Ball-morph: Definition, Implementation and Comparative Evaluation

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(Invited Paper)

Abstract—We define \textit{b-compatibility} for planar curves and propose three ball morphing techniques between pairs of \textit{b-compatible} curves. \textit{Ball-morphs} use the automatic ball-map correspondence, proposed by Chazal et al. [1], from which we derive different vertex trajectories (linear, circular, parabolic). All three morphs are symmetric, meeting both curves with the same angle, which is a right angle for the circular and parabolic. We provide simple constructions for these ball-morphs and compare them to each other and to other simple morphs (linear-interpolation, closest-projection, curvature-interpolation, laplace-blending, heat-propagation) using six cost measures (travel-distance, distortion, stretch, local acceleration, average squared mean curvature, and maximum squared mean curvature). The results depend heavily on the input curves. Nevertheless, we found that the \textit{linear ball-morph} has consistently the shortest travel-distance and that the \textit{circular ball-morph} has the least amount of distortion.

Index Terms—Morphing, Curve Interpolation, Medial Axis, Curve Averaging, Surface Reconstruction from Slices, Ball-map

1 INTRODUCTION

The animation of a planar curve may be specified by drawing the shape of the curve at specific time values. These drawings are called key-frames or simply \textit{keys}. Then, the problem is one of constructing an animation that continuously deforms the curve from one key to the next, while respecting the timing provided. Each segment of the animation between two consecutive keys is a morph and may be addressed independently, if one does not have to enforce derivative continuity across the keys. We explore here a new formulation of such morphs and their automatic construction and animation.

1.1 Problem statement

A variety of techniques have been proposed for computing automatically a morph between two curves \textit{P} and \textit{Q} in the plane (see [2] and [3] for examples). In this paper, we present a new family of three related morphs, which we call the \textit{ball-morphs}, and discuss two related issues: (1) How to compare different morphing solutions and (2) How do the \textit{ball-morphs} introduced here compare to each other and to other morphing approaches.

1.2 Motivation and applications

Morphing is a fundamental tool in animation design where \textit{in-between} [4] frames are produced from a sparse set of key-frames that are often designed by lead artists [5]. Although several successful attempts at automating the construction of in-between frames have been proposed [6], the artist responsible for in-betweening like to have control over correspondence and over the trajectories for selected landmarks or stroke end-points. These specifications are difficult to automate because they involve aesthetic judgement, style guidelines, and context semantics about the relative 3D motions of the strokes and their mutual occlusions.

Once these matching and control trajectories are given, the overall problem is naturally broken into a series of tight in-betweening tasks [7] [8]. These are viewed as tedious and hence are a prime candidate for artist-supervised automation. In most of such tight in-between tasks, the goal is to generate intermediate frames between two reasonably simple and similar curve segments.

The help the artist select the inbetween technique best suited to a particular need, we show and compare several of these techniques to better assess the strength of each. This paper is a modest—although we
hope useful—step in this direction. It may not be the final answer to tight in-betweening for several reasons: (1) The quantitative measures that we use may not reflect artistic concerns. (2) For practical reasons, we compare the proposed ball-morphs to our simple and un-optimized implementations of candidate techniques, and not to state-of-the-art solutions. More effective implementations of these competing approaches may exist. (3) We do not take into account the broader context of the whole animation, but instead focus on interpolating only the instances of the same stroke in two consecutive key-frames. Nevertheless, we hope that the experiments described here are useful and that the conclusions we draw from them about the specific benefits of the ball-morphs will help the reader appreciate their potential.

Furthermore, the problem (encountered in the segmentation of medical scans) of constructing a surface in 3D that interpolates between each pair of consecutive planar cross-sections may be solved [9] using the morphing between the projection, onto the same plane, of the two cross-section curves. This problem of surface reconstruction has been studied extensively [10][11][12][13][14]. Hence, we have included quality measures of the resulting surface in our set of metrics. As it was the case for tight in-betweening, our investigation of the benefit of ball-morphs to the problem of cross-section interpolation has limitations. For example, it only considers two consecutive slices, instead of building a smooth surface through the whole series, as proposed in [15]. However, because the circular ball-morph reaches the interpolated contours at right angles, the projection of these trajectories on the slice plane is $C^1$. We expect that this property may help researchers devise solutions that smoothly connect surface sections generated by ball-morphs. Furthermore, the approach is limited to b-compatible curves and hence is not suited for dealing with topological changes, as discussed for example in [16].

In these applications, the quality of the morph is important as one typically favors a solution where the animation or interpolating surface is smooth and free from self-intersections [17] and of unnecessary distortions. We show that when the curves are b-compatible, the ball-morph always satisfies these properties.

1.3 Contributions

We propose a family of three new morphing techniques (that we call ball-morphs) for which the correspondence and the vertex trajectories are both derived from the maximal disks and their tangential contact points with the curves.

We propose six cost measures for comparing morphs: travel-distance, distortion, stretch, local acceleration, average squared mean curvature, and maximum squared mean curvature.

We use these measures to compare the three ball-morphs to each other and also to a benchmark of five simple morphing techniques which we have implemented: linear-interpolation, closest-projection, curvature-interpolation, laplace-blending, heat-propagation.

