Length Constrained Multiresolution Deformation for Surface Wrinkling
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Abstract

We present a method for deforming piecewise linear 3D curves with constant length constraint. We show how this constraint can be integrated into a multiresolution editing tool allowing an intuitive control of the deformation’s extent and aspect. The constraint is enforced following two steps. A first step consists in approximating the initial length by modifying the multiresolution decomposition at some specified scale. In a second step the constraint is exactly enforced by constrained minimization of a smoothness criterion. This process then provides the core of an integrated wrinkling tool for soft tissues modelling. A curve on the mesh is deformed, providing a deformation profile which is propagated in a user-defined neighbourhood on the surface.

1 Introduction

Multiresolution (MR) analysis has received considerable attention in recent years in many fields of geometric modeling, computer graphics and visualization. It provides an efficient representation of complex functions at multiple levels of detail. Thus convenient handling of geometrical objects is possible. General multiresolution editing or deformation techniques for parametric curves have been explored in detail by Finkelstein and Salesin [13], Gortler and Cohen [15], Elber and Gotsman [11] using B-splines. In contrast, constrained multiresolution editing techniques have not been explored so much in the past. However, there are many application areas, including CAD/CAM, computer graphics and computer animation, where deformations under constraints are needed. Constraint enforcement offers additional control of a shape during the modeling process. Constraining linear geometric properties, like position, normal and point tangent combined with minimizing curve or surface energy [14, 6, 34] proves to be an effective tool both for sculpting models and for animating real behaviours of objects. Some other works deal with non-linear constraints, e.g. length preserving deformation of Bézier curves [27], prescribing the length of rational Bézier curves [29], and volume preserving deformation of solids [28] and implicit surfaces [8]. None of these works integrate constraints into multiresolution editing. Only recently, MR editing of closed planar curves, coupled with the constraint of enclosed area preservation has been developed [10, 19]. In [5] a specific MR representation is proposed for volume preserving surfaces. It is based on the encoding volume elements between levels instead of basis function coefficients. While area and volume are bi-linear and tri-linear functionals, the length constraint in contrary is non-linear. This is the reason why for example the MR area preserving methods [10, 19] can’t be directly adapted to the length. One of the purposes of the present paper is therefore to show how the non-linear constraint of length preservation can be incorporated into an MR editing tool.

Furthermore, the length constraint as well as constant area or constant volume belongs to the traditional principles of animation [22]. For example in computer animation of soft objects, where realistic deformations are required, realism can be achieved thanks to physical-based models [33, 7]. However, when low computation time has priority, typically for real-time animations, it is quite advantageous to mimic physical laws by some appropriated geometric properties. Area or volume preserving deformations can simulate inelastic material [28]. Length preserving deformation can mimic nearly non-compressible or non-stretch material like cloth [1, 9] or soft tissues. A characteristic behaviour of soft tissues is to form wrinkles. Wrinkles appear and disappear in order to compensate length changes. Therefore the second goal of the present paper is to handle length preserving curves specifically in order to control the generation of wrinkles through the MR representation when two curve points get closer.

Several models dedicated to wrinkle generation have been developed in the domain of computer graphics and computer animation. The first work [2] introduced bump mapping, that have often been used later on. The main
drawback of bump mapping is that it doesn’t modify the geometry of the object. Later works [37, 3] deal with static wrinkles for skin modelling which is significantly different from the problem of dynamic wrinkles this paper focuses on. Static wrinkles are present on the skin and they don’t change when the skin is moving. In contrast, dynamic wrinkles are the product of deformation of the tissues. They typically appear and disappear to absorb the length changes. Modelling dynamic wrinkles with geometric deformations has been handled in [35, 37, 38] but all these methods require the ridges to be drawn by the user.

In [17] Hadap et al developed a method for simulating dynamic wrinkles on meshes. A displacement map is built in order to preserve the surface area: the underlying deformation of a coarse mesh is used for modulating pre-defined wrinkle patterns. Though it is convincing for cloth rendering it suffers drawbacks for a general use: the patterns have to be drawn by the user, and the geometry is modified through the displacement mapping extra layer.

Jing et al [20] presented a wrinkling tool for footwear design. It is based on the propagation of a wrinkling profile defined on the border of nurb patches. The main drawback, relatively to our aims, is a tessellation step that prevents this method from being used for dynamic wrinkles. Moreover, only one single deformation is possible because the initial geometry is lost.

