Abstract

Many applications in geometric modeling, computer graphics, visualization and computer vision benefit from a reduced representation called curve-skeletons of a shape. These are curves possibly with branches which compactly represent the shape geometry and topology. The lack of a proper mathematical definition has been a bottleneck in developing and applying the curve-skeletons. A set of desirable properties of these skeletons has been identified and the existing algorithms try to satisfy these properties mainly through a procedural definition. We define a function called medial geodesic on the medial axis which leads to a mathematical definition and an approximation algorithm for curve-skeletons. Empirical study shows that the algorithm is robust against noise, operates well with a single user parameter, and produces curve-skeletons with the desirable properties. Moreover, the curve-skeletons can be associated with additional attributes that follow naturally from the definition. These attributes capture shape eccentricity, a local measure of how far a shape is away from a tubular one.

1. Introduction

The problem of representing a three dimensional shape with a one dimensional geometry (curves) appears in various applications of geometric modeling, computer graphics, visualization and computer vision. For example, in animation and tracking a ‘stick-figure’ is immensely useful which represents the main geometric entities of a shape with their connectivities. It allows registrations, deformations, matching and other operations in a more controlled manner because of the reduced dimension. The concept of curve-skeleton was born from these various needs which, roughly speaking, should be a curve possibly with branches in the ‘center’ of the shape. A related and much more well defined concept is the medial axis which is also referred as the skeleton. The medial axis consists of the centers of the maximal balls inscribed inside the shape. For a three dimensional shape, the medial axis, in general, has two dimensional components often referred as medial surface. Therefore, medial axis cannot be a substitute for one dimensional skeletons.

A main problem with computing curve-skeletons is that they are not well defined. Although desirable properties of these skeletons have been identified based on different applications, no mathematical definition has been formulated. To fill this void different procedural definitions leading to different methods have been proposed for computing curve-skeletons. They include, to name a few, topological thinning [BNB99], distance field based methods [ZT99] [BKS01] [BST03] [HF05], potential field based methods [CSYB05], and others [OK95] [Cos99] [VL00]. Cornea et al. [CSM05] give a comprehensive survey of these techniques. Although many of these methods produce curve-skeletons with a set of desirable properties, they are not completely satisfactory as pointed out in Cornea et al. [CSM05]. We believe that this limitation stems from the lack of a proper mathematical definition of curve-skeletons.

In this paper, we give a mathematical definition of the curve-skeletons. Since the curve-skeleton should be in the ‘middle’ of the shape it is natural to define it as some subset of the medial axis. What we aim for is to determine the ‘middle’ of the medial axis. Algorithms to thin the medial axis based on the distances from the boundary have been designed on this principle. The main problem in this approach is that a large part or the entire medial axis may have the same distance from the boundary, e.g., the medial surface of...
Figure 1: (a) Female model, (b) approximated medial axis rendered with MGF values, (c) extracted curve-skeleton rendered with the eccentricity values. Coloring scheme: red tone : big values, blue tone : small values, green tone : medium values.

a thin plate. It is not clear how the thinning process should proceed in such cases. One of our main contributions is to define a function on the medial axis whose singularity brings out its ‘middle’. The function is based on the geodesic distances between points where the maximal balls defining the medial axis touch the shape boundary. We call it the [*medial geodesic function*](MGF). In a sense, the medial geodesic function combines the intrinsic property of the bounding surface (by geodesic distances) with its embedding in three space (by the medial axis) thereby capturing the shape information comprehensively. Just as the singular points of the standard distance function gives the medial axis, the singular points of this function gives the curve-skeleton. Figure 1(b) and (c) show the medial geodesic function and the curve-skeleton of the Female model respectively.

Our definition allows additional shape information to be associated with the curve-skeleton. First, the medial geodesic function values given by the shortest geodesic distances between the points where the maximal balls touch the surface give the size information of the shape. Second, the ratios between the geodesic and the Euclidean circles passing through these touching points tell how far the shape is locally away from a tubular one. We refer to this ratio as *eccentricity*. The coloring in Figure 1(c) shows the eccentricity values associated with the curve-skeleton. Furthermore, our definition allows to map the curve-skeleton back to the surface easily. These extra features are useful for various applications.