1.4 Limitation

Our ball-morph constructions assume that the two curves have been registered and are sufficiently similar. We provide a formal definition of compatibility that captures these assumptions for the two situations considered here:

1) $P$ and $Q$ are each a simple closed loop.

2) $P$ and $Q$ are open curve segments and share the same two end-points.

Loosely speaking, our compatibility conditions require that each maximal disk [18] in the finite region bounded by the union of the two curves have exactly one contact point with each curve (see Fig. 2). Note that curves with concave sharp features relative to the symmetric difference of their interior are not b-compatible due to this condition.

Where $P$ and $Q$ are similar but not properly registered, one may consider combining a ball-morph with the animation of a rigid or non-rigid registration [19] or of a smooth space warp [20], as was done with other morphs for morphing images [21] and for tight-in-betweening [8]. Numerous solutions to the automatic registration problem have been proposed using ICP [22], automatically identified landmarks [23] [24] [25], or distortion minimizing parameterization [26] [27].

1.5 Structure of the paper

Section 2 briefly reviews prior art in curve morphing and slice interpolation. Section 3 provides a precise definition of b-compatibility and contrasts it with a previously proposed notion of normal compatibility. Section 4 presents our three ball-morphs and compares them to morphs obtained using closest projection and linear trajectories. Section 6 defines our six cost measures and explains our strategy for sampling and for a fair integration of these measures over the set of all trajectories. Section 8 discusses our results.

2 PRIOR ART

A large variety of techniques have been investigated for the automatic generation of in-betweening frames or animations that morph between two planar curves.
We only discuss techniques that are appropriate to the tight in-betweening problem addressed here. Hence, we do not discuss complementary techniques for registration or landmark (salient feature) identification.

First, we review techniques that assume that the correspondence between curve samples (or vertices of polygonal approximations of the curves) on both curves is either given by the artist or computed automatically during preprocessing using, for example, uniform geodesic sampling, minimization of area or travel [13], curvature-sensitive sampling [28], or optimization of matching to affine transformations [29].

If the correspondence is given, the simplest approach is to use a linear interpolation between corresponding pairs. Linear trajectories are computed between these pairs of points on P and Q in order to produce polygonal approximations of the evolving curve, the vertices of which move with time \( t \) as \( v_i = p_i + t(q_i - p_i) \). We refer to this solution as the linear-interpolation.

This naïve approach may lead to unpleasant artifacts, such as self-intersections in the intermediate frames (as, for example, pointed out by [30]). The linear-interpolation is oblivious to the relative orientation and curvature of the curves at the corresponding points.

To take the orientation and curvatures of both curves into account, a popular morphing technique proposed by [31] for polygonal curves interpolates the lengths of corresponding edges and the angles at corresponding vertices and uses optimization to ensure that the curve closes properly. We include a simple version of this approach, which we call curvature-interpolation in our benchmark set. When it is applied to open curves, we ensure that the interpolating frames meet, throughout the morph, at their two endpoints by retrofitting them through a trivial similarity transformation (rotation, scaling, and translation).

A different approach that takes into account the relative orientation and curvature of the two curves at the corresponding samples is to compute the local coordinates of each vertex in the coordinate system defined by its neighbors on each curve. Then, the corresponding local coordinates are averaged linearly to produce a desired set of local coordinates for a given frame. Iterative techniques may be used to construct a curve that satisfies the two endpoint constraints and minimizes the discrepancy between the actual and desired local coordinates. Variations of these techniques have been successfully used [32][33][34]. We include a simple version of this approach, which we call laplace-blending, in our benchmark set.

Vertex trajectories and correspondences may also be obtained by solving a PDE or by computing a gradient field that interpolates the two contours and then following the steepest gradient to obtain the trajectory of each point. Equivalently, the in-between frames may be obtained as iso-contours of that field. A heat propagation formulation may be used to characterize the desired field [35]. We include a simple version of this approach, which we call heat-propagation, in our benchmark set.

Several approaches for morphing closed curves use compatible triangulations [36] of their interior [37][38][3] or compatible skeletons to ensure rigidity [39][40]. Other approaches blend distance fields to both shapes [41][14].

Now, instead of relying on the global optimization or feature recognition techniques discussed above, let us focus on techniques that define an explicit geometric formulation of the correspondence. We separate them into three categories: (1) proximity-based, (2) orientation-based, and (3) both proximity-and-orientation-based.

The most popular distance-based approach is the closest-point projection, which to each point \( p \) on \( P \) maps a point \( q \) on \( Q \) that minimizes the distance to \( p \). Variations of this approach are used for Iterative Closest Point (ICP) registration [42]. We include a simple version of this approach, which we call closest-projection in our benchmark set.

The simplest orientation-based approach is the Minkowski morph [43] and yields satisfying results for convex shapes (see Fig. 3 left), even when the shapes are not aligned. For smooth curves, the approach establishes a correspondence between points with the same normal. Unfortunately, as shown in Fig. 3 (right), the approach may yield surprising, sometimes self-intersecting frames when the two curves are not convex. Hence, we do not include it in our benchmark.

