convergence of the lifted minimizers

In order to prove the convergence of our numerical approach, let us first establish that our discrete problem converges in values in the sense of proposition 5.2. For a function \( u \) defined on \( S_k^h \), we introduce its lifted function \( u^l \) onto \( S \) defined by the relation \( u^l(b) = u(x) \) for \( b \in S \) where \( x \) is the unique point of \( S_k^h \) which satisfies \( a(x) = b \).

Below, we focus our analysis in the piecewise linear case \( r = k = 1 \) which contains all the main ingredients of a proof for the general \((r,k)\) case. For every \( h > 0 \), the convex optimization problem (5.6) has a unique solution.

**Lemma 5.1.** The differential of the projection \( a \) onto \( S \), when restricted to the tangent space of \( S^h \), is the identity, up to order 2 in \( h \):

\[
Da|_{T^h} = Id + O(h^2).
\]

The second differential of \( a \), when restricted to the tangent space of \( S^h \), is null, up to order 1 in \( h \):

\[
D^2a|_{T^h} = O(h).
\]

**Proof.** The identity estimate on \( Da \) is a direct consequence of \[9\], equations (4.12), (4.13) and (4.11), and the fact that, following the notations of \[9\], lemma 4.1, we have \( \nu_{n+1}^2 = 1 - \sum_{j \leq n} \nu_j^2 \). The estimate on \( D^2a \) follows from the same equations, plus the identity \( D^2a(x) = -2\nabla d(x)D^2d(x) \).

Defining \( F(u) = \int_S \left( \left| \nabla_S u \right|^2 - mu \right) \), we have

**Proposition 5.2.** Let \( u_{m,h} \) be the solution of problem (5.6) for \( k = r = 1 \). Let \( Lu_{m,h} := \frac{u_{m,h}}{\left| \nabla_S u_{m,h} \right|_{L^\infty(S)}} \) be the 1-Lipschitz normalization of \( u_{m,h} \). Then, \( Lu_{m,h} \in H^1(S) \) and

\[
F(Lu_{m,h}) = \min_{\substack{u \in H^1(S) \\left| \nabla_S u \right| \leq 1 \\ u(b) = 0}} F(u) + O(h^{\frac{1}{2}}).
\]

**Proof.** step 1. Let \( u_m \) be the solution of problem (1.1). For \( \varepsilon > 0 \), let \( u_{m,\varepsilon} : S \to \mathbb{R} \) be defined by:

\[
u_{m,\varepsilon} = \begin{cases} 
\frac{d(x)^2}{2\varepsilon} & \text{if } d(x) \leq \varepsilon \\
\ u_m(x) - \frac{\varepsilon}{2} & \text{if } d(x) > \varepsilon.
\end{cases}
\]
According to [10, Lemma 3.3], we have \( u_m = d_h \) in a neighborhood of \( b \). Therefore, for \( \varepsilon > 0 \) small enough, we have \( u_m = d_h \) on \( B(b, 2\varepsilon) \). In particular, we deduce that \( u_{m,\varepsilon} \) is \( C^1 \) on \( S \). As \( d_h^2 \) is smooth in a neighborhood of \( b \), the gradient of \( d_h^2/2\varepsilon \) is \( O(\varepsilon^{-1}) \)-Lipschitz on \( B(b, \varepsilon) \). Moreover, as \( u_m = d_h \) on \( B(b, 2\varepsilon) \), the gradient of \( u_m \) is \( O(\varepsilon^{-1}) \)-Lipschitz on \( B(b, 2\varepsilon) \). According to lemma [10, Proposition 3.4], \( u_m \) is also locally \( C^{1,1} \) on \( S \setminus \{ b \} \). Therefore its gradient is \( O(\varepsilon^{-1}) \)-Lipschitz on \( S \setminus B(b, \varepsilon) \). All in all, we obtain that \( u_{m,\varepsilon} \) is \( C^{1,1} \) on \( S \), and the Lipschitz constant of its gradient is \( O(\varepsilon^{-1}) \). Furthermore, as \( d_h \) and \( u_m \) are both 1-Lipschitz, we have \( |\nabla u_{m,\varepsilon}| \leq 1 \). Now for \( \varepsilon > 0 \), consider