2. Definition

Let \( O \subseteq \mathbb{R}^3 \) be a space called *shape* bounded by a *connected manifold* surface \( S \). The *medial axis* \( M \subseteq O \) is the set of centers of the maximal balls inscribed in \( O \).

Giblin and Kimia [GK04] show that, generically, \( M \) consists of five types of points giving it a stratified structure. One type form two dimensional sheets, two of the types form curves and the rest two types remain as isolated points on the medial axis. Figure 2(a) shows four of these types.

Figure 2: (a) The stratified structure of the medial axis of a rectangular block. (b) One sheet of the medial axis. Red points \( x, y, z \) are on this sheet. Green points are their corresponding contact points on the surface and black paths are the shortest geodesic paths between these contact points. (c) Medial axis rendered with MGF values following the coloring scheme of the title figure. (d) Red lines, blue lines and green points are \( SK_2 \), \( SK_3 \) and the limit points of their union, respectively.

First we focus on the points that form sheets; shown with grey in Figure 2(a). The maximal inscribed balls of such points touch the surface \( S \) at exactly two distinct points. Let \( M_2 \subseteq M \) be the set of such points. Each point in \( M_2 \) has a neighborhood which is an open disk and hence \( M_2 \) is a 2-manifold. Since \( M \setminus M_2 \) has measure zero in \( M \), in general, \( M_2 \) covers most of \( M \).

2.1. Medial Geodesic Function (MGF)

For a point \( x \in M_2 \), let \( B_x \) be the maximal inscribed ball centered at \( x \) and \( a_x \) and \( b_x \) be the two points of \( S \) where \( B_x \) meets \( S \). Define \( f(x) \) to be the length of the shortest geodesic path on \( S \) between \( a_x \) and \( b_x \). We call \( f \) the *medial geodesic function*, MGF, of \( O \). Figure 2(b) shows the corresponding shortest geodesic paths for several points on \( M_2 \). If a point \( x \in M_2 \) have more than one shortest geodesic paths between \( a_x \) and \( b_x \), their lengths are equal. Hence MGF is well defined on \( M_2 \). Figure 2(c) shows the rendering of \( M_2 \) based on the MGF value. We will define the curve-skeleton in \( M_2 \) as the singular set of \( f \). Several properties of this singular set (some observed and some proved) motivates this definition.
By our assumption the surface \( S \) is connected, compact, and without boundary. Further, we assume that \( S \) is smooth (\( C^\infty \)). The function \( f \) is defined on \( M_2 \). Define \( \phi: M_2 \rightarrow S \times S, x \rightarrow (a_x, b_x) \). Using a technique similar to Attali et al. [AJDLO3], it can be shown that \( M_2 \) is a smooth manifold and \( \phi \) is differentiable provided \( S \) is smooth. We denote the lengths of the shortest geodesic between two points \( x \) and \( y \) in \( S, M_2, \) and \( \mathbb{R}^2 \) by \( d_S(x, y), d_M(x, y), \) and \( d(x, y) \) respectively. Considering \( d_S \) as a function from \( S \times S \) to \( \mathbb{R} \) we have \( f = d_S \circ \phi \). Let \( \alpha: M_2 \rightarrow \mathbb{R}^2 \) be the local coordinate function for \( x \in M_2 \). The map \( \alpha \) is a diffeomorphism since \( M_2 \) is a smooth manifold. We use \( \alpha \) to define \( \tilde{f}: \mathbb{R}^2 \rightarrow \mathbb{R} \) so that \( \tilde{f}(\alpha(x)) = f(x) \). From standard definition in differential geometry \( f \) is called differentiable at \( x \) if and only if \( \tilde{f} \) is differentiable at \( \alpha(x) \).

We argue that the singularities of \( f \), i.e., the points in \( M_2 \) where \( f \) is not differentiable has a Lebesgue measure zero. This would mean that the singularities of \( f \) constitute curves or isolated points on \( M_2 \), which we define as the curve-skeleton in \( M_2 \).