The ball-map [1], upon which the ball-morphs proposed here are based, takes into account both proximity and orientation.

3 Compatibility

3.1 Terminology and notation

Let us start by defining our terminology and notation.

Let \( P \) and \( Q \) be manifold curves in the plane. Recall that a manifold curve is free from self-intersections. When a curve is homeomorphic to a line segment, we call it a stroke. When it is homeomorphic to a circle, we call it a loop. We say that curves \( P \) and \( Q \) are disjoint, when their intersection is empty. We say that they overlap when their intersection contains at least one one-dimensional component. We say that they cross if they are not disjoint and do not overlap. We say that strokes \( P \) and \( Q \) are quasi-disjoint when they only intersect at their two endpoints.

Let \( S \) and \( T \) be arbitrary point sets in the plane. Let \( S^c \) denote the complement of \( S \). We use the notation \( S,i, S,b, \)
and $S.c$ to refer to the topological interior, boundary and closure of $S$. The notation $S.e$ refers to the topological exterior, defined as $S.e = \overline{S.c}$. A set is regularized [44] when $(S.i).c = S$. A set $S$ is finite when there exists a disk of finite radius containing it. Let $S \cup T$ denote their set-theoretic union, $S \cap T$ their intersection, and $S - T$ their difference. Let $S \oplus T$ denote their symmetric difference (XOR), defined as $(S \cup T) - (S \cap T)$.

Consider a one-dimensional subset $B$ of the plane. We say that $B$ is a border when there exists a finite regularized set $S$ such that $B = S.b$. Note that, in that case, $S$ is unique. When $P$ and $Q$ are loops, each one is a border, but the union of two loops needs not be a border (Fig. 4).

**Fig. 4.** Even when the union (right) of overlapping loops $P$ (blue) and $Q$ (blue drawn over orange) is not a border, we define their gap (green).

We define the inside $i(B)$ of a border $B$ as the interior $S.i$ of the finite regularized set $S$ such that $B = S.b$. Similarly, we define the outside $o(B)$ of a border $B$ as the exterior $(i(B)).e$ of the inside of $B$. Note that $i(B)$ and $o(B)$ are topologically open sets. To test whether a point $p$ that is not on $B$ lies in $i(B)$, shoot a ray $R$ (see black arrows in Fig. 9) from $p$ that does not intersect any non-manifold point of $B$ and is not tangent to $B$ anywhere. If $R$ crosses $B$ an odd number of times, then $p$ is in $i(B)$. Otherwise it is in $o(B)$.

Given a closed and regularized [44] set $S$, following [18], we say that a disk in $S$ is maximal if it is not contained in any other disk in $S$ and we define the medial axis as the closure of the union of the centers of maximal disks in $S$.

### 3.2 Topological validity

The set of valid configurations for which the ball-morphs can be computed has both topological and morphological limitations. In this subsection, we address the topological ones. First we present two general restrictions and one simplification.

We orient each curve and ensure that the orientations are compatible. For example, each loop is given a clockwise orientation and the strokes are oriented to have the same starting point. A configuration is invalid when $P$ and $Q$ overlap and have opposite orientations along any portion of the overlap.

When $P$ and $Q$ overlap with compatible orientations, we simplify the validity discussion by removing the overlap segments (except for their endpoints) and by identifying components of the remaining part as matching pairs of separate strokes. During the morph, the overlap segments remain static, hence we need not worry about them anymore. Each stroke of $P$ shares its endpoints with a corresponding stroke of $Q$. We discuss below the morph of a pair of such strokes. From this point throughout the paper, we assume that we have identified and separated the overlap portions and hence that $P$ and $Q$ are not overlapping.

Finally, to avoid further complications, when $P$ and $Q$ are strokes with common endpoints, we require that the oriented loop obtained by combining $P$ with reversed $Q$ ($Q$ for which we have reversed the orientation) has winding number (total number of turns made by the tangent vector as one traverses the loop) equal to one. Hence, configurations such as those in Fig. 5 are excluded.

**Fig. 5.** Invalid configurations with winding number greater than one.

The two restrictions (on reverse orientation overlap and on winding number) and the simplification (remove overlaps) discussed here reduce the number of topological configurations to be discussed. Amongst the remaining ones, only four are valid. We define and illustrate them below.

- **Configuration 1:** $P$ and $Q$ are disjoint loops and the union of their insides is not empty (Fig. 6).

**Fig. 6.** Valid configuration 1: Disjoint loops with overlapping insides.

- **Configuration 2:** $P$ and $Q$ are quasi-disjoint strokes (disjoint except for their shared endpoints) (Fig 7).

**Fig. 7.** Valid configuration 2: Quasi-disjoint strokes, with common endpoints

- **Configuration 3:** $P$ and $Q$ are crossing loops and the intersection $i(P) \cap i(Q)$ of their insides has a single non-empty connected component (Fig. 8).

**Configuration 4:** $P$ and $Q$ are crossing strokes (Fig. 9) and the outside $o(P \cup Q)$ of their union is connected.