\[
v_{h,\varepsilon} := \frac{I_h u_{m,\varepsilon}}{\|\nabla_{S_h} I_h u_{m,\varepsilon}\|_{L_{\infty}(S_h)}},
\]

where \( I_h u_{m,\varepsilon} \) is the \( P^1 \) Lagrange interpolation of \( u_{m,\varepsilon} \) on \( S_h \). For \( x \in S_h \), observe that we have the relation \( I_h u_{m,\varepsilon}(x) = I_h (u_{m,\varepsilon} \circ a)(x) \) which says that \( I_h u_{m,\varepsilon} \) is the standard (flat) interpolation of the composed function \( u_{m,\varepsilon} \circ a \). From lemma [5.1] we know that on every triangle, the differential of \( a \) is \( O(h) \)-Lipschitz, and \( a \) is \( O(1) \)-Lipschitz. As the gradient of \( u_{m,\varepsilon} \) is \( O(\varepsilon^{-1}) \)-Lipschitz, we deduce that on every triangle, the gradient of \( u_m \circ a \) is \( O(\varepsilon^{-1}) \)-Lipschitz. By the quasi uniformity of the mesh, we obtain the uniform interpolation estimates on \( S_h \):

\[
I_h u_{m,\varepsilon}(x) = (u_{m,\varepsilon} \circ a)(x) + O(\varepsilon^{-1}h^2)
\]

and

\[
\nabla_{S_h} I_h u_{m,\varepsilon}(x) = \nabla_{S_h} (u_{m,\varepsilon} \circ a)(x) + O(\varepsilon^{-1}h).
\]

With lemma [5.1] we deduce for all \( x \in S_h \),

\[
\nabla_{S_h} I_h u_{m,\varepsilon}(x) = \nabla_S u_{m,\varepsilon}(a(x)) + O(\varepsilon^{-1}h).
\]

Recall that we have \( |\nabla_S u_{m,\varepsilon}|_{L_{\infty}(S)} = 1 \). Therefore the last identity yields

\[
|\nabla_{S_h} I_h u_{m,\varepsilon}|_{L_{\infty}(S_h,\varepsilon)} = 1 + O(\varepsilon^{-1}h).
\]

Thus, \( v_{h,\varepsilon} = I_h u_{m,\varepsilon}(1 + O(\varepsilon^{-1}h)) \), and so

\[
F_h(v_{h,\varepsilon}) = F_h(I_h u_{m,\varepsilon}) + O(\varepsilon^{-1}h).
\]

Applying lemma [5.1] again, with a simple change of variable, we find that for any function \( f : S_h \to \mathbb{R} \),

\[
\int_{S_h} f \circ a = \int_S f + O(h^2).
\]

Recalling [5.7] and [5.8], we obtain

\[
F_h(I_h u_{m,\varepsilon}) = F(u_{m,\varepsilon}) + O(\varepsilon^{-1}h).
\]

Furthermore, we have

\[
\int_S |u_{m,\varepsilon} - u_m| \leq O(\varepsilon) \quad \text{and} \quad \int_S |\nabla u_{m,\varepsilon} - \nabla u_m|^2 \leq O(\varepsilon^2),
\]

\[
\int S |u_{m,\varepsilon} - u_m| \leq O(\varepsilon)
\]

and

\[
\int S |\nabla u_{m,\varepsilon} - \nabla u_m|^2 \leq O(\varepsilon^2),
\]
so
\[ F(u_{m, \epsilon}) = F(u_m) + O(\epsilon). \]