**Property 1** The singularity of \( f \) has measure zero in \( M_2 \).

To prove that the singularity of \( f \) has measure zero, we show that \( \tilde{f} \) is locally \( k \)-Lipschitz (defined below) for some \( k > 0 \). For then the singular set of \( \tilde{f} \) has measure zero by Rademacher’s theorem [Fed96]. It follows that the singular set of \( f \) has measure zero by local coordinate maps.

A function \( g: \mathbb{R}^n \rightarrow \mathbb{R} \) is called locally \( k \)-Lipschitz near a point \( x \in \mathbb{R}^n \) if for some \( \varepsilon > 0 \)

\[
g(x) - g(y) \leq k||x - y||
\]

for any \( y \in \mathbb{R}^n \) where \( ||x - y|| \leq \varepsilon \).

**Observation 1** The function \( \tilde{f}: \mathbb{R}^2 \rightarrow \mathbb{R} \) is locally \( k \)-Lipschitz for some \( k > 0 \).

**Proof** For some \( \varepsilon > 0 \) and a point \( x \in M_2 \), let \( y \) be any point where \( \partial \phi(x), \partial \phi(y) \leq \varepsilon \). Consider the shortest geodesics \( \gamma_a \) between \( a_x \) and \( b_x \), and \( \gamma_b \) between \( a_y \) and \( b_y \). The lengths \( |\gamma_a| \) and \( |\gamma_b| \) satisfy

\[
|\gamma_a| \leq |\gamma_b| + d_M(a_x, a_y) + d_M(b_y, b_x).
\]

Since \( \phi \) is differentiable, there is some \( k_1 > 0 \) so that we have \( \max\{d_S(a_x, a_y), d_S(b_x, b_y)\} \leq k_1 d_M(x, y) \) when \( x \) and \( y \) are sufficiently close. Also since the local coordinate function \( \alpha \) is differentiable, we have \( d_M(x, y) \leq k_2 d(\alpha(x), \alpha(y)) \) for some \( k_2 > 0 \). Therefore,

\[
\tilde{f}(\alpha(x)) = |\gamma_a| \leq |\gamma_b| + 2k_1 k_2 d(\alpha(x), \alpha(y)) = \tilde{f}(\alpha(y)) + 2k_1 k_2 d(\alpha(x), \alpha(y))
\]

proving that \( \tilde{f} \) is locally \( 2k_1 k_2 \)-Lipschitz.

We do not have rigorous mathematical proofs for the next two properties though we conjecture them to be true. We have observed the properties from experiments as well.

**Property 2** There is no local minimum of \( f \) in \( M_2 \).

This property can be argued roughly as follows. Since \( M_2 \) is a manifold, any point \( x \) in \( M_2 \) has a neighborhood \( N \subset M_2 \), which is a disk. Let \( \gamma \) be the geodesic path between \( a_x \) and \( b_x \). In any small enough neighborhood of \( x \), there is a point \( y \) such that \( a_y \) is on \( \gamma \). If \( b_x \) is also on \( \gamma \) then \( f(x) > f(y) \) and we are done. However, in general, \( b_x \) may not be on \( \gamma \). But we observe that \( b_x \) is close to \( \gamma \) and hence it is likely that \( f(x) > f(y) \) still holds.

**Property 3** At each singular point \( x \) of \( f \) there are more than one shortest geodesic paths between \( a_x \) and \( b_x \).

A rough argument why the above property is true may proceed as follows. We have \( f = d_S \circ \phi \) and \( \phi \) is differentiable on \( M_2 \). Therefore, \( f \) is differentiable at a point \( x \in M_2 \) if \( d_S \) is differentiable at \( \phi(x) \). Suppose that there is only a single shortest geodesic path \( \gamma \) between \( a_x \) and \( b_x \). Then in a sufficiently small neighborhood \( N \) of \( (a_x, b_x) \) in \( S \times S \), all geodesic paths between \( a \) and \( b \) for \( (a, b) \in N \) smoothly converge to \( \gamma \) as \( (a, b) \) approaches to \( (a_x, b_x) \). This means \( d_S \) is smooth at \( (a_x, b_x) \) and so is \( f \) at \( x \) contradicting that \( f \) is singular at \( x \).