Larbulotte and Cani [21] propose to create wrinkles at different scales on a mesh by normal displacement. The mesh is first projected on some mean-tangent plane. An height-function is built on this plane by the propagation of a 2-dimensional profile curve. Then this function weights the normal displacement of the vertices. But the profile is independent of the mesh’s geometry, leading to very regular deformations. Moreover the projection on a plane hinders the use on high-curvature regions.

The contribution of this paper is to provide a multiresolution editing tool for 3D piecewise linear curves which allows to satisfy the non-linear constraint of length preservation. Beside general deformations through select-and-drag control points, the present paper focuses on a particular type of deformation: the generation of dynamic wrinkles in case of compressing the curve or part of it in order to absorb the length changes. The approach is purely geometric in the sense that the scale and frequency of the generated wrinkles are not controlled by some physical law, but by modifying curve coefficients at different multiresolution levels. Then we show how it can be used for general surface deformation through the extraction of a profile curve on the surface. Our algorithm is then applied to this curve. Eventually surface wrinkles are created by propagating and by attenuating smoothly the wrinkles on both sides of the curve.

The paper is organized as follows. Section 2 sets the basic tools and details the length preserving deformation method for 3D curves. In Section 3 the application to surface wrinkling is presented. Eventually, Section 4 concludes and gives possible future works.

2 3D curves

In a classical MR editing environment the user chooses a resolution level at which the curve is modified by displacing one coarse control point. The shape of the curve changes more or less locally depending on whether a low or a high resolution level has been chosen.

The same procedure can be applied when integrating the additional constraint of keeping the curve length constant during deformation. The contribution of the present section is a method for solving the length constraint in a MR editing environment. It is in particular used for creating wrinkles whose frequency can be controlled by the user through the MR representation of the curve. This method is a generalization of the 2D method presented in [30], with a specific solution of the 3D orientation problems.

In a first time we present the length measure and constraints in Section 2.1. It is followed in Section 2.2 by an overview of the method whose two main steps are detailed in Sections 2.3 and 2.4.

2.1 Curve length

Let \( c(t) = (x(t), y(t), z(t)) \) be a parametric curve lying in a space having a multiresolution analysis (see Appendix A for details). The length of \( c(t) \) is given by
\[
L = \int \sqrt{(x'(t))^2 + (y'(t))^2 + (z'(t))^2} dt.
\]
The curves the present paper is dealing with are continuous and piecewise linear. In that case the length simplifies to
\[
L = \sum_{i=0}^{N-2} \| c_{i+1}^n - c_i^n \|_2,
\]
where \((c_i^n)_{i=0,...,N-1} = (x_i^n, y_i^n, z_i^n)\) denote the control points at the finest level \( n \) following the notations of Appendix A.

For the multiresolution representation of piecewise linear curves the scheme based on the Lazy wavelets [32] (see Fig. 2) is used:
\[
\begin{align*}
\begin{cases}
  c_{i+1}^j &= c_{2i+1}^{j+1} \\
  d_{i+1}^j &= c_{2i+1}^{j+1} - \frac{1}{2} (c_{2i+1}^{j+1} + c_{2i+2}^{j+1}).
\end{cases}
\end{align*}
\] (1)
where \( c_i^j \) and \( d_i^j \) are the coefficient of the MR curve at level \( j \) (see Eq. 8).
The reasons for the particular choice of working only with piecewise linear curves are threefold. First, the length preserving deformation method presented here is intended to work dynamically. Linear curves in contrast to higher order polynomial curves simplify and accelerate the length computations. Second, the particular effect of wrinkle creation in response to the motion of control points is a purely visual effect. Therefore, on a computer screen no visual difference can be observed between displaying a dense polygonal curve or a rasterrized polygonal curve. Third, the algorithm will be used in Section 3 to create dynamic surface wrinkles, the surface being represented as a fine triangular mesh. Thus all curves on the surface are piecewise linear.

In the case of piecewise linear curves one can choose either to keep the total length constant or to preserve the length of each segment. We choose the second way because of two main reasons:

- It ensures the balance between segment’s length that is to say the control points don’t gather in a small part of the curve.
- It allows the length constraints to be expressed in such a way that computationally inefficient square roots evaluations can be avoided.