Note that we can decompose configurations 3 and 4 into one or more instances of configurations 2. In general, such a decomposition of $P$ and $Q$ into quasi-disjoint strokes may not be desired, since it imposes artificial constraints, forcing the evolving curves to interpolate...
For topological configuration 1, we require that the medial axis of the gap $G(P,Q)$ be a loop. Figure 10 shows a geometrically valid and a geometrically invalid configuration 1.

For topologically configuration 2, we require that the medial axis of the gap $G(P,Q)$ be a strokes with the same endpoints as $P$ and $Q$. This additional precision is necessary to exclude situations such as the one depicted in Figure 11.

For topological configuration 1, we require that the medial axis of the gap $G(P,Q)$ be a loop. Figure 10 shows a geometrically valid and a geometrically invalid configuration 1.

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the set of configurations to which the ball-morph may be applied. In this subsection, we dispel this perception by comparing them to conditions required by the popular closest-projection-morph.

The closest-projection of a point \( m \) onto a curve \( P \) is a set of points \( p \in P \) for which \( \text{distance}(m, p) = \text{distance}(m, P) \).

\( P \) and \( Q \) are \( c \)-compatible (i.e., closest-projection compatible) when every point of \( P \) or \( Q \) has a single closest projection on the other curve. Figure 13 shows a configuration where \( P \) and \( Q \) are \( b \)-compatible but not \( c \)-compatible.

Sufficient conditions for \( b \)-compatibility and \( c \)-compatibility have been discussed in [45], [46] and [1]: \( P \) and \( Q \) are \( b \)-compatible if \( H(P, Q) < h \) and \( c \)-compatible when \( H(P, Q) < ef \), where \( H(P, Q) \) denotes the Hausdorff distance [47] between \( P \) and \( Q \), \( f \) is the smallest of their minimum feature sizes [48], and \( c = 2 - \sqrt{2} \approx 0.5858 \). Note that a significant set of configurations satisfy the condition for \( b \)-compatibility, but not for \( c \)-compatibility. Hence, we argue that \( b \)-compatibility conditions are in fact less restrictive than the \( c \)-compatibility ones.

4 Ball-morphs

In this section, we describe the correspondence used for our ball-morphs and present the various options for ball-morph trajectories between corresponding pairs of points. The definitions are independent of the nature of the two curves and of their representation. We have implemented these techniques for two domains: (1) smooth \((C^1)\) piecewise circular curves \((\text{PCCs})\) [49], and (2) relatively smooth polygons (such as those obtained through smoothing or subdivision). Our implementation on pairs of \( b \)-compatible piecewise-circular curves is numerically precise and yields the theoretically correct ball-morph. Clearly, the implementation for polygons is not theoretically correct. Indeed, the polygonized versions of two \( b \)-compatible curves are not \( b \)-compatible, because the region they bound must have convex vertices and hence bifurcations in its medial axis. Nevertheless, when the polygonal curves are reasonably smooth and densely sampled, our polygonal algorithm computes ball-morphs that closely approximate the ball-morphs of the original smooth curves and are acceptable for animation or surface reconstruction. Because most

Fig. 12. The blue stroke can be expressed as the normal offset (top) or as the ball-offset (bottom) of the orange stroke.

Fig. 13. Two curves that are \( b \)-compatible (left), but not \( c \)-compatible (right).

Fig. 14. To obtain the point \( q \) on \( Q \) that corresponds, through the ball-map, to point \( p \) on \( P \), we compute the smallest positive \( r \) such that \( m = p + r\bar{N}_P(p) \) is at distance \( r \) from \( Q \) and return its closest projection \( q \) on \( Q \). Point \( m \) is on the median and defines the center of the circle tangent at both \( p \) and \( q \). The circular (black) and parabolic (purple) ball-morph trajectories are defined by the inscribing isosceles triangle \( \Delta mnq \). The linear ball-morph trajectory is the line segment \( pq \). Other morphing schemes to which we compare our ball-morphs work on polygonal curves, we use the polygonal ball-morph implementation to ensure consistency in our experiments.

4.1 Ball-map correspondence

Consider the maximal disk centered at point \( m \in M \). The ball-map [1] establishes the correspondence between the closest projection \( p \) of \( m \) onto \( P \) and the closest projection \( q \) of \( m \) onto \( Q \). The maximal disk \( D \) centered at \( m \) touches \( P \) at \( p \) and \( Q \) at \( q \), as shown in Fig. 14. The ball-map may be viewed as a continuous version of an approach proposed by [41] for establishing correspondences between surfaces by considering their distance fields. A uniform sampling of the ball-map correspondence may be computed in several ways: (1) By initially computing \( M \) as the medial axis of \( G(P, Q) \) between two curves \( P \) and \( Q \) using efficient medial axis construction techniques [50][51] and then generating the closest projections \( p \) and \( q \) for a set of uniformly spaced sample points \( m \in M \); (2) By computing the radii of the maximal disks that touch \( P \) at a set of uniformly spaced samples \( p \); or (3) By simultaneously advancing the corresponding points, \( p \) and \( q \), until one of them has travelled from the previous sample on its curve by a prescribed geodesic distance. To ensure a fair comparison with techniques that lack the symmetry of the ball-morphs, we will use the second (asymmetric)
approach, although the first one yields the best results.