### 2.2. Skeleton definition

We observe that the behavior of the medial geodesic function is like a distance function. First, MGF is continuous and differentiable everywhere on \( M_2 \) except at a measure zero set in \( M_2 \). Second, we have observed that property 2 and 3 hold in practice. This means MGF has no local minimum on \( M_2 \) which is open in \( M \) and the singularity of \( f \) occurs roughly in the ‘middle’ of \( M_2 \).

We define the curve-skeleton in \( M_2 \), denoted by \( S_{\text{sk}} \), as the set of singular points of MGF on \( M_2 \). To extend the definition beyond \( M_2 \), we use a different characterization of the singular points by means of divergence. It is reminiscent of the use of divergence for defining medial axis by Siddiqi et al. [SBTZ99]. The gradient of MGF, \( \nabla f \), defines a vector field on \( M_2 \) except at the singular points. The divergence of the vector field at point \( x \), \( \text{div}(x) \), is the net outward flux per unit area on \( M_2 \) taken over a neighborhood \( D \) shrinking to zero, i.e.,

\[
\text{div}(x) := \lim_{A \to 0} \frac{\int_C <\nabla f, n> dC}{A}
\]

where \( A \) is the area of \( D \), \( C \) is the boundary of \( D \) and \( n \) is the outward normal at a point on \( C \) as shown in Figure 3(a). The divergence is negative at the singular points but 0 everywhere else. In other words, \( S_{\text{sk}} \) consists of the points where the gradient flow of MGF sinks into.

Next we consider the set of points \( M_3 \subset M \) where \( M_3 \) constitutes curves lying at the intersection of the closure of three sheets in \( M_2 \). The thick black lines in Figure 2(a) are such curves. The maximal inscribed ball of such a point touches \( S \) at three points. Although MGF is not well defined for these points, we consider MGF defined on their neighborhoods.
Let $x$ be such a point. The neighborhood of $x$ in $M_2$ consists of three topological half disks. Consider one of them, say $HD$. We define the divergence with respect to $HD$ for $x$, $\text{div}(x)|_{\text{HD}}$, as follows.

\[
\text{div}(x)|_{\text{HD}} := \lim_{\text{HA} \to 0} \frac{\int_{\text{HC}} \langle \nabla f, n \rangle \text{d}HC}{\text{HA}}
\]

where $\text{HA}$ is the area of the half disk $HD$. $HC$ is the boundary of $HD$ on $M_2$ as in Figure 3(b). The point $x$ is on the curve-skeleton if $\text{div}(x)|_{\text{HD}}$ is negative for all three half disks in the neighborhood of $x$. Basically, $x$ is on the curve-skeleton if the gradient flow of MGF from all three local neighboring sheets sink into it. Let $Sk_3$ denote such set of points.

Now consider the rest of the types of points in $M$. One type of points form the boundary curves of $M$ where two contact points of the maximal inscribed ball with the surface coincide. In case of the rectangular block shown in Figure 2(a), these curves are the twelve edges of the block. The rest two types of points are the isolated points on the medial axis where at least two curves meet. We do not explicitly define the curve-skeleton for these three types of points since they are either the boundary or the isolated points on the medial axis. A point of one of these three types is on the curve-skeleton if it is the limit point of $Sk_2$ or $Sk_3$.

Finally, we define the curve-skeleton of $O, Sk_O$ as the closure of $Sk_2 \cup Sk_3$. Figure 2(d) shows the curve-skeleton for a rectangular block.

3. Algorithm Overview

In general it is extremely hard to compute the curve-skeleton exactly as we defined. It is well known that exact medial axis computation is hard due to numerical instability associated with the computations. Performing an exact computation based on the analysis of a function defined on the medial axis would be even harder. We bypass this difficulty by computing an approximation of the curve-skeleton. Extensive experiments show that the algorithm is effective in practice.