Combining this with (5.9) and (5.10), we find
\[ F_h(u_{h, \epsilon}) = F(u_m) + O(\epsilon^{-1}h) + O(\epsilon). \]

Choosing \( \epsilon = h^{\frac{1}{2}} \), this yields
\[
\min_{u \in H^1(S_h)} F_h \leq \min_{u \in H^1(S)} F + O(h^{\frac{1}{2}}).
\]

**step 2.** Symmetrically, let \( u_{m,h} \) the solution of the discrete problem (5.6), \( u^l := u_{m,h} \circ (a_{|S_h})^{-1} \) its lifted version on \( S \), and \( L_{u_{m,h}}^l := \frac{u^l}{\|\nabla_{S_h} u^l\|_{L^\infty(S_h)}} \). We show as before, using the equation \( u_{m,h} = u_{m,h} \circ a \), that \( F(L_{u_{m,h}}^l) = F_h(u_{m,h}) + O(h) \). With (5.11), this implies
\[
\min_{u \in H^1(S)} F \leq F(L_{u_{m,h}}^l) \leq \min_{u \in H^1(S)} F + O(h^{\frac{1}{2}}),
\]
which concludes the proof of the proposition.

We can now establish the convergence of the minimizers:

**Proposition 5.3.**
\[
\left| \nabla u_{m,h}^l - \nabla u_m \right|_{L^2(S)}^2 = O(h^{\frac{1}{2}}) \quad \text{and} \quad \left| u_{m,h}^l - u_m \right|_{L^1(S)} = O(h^{\frac{1}{2}}).
\]

**Proof.** Consider \( v = \frac{1}{2}(L_{u_{m,h}}^l + u_m) \). Then, \( v \) is admissible for problem (1.1), so \( F(v) \geq F(u_m) \). Moreover, the following algebraic identity holds
\[
F(v) = \frac{1}{2} F(L_{u_{m,h}}^l) + \frac{1}{2} F(u_m) - \frac{1}{4} \int_S |\nabla_S u_m - \nabla_S L_{u_{m,h}}^l|^2.
\]

Therefore, we have
\[
\frac{1}{2} F(L_{u_{m,h}}^l) - \frac{1}{2} F(u_m) \geq \frac{1}{4} \int_S |\nabla_S u_m - \nabla_S L_{u_{m,h}}^l|^2,
\]
which proves, with proposition 5.2 that
\[
\left| \nabla L_{u_{m,h}}^l - \nabla u_m \right|_{L^2(S)}^2 = O(h^{\frac{1}{2}}).
\]

Moreover, we have
\[
F(L_{u_{m,h}}^l) - F(u_m) = \int_S \left( |\nabla L_{u_{m,h}}^l|^2 - |\nabla u_m|^2 \right) - m \int_S (L_{u_{m,h}} - u_m).
\]
The last two equations imply

$$|L u_{m,h}^l - u_m|_{L^1(S)} = \mathcal{O}(h^{\frac{1}{2}}).$$

(5.13)

As in the proof of proposition 5.2 using the relation $u_{m,h} = u_{m,h}^l \circ h$, we show that $L u_{m,h}^l = u_{m,h}^l (1 + \mathcal{O}(h^2))$. Together with (5.12) and (5.13), this concludes the proof.

We just proved that the sequence of the lifted minimizers converges with an order at least $1/2$ to the minimizer of problem (1.1). By analogy with the more standard variational context [6, 9], we expect a convergence of order $\mathcal{O}((h^r + h^{k+1})^{\frac{1}{2}})$ using an approximation of orders $(r, k)$.