The details of the construction of this mapping for the case when \( P \) and \( Q \) are piecewise-circular and when they are polygonal approximations of smooth curves are provided below.

### 4.2 Ball-morph trajectories

For each maximal disk, we consider five paths (curve segments) from \( p \) to \( q \) (Fig 14):

- **Hat:** The broken line segment from \( p \) to \( m \) to \( q \) (Fig 14 green).
- **Linear:** The straight line segment from \( p \) to \( q \) (Fig 14 yellow).
- **Tangent:** The shorter of the two circular arc segments of the boundary of \( D \) that joins \( p \) and \( q \) (Fig 14 green).
- **Circular:** The circular arc segment that is orthogonal to \( P \) at \( p \) and to \( Q \) at \( q \) (Fig 14 green).
- **Parabolic:** The parabolic arc segment that is orthogonal to \( P \) at \( p \) and to \( Q \) at \( q \) (Fig 14 green).

The circular and parabolic paths are trivially defined by their enclosing isosceles triangle \( \Delta \text{P} \text{m} \text{q} \). The parabolic path is the quadratic Bézier curve with control vertices \( p \), \( m \) and \( q \). The center of the circle supporting the circular path is the intersection of the tangent to \( P \) at \( p \) and the tangent to \( Q \) at \( q \).

All paths, including the linear path, are symmetric in that the angles where they meet \( P \) and \( Q \) are equal. Swapping the role of \( P \) and \( Q \) does not affect these segments. Hence, the ball-morphs derived here are symmetric and may be inverted easily by swapping the role of \( P \) and \( Q \).

Let \( I \) be the midpoint of the linear path and let \( L \) be the set of all points \( I \). \( L \) is the midpoint locus proposed by Asada and Brady [32]. Let \( T \) be the midpoint of the tangent path and \( T \) be the set of all points \( t \). \( T \) is the Process-Infering Symmetry Axis (PISA) proposed by Layton [53] as a variation of the medial axis. Let \( N \) be the midpoint of the circular path and \( N \) be the set of all points \( N \). Let \( b \) be the midpoint of the parabolic path (quadratic B-spline) and \( B \) be the set of all points \( b \). The construction of these 4 points, along with \( m \) is illustrated in Fig. 14. The curves \( M \), \( L \), \( T \), \( N \) and \( B \) usually differ from one another, but may all be viewed as averages of \( P \) and \( Q \).

A ball-morph advances, with time, each point \( p \) according to uniform arc-length parameterization along one of the five aforementioned paths. A result for the circular ball-morph is shown in Fig. 1 using seven inbetween frames.

### 5 Implementation details

In this section, we provide implementation details for PCCs and then for polygonal curves.

#### 5.1 Details of the ball-map construction for PCCs

We include here the details of an exact implementation (except for numerical round-off errors) for the case of piecewise-circular curves in 2D, where \( P \) and \( Q \) are each a series of smoothly connected circular-arc edges.

We first explain how we compute the ball-map of a sample point. Then, we explain how to produce these samples and how to reduce the computational complexity of the whole process. We sample points on one curve and for each such sample, say \( p \) on \( P \), we compute the corresponding point \( q \) on \( Q \) as explained below. Here we assume that \( p \) is not on \( Q \), as discussed above.

Consider the parameterized offset point \( m = r \bar{N}_P(p) \), whose distance from \( p \) is defined by the parameter \( r \). \( \bar{N}_P(p) \) is the normal of \( P \) at \( p \) and it is oriented so that it points towards the interior of the gap \( G(P,Q) \). By construction, \( m \) is the center of a circle of radius \( r \) that is tangent to \( P \) at \( p \). We want to compute the smallest positive \( r \) for which \( m \) is at distance \( r \) from \( Q \), and hence for which the circle is tangent to \( Q \).

First consider a circular edge \( Q_i \) of \( Q \) with center \( c \) and radius \( s \) (Fig. 15). We compute \( r_1 \) and \( r_2 \) as the roots

\[
\frac{s^2 - \bar{c}^2}{2\bar{N}_P(p) \cdot \bar{c}^2 \pm 2s} \\
\]

of

\[
\bar{c}^2 = (r \pm s)^2.
\]

In order for \( m \) to define the center of a disk of radius \( r \) that is tangent to \( Q \), \( m \) must have either a minimum or maximum distance of \( r \) from \( Q \), or in other words, \( m \) must be at distance \( r \pm s \) from \( c \), as defined by Equation 2. Substituting \( m = r \bar{N}_P(p) \) and expanding yields a second degree equation in \( r \), the roots of which are given by Expression 1.

We apply the above approach to all edges \( Q_i \) of \( Q \). We compute the \( r \)-value for a circle supporting each edge, compute the corresponding candidate point \( q \) on the circle, discard it if it is not contained within the arc (such as \( q_2 \) in Figure 15), and select amongst the retained \((r,q)\) pairs with the one with the smallest \( r \)-value.

Since we assume that \( P \) and \( Q \) are \( b \)-compatible, there is exactly one \((r,q)\) pair for each point \( p \in P \).

The above process computes the ball-map correspondence for any desired sampling of \( P \) or \( Q \).