Assume that the input is a shape represented by a polygonal surface. Ideally, the approximation algorithm can be described in the following three steps. First, compute a polygonal approximation of the medial axis. Second, compute the gradient of MGF for the center of each medial axis facet (polygon), which approximates the gradient for all points in that facet. Third, mark the edges of the medial axis facets with negative divergence as skeleton edge. The output curve-skeleton consists of all marked skeleton edges. Figure 4(a) illustrates a curve-skeleton computed with this strategy. This result, however, is an accident. In practice, the polygonal approximation of the medial axis will most likely be worse and hence the computation of the gradient of MGF will be less accurate. As a result, the curve-skeleton computed by the above three steps will most likely be disconnected as Figure 4(b) illustrates. To overcome this problem caused by discretization and approximation error, we introduce an additional step of medial axis erosion to recover the missing part. The medial geodesic function guides the erosion, which is the reason why the curve-skeleton is roughly in the middle of the medial axis. Specifically, the erosion only removes pieces of the medial axis from the boundary of the subset that has yet not been eroded. Also, among all erodable elements, the one with the smallest MGF value is eroded first. At the same time the edges marked in the third step are never allowed to be eroded. Figure 4(c) shows a stage where three facets still need to be eroded. Figure 4(d) shows the extracted curve-skeleton.

Extraction Algorithm:

step 1: (MA approximation) Compute a polygonal approximation of the medial axis using the input polygonal surface.

step 2: (MGF approximation) Approximate the medial geodesic function and its gradient for the points inside each medial axis facet.

step 3: (Marking) Mark the edges of the medial axis facets with negative divergence as skeleton edges.

step 4: (Erosion) Erode the medial axis in the increasing order.
Figure 5: The first row shows the extracted curve-skeletons for surfaces with genus 0. The second row shows the extracted curve-skeletons for surfaces with genus more than 1. The skeleton edges are colored based on their eccentricity values.

The extracted curve-skeleton is centered because of the following two reasons. First of all, the curve-skeleton is a subset of the approximated medial axis and hence is centered with respect to the distance field defined by the surface. Second, by property 3, a point $x$ in $M_2$ is on the curve-skeleton only if there are multiple shortest geodesic paths between two touching points, $a_x$ and $b_x$. This means that the point $x$ is in the middle of $M_2$. Different examples given in this paper also show the centeredness of the curve-skeleton.

A curve-skeleton should remain stable against small changes in the shape. In particular, small changes introduced by noise should not affect the curve-skeleton significantly. Although the medial axis based on which we extract the curve-skeleton is not stable under shape perturbations [Wol92, ACK02, AJDE04], the curve-skeleton remains stable. The reason is that the unstable parts of the medial axis do not contribute to our curve-skeleton. Figure 6(a) and (b) show a noisy Hand (we generate it by perturbing the points of a smooth Hand shown in Figure 13) and its medial axis respectively. As we can see the medial axis has a number of spikes because of noise. However, these spikes are close to the boundary and hence have small MGF value as the coloring of the medial axis shows. The erosion process erodes these spikes before reaching the “middle” of the medial axis, where the curve-skeleton is. Figure 6(c) shows the curve-skeleton of noisy Hand, which is almost the same as the one of smooth Hand in Figure 8 though it has more wiggles.

The curve-skeleton computed by our algorithm remains stable under certain deformations where the topology of the shape does not change and the geodesic distance between any two points on the surface does not change much. Deform-
The curve-skeletons of other models in this paper such as Female, Boy, Hand and Horse also show their ability to detect the junctions. Junction detection is a key property of a curve-skeleton which many applications depend on such as mesh decomposition and shape matching.