5.3 Convergence in measure to the elastic set

Let us recall that the set $E_{m,\lambda}$ is defined by

$$E_{m,\lambda} = \left\{ x \in S \setminus \{b\} : |\nabla_s u_m(x)|^2 \leq 1 - \frac{\lambda^2}{u_m^2(x)} \right\}.$$

Proposition 5.4. For any $\lambda > 0$ and $\varepsilon > 0$ with $\varepsilon < \lambda/2$, let us define

$$E_{m,\lambda,h} := \left\{ x \in S \setminus \{b\} : |\nabla_s u_{m,h}^l(x)|^2 \leq 1 - \frac{\lambda^2}{(u_{m,h}^l(x))^2} \right\}.$$

Then, we have

$$|E_{m,\lambda + \varepsilon} \setminus E_{m,\lambda,h}| = \mathcal{O}(h^{\frac{1}{2}}) \quad \text{and} \quad |E_{m,\lambda,h} \setminus E_{m,\lambda - \varepsilon}| = \mathcal{O}(h^{\frac{1}{2}}).$$

Proof. By definition of $E_{m,\lambda}$ and $E_{m,\lambda,h}$, we have

$$E_{m,\lambda + \varepsilon} \setminus E_{m,\lambda,h} \subset \left\{ \frac{|\nabla u_{m,h}^l|^2 - |\nabla u_m|^2}{u_m^2} > \frac{(\lambda + \varepsilon)^2}{u_m^2} - \frac{\lambda^2}{(u_{m,h}^l)^2} \right\}.$$

Therefore, on $E_{m,\lambda + \varepsilon} \setminus E_{m,\lambda,h}$, we have

$$|\nabla u_{m,h}^l|^2 - |\nabla u_m|^2 > \frac{(\lambda + \varepsilon)^2}{u_m^2} - \frac{\lambda^2}{u_m^2} - \frac{2}{\min(u_m, u_{m,h}^l)^3} |u_m - u_{m,h}^l| \geq \frac{2\varepsilon\lambda + \varepsilon^2}{u_m^2} - \frac{2}{\min(u_m, u_{m,h}^l)^3} |u_m - u_{m,h}^l|,$$

where $\text{diam} S$ is the diameter of $S$. By definition of $E_{m,\lambda}$, we also have $E_{m,\lambda + \varepsilon} \subset \{ u_m \geq (\lambda + \varepsilon) \}$, so on $E_{m,\lambda + \varepsilon} \setminus E_{m,\lambda,h}$,

$$|\nabla u_{m,h}^l|^2 - |\nabla u_m|^2 > \frac{2\varepsilon\lambda + \varepsilon^2}{(\text{diam} S)^2} - \frac{2}{(\lambda + \varepsilon + \mathcal{O}(h^{\frac{1}{2}}))^3} |u_m - u_{m,h}^l|. $$
So for $h$ big enough, we have

\[ E_{m,\lambda+\varepsilon} \setminus E_{m,\lambda,h} \]

\[ \subset \left\{ \left| \nabla u_{m,h}^I \right|^2 - \left| \nabla u_m \right|^2 + 2\lambda \left| u_m - u_{m,h}^I \right| > \frac{2\varepsilon\lambda + \varepsilon^2}{(\text{diam } S)^2} \right\}, \]

\[ \subset \left\{ \left| \nabla u_{m,h}^I \right|^2 - \left| \nabla u_m \right|^2 > \frac{2\varepsilon\lambda + \varepsilon^2}{2(\text{diam } S)^2} \right\} \cup \left\{ \left| u_m - u_{m,h}^I \right| > \frac{2\varepsilon\lambda + \varepsilon^2}{4\lambda(\text{diam } S)^2} \right\}. \]

Now from proposition 5.3 we know that for any $\eta > 0$, we have the following estimates

\[ \eta \left| \left\{ \left| \nabla u_{m,h}^I \right|^2 - \left| \nabla u_m \right|^2 > \eta \right\} \right| \leq \int_S \left| \nabla u_{m,h}^I \right|^2 - \left| \nabla u_m \right|^2 = O(h^{\frac{1}{2}}), \]

and

\[ \eta \left| \left\{ \left| u_m - u_{m,h}^I \right| > \eta \right\} \right| \leq \int_S \left| u_m - u_{m,h}^I \right| = O(h^{\frac{1}{2}}). \]

This gives the estimate $|E_{m,\lambda+\varepsilon} \setminus E_{m,\lambda,h}| = O(h^{\frac{1}{2}})$. The other estimate is proved by the same method.

\[ \square \]

Remark 5.5. We expect a convergence of order $O((h^r + h^{k+1})^{\frac{1}{2}})$ using an approximation of orders $(r,k)$.