To accelerate the computation of the ball-map for \( b \)-compatible PCCs and produce a sampling-independent representation from which different sampling densities
which is convenient for the interactive design of curves. For disjoint loops, the lacing starts at any vertex of $P$ and terminates when it returns to the starting point. For quasi-disjoint strokes, the lacing starts at one common endpoint and finishes at the other endpoint. A small variation of this approach, described in [54], permits in linear time to either perform the lacing when the two curves are $b$-compatible or to detect that they are not.

5.2 Details of ball-map construction for PLCs

Because they are not smooth, piecewise-linear curves cannot be $b$-compatible. However, we propose here an approach for computing an approximate ball-map, treating piecewise-linear curves as approximations of smooth curves. We treat vertices as circular arcs with infinitely small radii and ignore incompatibilities as long as adjacent ball-maps do not intersect.

First we show how to compute the ball-map from a point $p$ on $P$ to a polygonal curve $Q$ by computing the radius of a circle that is tangent to $P$ at $p$ and touches $Q$ at a vertex or an edge. Assuming $p$ is not a point of intersection between $P$ and $Q$. We estimate the normal $\hat{N}_P(p)$ to $P$ at $p$ such that it is orthogonal to the line passing through the two neighboring vertices of $p$ along $P$ and that it points towards the interior of the gap. First, for every vertex $q$ of $Q$, we compute the $r$-value for a ball with center $m = p + r\hat{N}_P(p)$ such that $|m - p| = |m - q| = r$ as follows:

$$r = -\frac{\mathbf{p}^2}{2\mathbf{p} \cdot \hat{N}_Q(q)}$$

(3)

Notice that we do not need to check explicitly that the direction of vector $\mathbf{q}_m$ is a plausible normal to $Q$ at $q$. If it were not, the ball-map for edges of $Q$ incident upon $q$, computed as explained below, would return a smaller radius.

For every edge $Q_i$ of $Q$ with oriented edge-normal $\hat{N}_Q(Q_i)$ and vertices $c, d$, we compute the $r$-value for a ball with center $m = p + r\hat{N}_P(p)$ as follows:

$$r = \frac{\mathbf{c} \cdot \hat{N}_Q(Q_i)}{1 - \hat{N}_P(p) \cdot \hat{N}_Q(Q_i)}$$

(4)

The corresponding point $q$ is then computed as:

$$q = p + r\hat{N}_P(p) - r\hat{N}_Q(Q_i)$$

(5)

The candidate mapping is discarded if $q$ lies outside the bounds of $Q_i$.

The minimum $r$ ball-map candidate among all vertex-vertex and retained vertex-edge mappings is then selected for point $p$.

For conciseness, we omit the discussion of singular configurations where the denominators of these equations are 0. We trap these using a numeric tolerance and use trivial formulations for the corresponding singular (parallel or coincident) cases.

The ball-map construction for PLCs presented above may fail when the curves are insufficiently smooth or
when the sampling process is not sufficiently dense. Our experiments show that the application of a few simple subdivision steps [55] produces curves for which our construction works without problem. In fact, all results in this paper were computed using the PLC method described here unless explicitly stated otherwise. When working on PLCs, we use a modified version of lacing.

Recall that in the PCC lacing algorithm, two points \((p, q)\) “walk” on each curve, where the maximum step-size is determined by the length of the current circular-arc. Instead, we now set a constant step-size \(d\text{Step}\). Therefore at each step, \(p'\) and \(q\) are computed by walking along \(P\) and \(Q\), respectively, a geodesic distance of \(d\text{Step}\). Our experiments show that using a step-size that is twice larger than the longest edge of the subdivided curves works well.

6 MEASURES

We first discuss how we sample space and time. Then, we provide details of the measures used here to compare morphs.

Three of the studied morphs (linear-interpolation, curvature-interpolation, and laplace-blending) assume a given correspondence. For simplicity, we use a uniform arc-length sampling to produce the same number of uniformly distributed samples on each curve. The three ball-morphs use the ball-map correspondence. The other morphs compute their own correspondence. This sampling disparity makes it difficult to compute measures for a fair comparison.

Consider for example the problem of measuring the average travel-distance. This should be the integral of travel distances. The problem is how to fairly select the integration element. If for example we use the linear-interpolation-morph, then the average distance measured for a set of uniformly distributed samples will depend on whether we start form \(P\) or \(Q\). Since the average travel-distance is a property of the mapping, and not the sampling, a measure that so blatantly depends on the sampling is clearly incorrect.

To overcome this problem, each reported measures is the average of two measures, one computed by sampling \(P\) and one computed by sampling \(Q\). For the first measure, we sample the departure curve \(P\) using a dense set of samples that are uniformly distributed on each curve so as to be separated by a prescribed geodesic distance \(u\). For each sample \(p_i\) on \(P\), we compute the corresponding point \(q_i\) on the arrival curve \(Q\) so that \(q_i\) is the image of \(p_i\) by the mapping associated with the particular morphing scheme. We compute a measure \(m_i\) associated with the trajectory from \(p_i\) to \(q_i\) and the associated weight \(w_i = (|p_i| - |p_i - p_{i+1}| + |q_i - q_{i+1}| + |p_i - q_{i+1}| + |q_i - p_{i+1}|)/4\). Then, we report the normalized weighted average \((\Sigma w_i m_i)/(\Sigma w_i)\). For the second measure, we sample the arrival curve \(Q\) as before using the same geodesic distance \(u\). For each sample \(q_i\) on \(Q\), we compute the corresponding point \(p_i\) on the departure curve \(P\), so that \(q_i\) is the image of \(p_i\) by the mapping associated with the particular morphing scheme. Then, we proceed as above.