4.1. Shape eccentricity

In addition to those mentioned in [CSM05], the curve-skeleton extracted by our algorithm can be associated with two other attributes, which make them encode shape information more comprehensively. Our definition and algorithm allow easy computation of these two quantities for each skeleton edge. Consider the hand in Figure 8. We approximate the medial axis using a subset of the Voronoi diagram and hence each skeleton edge \( E \) is a Voronoi edge, whose dual Delaunay triangle \( t \) has three vertices on the surface (red points). The geodesic paths between each pair of them together form a ‘circle’ (red circles), called the geodesic circle of \( E \). The length of this circle, denoted \( g(E) \), captures the local size of \( O_E \) where \( O_E \) corresponds to the skeleton edge \( E \). Let \( c(E) \) denote the length of the circumcircle of the dual triangle \( t \) (blue circles). The ratio \( \frac{g(E)}{c(E)} \) essentially tells how much \( O_E \) deviates from a tubular shape. We call this ratio the eccentricity of \( E \), denoted by \( e(E) \). In Figure 8, the fingers have small eccentricity value and are close to tubular shape while the palm has a big eccentricity value and is more flat. In Section 5.6 we detail the algorithm for identifying tubular/flat regions of a shape using eccentricity.

5. Algorithm details

In this section we give the detailed description of each step of the extraction algorithm described in section 3. We assume that the input surface is a connected triangulated surface, denoted by \( T \).
5.1. Medial axis approximation

In this work we follow the Voronoi diagram filtration of Dey and Zhao [DZ03] to approximate the medial axis. This approach filters the Voronoi diagram of a set of vertices on the surface and retains a set of Voronoi facets to approximate the medial axis. The main problem with this medial axis computation is that the filtration, often guided by some input parameters, leave some unwanted spikes or holes in the approximate medial axis. However our algorithm is not affected by this shortcoming as we avoid the filtration and instead take all the Voronoi facets inside the space bounded by $T$ as a preliminary approximation of the medial axis, denoted by $M_T$. Eventually this superset of the medial axis gets eroded by our algorithm. Figure 6(b) and Figure 9(a),(b) show the approximated medial axes of Hand and T-shape respectively.

Note that the medial axis of a triangulated surface can be computed exactly [CKM04]. However, this exact computation is expensive. Fortunately, the medial axis approximated by the Voronoi diagram serves our purpose as long as the vertices of the triangulated surface form a dense sampling of the original object.

5.2. MGF Approximation

For a Voronoi facet $F$ on the approximated medial axis, let $a_F$ and $b_F$ be two endpoints of the dual Delaunay edge of $F$, as in Figure 9(c). We compute the shortest geodesic distance $f(F)$ between $a_F$ and $b_F$ by using the algorithm presented by V. Surazhsky et. al [SSK*05]. The black curve in Figure 9(a) shows the shortest geodesic path between $a_F$ and $b_F$ computed by their algorithm. We approximate the medial geodesic function for any point inside $F$ with this quantity $f(F)$. The medial axis in Figure 9(b) is rendered with different colors for different ranges of MGF values.

We approximate the gradient of the MGF for any point inside a Voronoi facet $F$ as follows. First, we compute the tangent directions $v_a$ and $v_b$ of the shortest geodesic path at the two endpoints $a_F$ and $b_F$ respectively. Next, we project these two vectors onto the Voronoi facet $F$ and approximate the gradient $\nabla f$ for any point inside $F$ using the normalized summation $v(F)$ of these two projected vectors. The white arrows in Figure 9(e) show the vector $v(F)$ for the Voronoi facets inside the white circle marked in Figure 9(b).

5.3. Marking

Because of the erosion process, we do not need to mark all skeleton edges in this step. In fact, the actual purpose of the marking step is to find those skeleton edges which form the boundary of the curve-skeleton. Consider a Voronoi edge $E$ and let $F$ be any Voronoi facet incident on it as shown in Figure 9(d). The dot product $d_E(F)$ of $v(F)$ and the inward normal $n$ of the edge $E$ towards $F$, approximates the flux flowing into any point on $E$ from the neighborhood in $F$. We mark the Voronoi edge $E$ as a skeleton edge if $d_E(F) < \theta$ for any incident Voronoi facet $F$ where $\theta$ is an input parameter. Actually $\theta$ is the only input parameter for the entire curve-skeleton extraction algorithm. We will show its effect later in this section. In addition, to avoid small branches in the final curve-skeleton, at most one edge is allowed to be marked as skeleton edge for a Voronoi facet in this step. Also we do not mark edges on the boundary of $M_T$. Notice that, whenever an edge is marked, its two endpoints are also marked. Figure 9(a) shows the skeleton edges (red) marked after this step. As we can see they are in the ‘middle’ of the medial axis but are disconnected. Since MGF has no local minimum in $M_2$ as Property 2 claims, the skeleton edges marked in this step can not form a loop.