Sections 3, 4 and 5 together justify that the set $E_{m,\lambda,h}$ is a good approximation of the cut locus of $b$ in $M$, if $m$ is big enough, and $\lambda$ and $h$ are small enough.

6 Numerical illustrations

6.1 Cut locus approximation

We established the convergence of the minimizers of solutions of problems (1.1) when $h$ tends to 0. For a fixed $h > 0$, this convex discrete problem is of quadratic type with an infinite number of conic pointwise constraints. By the way, it is important to observe that for $k = r = 1$, the gradient pointwise bounds for a function of $P^1$ is equivalent to a single discrete conic constraint on every triangle with respect to the degrees of freedom of $P^1(T_h)$.

Nevertheless, we observed in our experiments that using $P^1$ elements may lead to approximated cut loci with some tiny artificial connected components. Motivated by this lack of precision, we use in all following illustrations elements of order $r > 1$.

For the general case $r > 1$, the bound constraint on the gradient can not be easily reduced to a finite set of discrete constraints. In our experiments, we approximated the constraint $|\nabla_{S_h^k} u|_{L^\infty(S_h^k)} \leq 1$ by forcing this constraint only on a finite number of points of the mesh. In practice, we imposed these constraints on the Gauss quadrature points of order $g$ on every triangle of $T_h$. 

11
We illustrate in figures 1, 2, 3 and 4 the approximation of the cut locus provided by our approach. These computations have been carried out on meshes of approximated $10^5$ triangles for $k = 2$ and $r = 3$ using high precision quadrature formula associated to 17 Gauss points on every element of the mesh. Moreover, for $r = 3$, we imposed the conic gradient constraints on the $g = 9$ Gauss points of every triangle. In order to solve the resulting linear conic constrained quadratic optimization problem, we used the JuMP modeling language and the finite elements library Getfem++ combined with Mosek optimization solver [2]. For such a precision, the optimization solver identified a solution in less than one hour on a standard computer.

Figure 1: Three different views of the approximation of a cut locus on a standard torus

Figure 2: Three different views of the approximation of a cut locus on a standard torus, without representing the surface
Figure 3: Three different views of the approximation of a cut locus on a torus of genus 2

Figure 4: Three different views of the approximation of a cut locus on a torus of genus 2, without representing the surface
6.2 Approximation of the boundary of Voronoi cells

All previous theoretical results still hold if we replace the source point \( b \) by any compact subset of the surface \( S \). For instance, if \( b \) is replaced by a set of points, the singular set of the distance function can be decomposed as the union of the boundary of Voronoi cells and the cut loci of every point intersected with its Voronoi cell. As a consequence, if the distribution of source points is homogeneous enough, that is every Voronoi cell is small enough, the singular part of the distance function will be exactly equal to the boundary of the Voronoi cells. We illustrate this remark in the following experiments. We used exactly the same framework as in previous sections and just replaced the pointwise condition at \( b \) with the analogous pointwise Dirichlet conditions at every source point. Figure 5 and 6 represent the Voronoi diagrams obtained with 10, 30 and 100 points for surfaces of genus 2 and 3. The expected computational complexity is exactly of the same order as with a single source point.

Figure 5: Approximation of the Voronoi cells on a torus of genus 2 of 10, 30 and 100 points. Every column represents two different views.
Figure 6: Approximation of the voronoi cells on a torus of genus 3 of 10, 30 and 100 points. Every column represent two different views.

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