We have implemented the following six measures of morph quality.

Travel-distance: For each sample \(p_i\), we measure \(m_i\) as the arc length of the trajectory to the corresponding point \(q_i\). Then, as explained above, we report the weighted average of these from \(P\) to \(Q\) and vice-versa.

Stretch: We define stretch \(S(P, Q)\) as the average of the integral over time of the stretch factor for an infinitesimal portion of the curve. We compute its discrete approximation as follows. Let \(p\) and \(p'\) be consecutive samples on \(P\). Let \(L(p, t)\) be the length of the segment of \(P(t)\) between \(p(t)\) and \(p'(t)\). We compute \(S(P, Q)\) as

\[
S(P, Q) = \sum_{t \in [0, 1]} \left( \sum_{p \in P} |L(p, t + \epsilon) - L(p, t)| \right) + \sum_{t \in [0, 1]} \left( \sum_{q \in Q} |L(q, t + \epsilon) - L(q, t)| \right)
\]

Acceleration: Acceleration is defined as the derivative of the expression of velocity in the local, time-evolving frame, and measures the lack of steadiness [19] of the motion.

Let \(p_i\) denote the position of a sample \(p\) at a time \(t\). We approximate the instantaneous velocity of \(p_i\) by the vector \(p_{i+\epsilon} - p_i\). For each such velocity on a morph trajectory, we compute two barycentric coordinate vectors \(B_L(p_i)p_{i+\epsilon}\) and \(B_R(p_i)p_{i+\epsilon}\) relative to the left and right neighboring triangles \(L_i\) and \(R_i\) as shown in Fig. 18. The steadiness at a point \(p_i\) is then computed as:

\[
g_i = \frac{1}{2} \| B_L(p_i - p_{i-\epsilon}) - B_R(p_{i-\epsilon} - p_i) \| + \frac{1}{2} \| B_L(p_i - p_{i+\epsilon}) - B_R(p_{i+\epsilon} - p_i) \|
\]

We compute the acceleration measure \(m_i\) as the sum of the \(g_i\) terms over the trajectory of each point \(p_i\) and report their weighted average, as described above.

Distortion: At each point along the evolving curve and at each time, the amount of distortion is proportional to \(1/cos\theta\), where \(\theta\) is the angle between the direction of travel and the normal to the evolving curve. Let \(p\) and \(p'\) be consecutive samples on \(P(t)\) and \(L(p, t)\) define the unit vector in the direction \(pp'\). Let \(V(p, t)\) define the unit vector in the direction \(pp_{i+\epsilon}\). We compute

\[
r_i = \sum_{t \in [0, 1]} \frac{1}{2} \left( L(p, t) + L(p, t + \epsilon) \right) \cdot \frac{1}{2} \left( V(p, t) + V(p', t) \right)
\]

where \(\cdot\) denotes dot product.
It was shown in [1] that the circular ball-morph is free from distortion when morphing between linear segments (in 2D) of \( P \) and \( Q \) (Fig. 19).

6.1 Mesh measures

In addition to the 2D measures, we also present results of 3D measures of average squared mean curvature and maximum squared mean curvature [56] of the resulting triangle mesh surfaces constructed by interpolating the input curves along the \( z \)-axis. In applications of surface reconstruction from 2D planar contours, smoothness of the resulting reconstruction is often desirable (see Fig. 20).

7 Composite ball-morphs

To produce morphs between curves that are not \( b \)-compatible, we compute the relative blendings [57] \( P' = R_Q(P) \) and \( Q' = R_P(Q) \) and the ball-morph \( M_0 \) between the resulting curves \( P' \) and \( Q' \). The relative blendings of two curves \( P \) and \( Q \) are computed by trimming away the parts of the curve that violate the conditions of \( b \)-compatibility and replacing them with circular arcs defined by maximal disks in the gap \( G(P, Q) \).

This solution produces only the central part of the morph, which must be concatenated in time and possibly on both ends with other extension-morphs that fill the incompatible features. There are 3 situations for computing the next extension-morph (morph \( M_1 \)):

1) If \( Q' \) and \( Q \) are \( b \)-compatible we compute their ball-morph \( M_1 \).

2) Else if the closest-projection-morph from \( Q \) to \( Q' \) is a homeomorphism, we produce an extension-morph \( M_1 \) with the reversed straight line trajectories (Figure 21-left)

3) Otherwise, we compute the relative blending \( Q'' = R_{Q'}(Q) \), then compute the ball-morph \( M_1 \) between \( Q' \) and \( Q'' \).