5.4. Erosion

The erosion proceeds by collapsing facets, edges and vertices from $M_T$ gradually. Consider $M_T$ as a cell complex consisting of three types of cells: Voronoi facets, Voronoi edges and Voronoi vertices. A cell $\tau$ is a face of another cell $\sigma$ if $\tau$ is on the boundary of $\sigma$. We also say $\sigma$ is a coface of $\tau$. A pair $(\tau, \sigma)$ is a face-coface pair if $\tau$ is a face of $\sigma$. In our case, there are three types of such pairs: (edge, facet), (vertex, facet) and (vertex, edge). The erosion actually proceeds by collapsing face-coface pairs. One way to think of collapsing a pair $(\tau, \sigma)$ is to push every point on $\sigma$ and $\tau$ onto the
other boundary of \( \sigma \), see Figure 10. It is the same as \( \sigma \) and \( \tau \) being eroded. A face-coface pair \((\tau, \sigma)\) is collapsible if \( \sigma \) is the only coface of \( \tau \) that has not been collapsed so far. If we only collapse the collapsible pairs, the spaces before and after the collapse remain homotopy equivalent.

The MGF value guides the erosion. The assignment of MGF value to every facet in \( M_T \) is described in section 5.2. An edge and a vertex is assigned an MGF value equal to the maximal of the MGF values of the facets incident to it. The erosion removes a pair \((\tau, \sigma)\) where \( \tau \) is not a marked skeleton edge or a vertex in the marking phase and \( f(\tau) \) is the least among all such collapsible pairs.

step 1: Initialize a priority queue \( Q \) with all the initial collapsible pairs, the one having the smallest \( f(\tau) \) being at the top. These pairs are the edges on the boundary of \( M_T \) and their cofaces.

step 2: Pop the top element \((\tau, \sigma)\) out of \( Q \) and erode \( \sigma \) and \( \tau \). The erosion of these two cells may make some adjacent pairs collapsible. If neither of the two cells in a new collapsible pair has been marked in the marking step, push the pair into \( Q \). Repeat step 2 until \( Q \) is empty.

Figure 10: (a) Collapsing \((\tau, \sigma)\), (b) a series of collapses, red edges and vertices are marked in the marking step.

Table 5.4 shows the timing of the algorithm for some models.

### 5.5. Effect of \( \theta \)

Finally we describe the effect of the only user parameter \( \theta \), which is used in the marking step to identify the points with certain negative divergence of the gradient field of MGF. As \( \theta \) decreases, the condition for points being on the curve-skeleton becomes more strict and the resulting curve-skeleton becomes less detailed. Formally, let \( \text{SK}_\theta^O \) be the curve-skeleton of a shape \( O \) extracted with parameter \( \theta \). We have \( \text{SK}_{\theta_1}^O \subseteq \text{SK}_{\theta_2}^O \) if \( \theta_1 < \theta_2 \). Figure 11 shows a series of curve-skeletons of Protein with different \( \theta \) values. As \( \theta \) decreases, the curve-skeleton corresponding to the less prominent features go away. When \( \theta \) reaches a value a little less than \(-1\), no edge and point is marked in the marking step since the condition \( d_E(F) < \theta \) is not satisfied for any edge \( E \) and any of its incident facet \( F \) and the curve-skeleton reaches its simplest form consisting of loops only. Different \( \theta \) can be chosen for different applications of the curve-skeleton.

Figure 11: A series of curve-skeletons of Protein with different \( \theta \) values.

### 5.6. Computing tubular regions

Many shapes have pronounced features which are perceived to be tubular and flat. This shape information should be readable from the medial axis which encodes the shape compactly. A recent work of Goswami et.al [GDB06] points toward this direction. As a subset of the medial axis, the curve-skeleton together with the eccentricity value can indicate how much the shape differs locally from a tubular one.

First, all skeleton edges are classified into two types, ones with eccentricity values less than a “threshold” and the rest. Second, we compute the geodesic circle for each skeleton edge \( E \) and attach the mesh triangles intersected with the geodesic circle with an id number (say 1) if \( e(E) \) is less than the threshold and another id number (say 0) otherwise. Figure 12(a) shows the result after this step where the mesh triangles with only one id number are rendered with blue if id number is 1 or white if it is 0, those with no id number are rendered with grey color, and those with two id numbers are rendered with red color. To compute a component we carry out a depth first search starting from a mesh triangle with only one id number and walk to the adjacent triangles until a triangle with a different id number is reached. Figure 12(b) shows the tubular part (in blue) for Genus3 with a threshold 1.41.

Figure 13 shows the identification of tubular parts for two other models.

### 6. Comparisons

In this section, we make a brief comparison between our algorithm and some existing ones. Our algorithm extracts the
curve-skeleton by eroding the medial axis. Unlike other medial axis based algorithms, the extracted curve-skeleton by our algorithm remains stable against the noise even though the medial axis may not. Distance field based methods works nicely and efficiently for the tubular objects, see [HF05] for recent results. Since the points with the same distances from the boundary of some shapes may form surface patches, the distance field based methods may face difficulty in extracting curve-skeletons for those shapes. The potential field based method is proposed to fix this problem by taking into account the surface area, see for example [CSYB05]. However, in practice, the potential field based method still may fail for the shapes containing thin flat parts. Imagine a very thin flat plate. The variation of the potential field can be very subtle in the middle of this shape since the surface area on the sides (the only place producing different potentials) are relatively small and far away from the middle. Figure 14 shows the comparison results between our method and the potential field based method of [CSYB05]. Notice that the ears of Dog are thin and flat. The surfaces shown in Figure 14(a),(c) are the extracted isosurfaces of the volume data. We choose the parameter 'Potential field strength' to be 8 and 6 and the parameter 'highest divergence points' to be 65 and 20 for (a) and (c) respectively, which give the best results. The curve-skeleton extracted using potential field may not be connected (a) or may not be centered (ears in (c)). One more advantage of our method is that our algorithm needs only one user supplied parameter as opposed to many for the existing algorithms.

7. Discussions and Future work

In this paper, we introduce a mathematical definition of curve-skeletons for 3D shapes with connected manifold boundary. We present an algorithm to approximate these curve-skeletons. Extensive experiments show that the approximation algorithm is effective in practice. We also show that the extracted curve-skeletons enjoy many nice properties. It is appropriate to mention that our definition only works for shapes with connected boundary. For otherwise the geodesic distances are undefined between points on different connected components of the boundary.

There are pathological cases where our algorithm fails to extract a curve-skeleton though our definition provides a 1D curve. In these cases, although the boundary of the shape is connected, its medial axis contains a closed surface preventing the erosion process to proceed. One such example can be derived from the famous “house with two rooms” [Hat02], which is a contractible two dimensional subspace of $\mathbb{R}^3$. A small thickening of this “house with two rooms” creates a shape whose medial axis contains a closed surface.

It would be interesting to apply the curve-skeletons extracted by our algorithm in various applications such as shape matching, mesh decomposition, and animation. We plan to address this issue in future work.

8. Acknowledgement

We acknowledge the support of the NSF grants DMS-0310642 and CCR-0430735. We thank AIM@SHAPE, Stanford University, Princeton University, Sumner and Popovic [SP04] and K. Zhou et al [ZHS'05] for providing 3D models, and N. Cornea et al. [CSYB05] for making their code available.

References


Figure 14: (a) and (c) The curve-skeletons extracted by the potential field method. (b) and (d) The curve-skeletons extracted by our algorithm, \( \theta = -0.3 \) for both models.