Note that in all cases, the trajectories of \( M_1 \) leave \( Q' \) along its local normal and are hence smoothly joined with the trajectories of \( M_0 \). If we run into situation (3), we recurse on the gap between \( Q'' \) and \( Q \). This process fills the gap between \( Q' \) and \( Q \) by a series of ball-morphs and possibly a final closest-projection-morph, generating piecewise-circular trajectories (Fig. 21 right) with possibly a straight line at the end of each one. Note that without handling situation (2), the recursive process would never converge since a ball will never reach the sharp feature. We have produced a concatenation of morphs \( M_1, M_2, M_3... \). We perform a similar iterative process to invade the gap between \( P' \) and \( P \), producing in this manner a series of morphs, which we reference with negative integers: \( M_{-1}, M_{-2}, M_{-3}... \). The final combined morph is: \( ... M_{-3}, M_{-2}, M_{-1}, M_0, M_1, M_2, M_3 ... \).

Fig. 22 shows a simple example which produces a composite of four circular ball-morphs.

Although different synchronization approaches are possible, the simplest one is to move each sample at
a constant speed along its PCC trajectory during the desired interval. Note that smooth inbetween curves are not generated when using the composite ball-morph. However, the inbetween curves and trajectories are similar to those produced by the heat-propagation-morph, as shown in Fig. 23.

8 Results

We first compare the ball-morphs to our benchmark set using two different test cases, as shown in Figures 24 and 25. Then, we compare two of the ball-morphs to the best two other morphs (laplace-blending and heat-propagation) on a test case between an apple and a pear. The closest-projection-morph is not shown for these tests because the pairs of curves are not c-compatible (Fig. 13).

The first test case (Fig. 24) shows a morph between a circle and an ellipse. Our experiments demonstrate that the average travel-distance is the shortest when using the linear ball-morph and that the circular ball-morph has the least amount of distortion. Note that the heat-propagation-morph is similar in terms of appearance to the circular ball-morph. However, due to its reliance on a discrete grid and other sampling issues, it is very susceptible to acceleration and squared mean curvature errors in regions where $P$ and $Q$ are very close relative to the chosen grid size. The minimum squared mean curvature measures (maximum and average) are produced by the curvature and laplace-blending-morphs.

The test case shown in Fig. 25 shows a set of symmetric ‘S’ shaped curves. This example highlights the strength of the morphs which compute their own correspondence (heat-propagation, ball-morphs). The other morphs, which define correspondence through uniform arc-length parameterization, exhibit extreme distortion and travel lengths and also produce self-intersections with the original curves. As with the previous example, the travel-distance and distortion measures are minimized by the linear and circular ball-morphs, respectively. The heat-propagation-morph is very similar in terms of appearance and measure to the family of ball-morphs and produces meshes with the smallest values of maximum and average squared mean curvature.

The test case in Fig. 26 uses contours representing an apple and a pear. We show the “best” four morphs (linear ball-morph, circular ball-morph, heat-propagation and laplace-blending) and compare their measures. Travel distance and distortion are still minimized for the linear and circular ball-morphs, respectively. The laplace-blending approach performs best in terms of stretch. The ball-morphs perform the best in terms of acceleration and the worst in terms of squared mean curvature of the resulting meshes.

The parabolic ball-morph is barely distinguishable from the circular one, even though the surface it produces has a higher maximum squared mean curvature. It may be preferred in some applications, where the non-rational quadratic parameterization of the trajectories simplifies numeric calculations.

9 Conclusion

We have proposed a family of morphs between curves which are b-compatible. All are based on variations of the medial axis construction. We have compared them to one another and to several other simple morphs. We used four measures of morph quality in our comparison, as well as surface measures for comparing them as surface reconstruction techniques.

Although the heat-propagation-morph produces very similar results to the ball-morphs, it has the disadvantage of requiring rasterization to a grid and a PDE solve. However, this method easily maps to more extreme cases that are not b-compatible without need for special extensions (other than a higher resolution grid).

We conclude that for the cases of b-compatible shapes, the ball-morphs offer a precise and desirable result in terms of distortion, travel-distance, as well as curvature.

Ball-morphs have many advantages. For example [1], the circular ball-morph produces curves of $C^{k−1}$ continuity that do not intersect one another for input curves of $C^k$ for $k ≥ 2$.

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References


Fig. 24. Morph results for a circle and ellipse, showing the morph curves (top), the morph trajectories (middle) and the surface created by sweeping the evolving curve, changing its height at a constant rate (bottom). Also displayed are the measures for each morph with the smallest (best) value of each measure highlighted in orange.

Fig. 25. Morph results for a set of ‘S’-shaped curves, showing the morph curves (top), the morph trajectories (middle) and the surface created by sweeping the evolving curve, changing its height at a constant rate (bottom). Also displayed are the measures for each morph with the smallest (best) value of each measure highlighted in orange. Note that some of the morphs do not remain within the bounds of the input curves.
Fig. 26. Morph results for a set of apple and pear shaped curves, show the morph trajectories (top) and the surface created by sweeping the evolving curve, changing its height at a constant rate (middle). Also displayed are the measures for each morph with the smallest (best) value of each measure highlighted in orange.


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