The length constraints on a deformed curve defined by \( C^n = (c^i_n)^T \) can now be reformulated as follows:

\[
f_i(C^n) = \Delta x_i^2 + \Delta y_i^2 + \Delta z_i^2 - l_i^2 = 0; \quad i = 0, \ldots, N-2,
\]

where \( l_i \) is the reference length of segment \( [c_i; c_{i+1}] \) before deformation, \( \Delta x_i = x_{i+1} - x_i \), \( \Delta y_i = y_{i+1} - y_i \) and \( \Delta z_i = z_{i+1} - z_i \).

### 2.2 Overview of the deformation method

We present here how to deform a 3D curve while enforcing the previous length constraints. The multiresolution representation of the curve is at the center of this process. It provides a friendly manipulating tool for the user: the curve can be deformed globally through a few coarse control points while preserving the details. Moreover the use of several decomposition levels allows to precisely control the deformation scale at each step of the process.

Three levels of decomposition (corresponding to the three columns in figure 1) are involved: the finest level \( n \), the editing level \( e \) and an intermediate level \( w \) corresponding to the wrinkling scale if required by the length constraint. Hence \( C^1 \) denote the gathering of the MR coefficients \( e^1, d^1, d^{1+1}, \ldots, d^{n-1} \) (see Appendix A). The deformation process follows the loop of figure 1. It can be iterated several times, basically at each time step in a select-and-drag editing process. During the deformation process the curve undergoes several transitions and thus admits different geometric stages i.e. different values of the MR coefficients. Four different stages of \( C \) are involved, labeled by a different index:

- \( C_R \) is the initial curve and also the reference curve for the length;
- \( C_D \) is the deformed curve, i.e. after it has been edited and before any length preservation action;
- \( C_A \) is the attracting curve that originates from the explicit length preservation step;
- \( C_F \) is the final curve. It will replace \( C_R \) for the next deformation.

![Figure 1. Editing loop: each box is a state of the curve; each horizontal transition is a change in the decomposition level; each vertical transition is a modification of the curve.](image)

Let us describe the role of each transition in figure 1:

- **Decomposition**: The curve is decomposed into \( C_R^e \) at the level \( e \), chosen by the user, following equation 1.
- **Edition**: The coarse control polygon \( C_R^w \) a level \( e \) is modified by the user. The corresponding deformed curve \( C_D \) has not the same length as the reference curve \( C_R \). Notice that the choice of level \( e \) determines the extent of the deformation.
- **Reconstruction**: The deformed curve is then partially reconstructed at level \( w (e \leq w < n) \), chosen by the user to be the wrinkling scale.
- **Explicit length preserving**: An attracting curve \( C_A \) is constructed. This curve is obtained by modifying \( C_D^w \) at level \( w \) using the details in order to obtain a control polygon \( C_A^{w+1} \) whose edges have the same length as the control polygon segments of \( C_R \) at level \( w+1 \). This step is detailed in Section 2.4. Modifying details at level \( w \) is equivalent to modifying corresponding control points at level \( w+1 \). The choice of an intermediate level \( w \) different from \( e \) for length approximation increases the number of coarse control points defining the same portion of the curve. The closer to the highest level \( n \) is \( w \), the higher is the number of control.
points for length approximation, and the higher is the frequency of the wrinkles in case of need. If the curve is stretched \( w \) acts as a kind of stiffness parameter.

- **Reconstruction:** The attractive curve is then completely reconstructed by re-inserting the details of \( C_R \) at the levels \( w + 1, \ldots, n \). After reconstruction, the length of \( C_A \) is close to satisfy the length constraint. Note that the use of a lazy wavelet scheme may lead to some sharp features in the reconstructed curve. Hence a smoothing is useful.

- **Length preserving by smoothing:** An optimization method applied to \( C_A \) leads to the final curve \( C_F \). It precisely satisfies the length constraints, has a smooth shape, and it is close to the attracting curve \( C_A \). This step is described in Section 2.3.

2.3 Optimization step

We detail here the step of exact length enforcing via optimization. After presenting the basic ideas and the calculus we explain why it is useful to combine it with the explicit length preserving step.

Starting from \( C_A \), the problem to solve consists in length enforcing, i.e. finding a final curve \( C_F \):
1. whose length is \( L_R \);
2. that is close to \( C_A \);
3. that is smoother than \( C_A \).

The length is a strong constraint while the closeness and the smoothness are soft constraints. Hence let us choose a minimization method constrained by the length. The objective function contains a smoothness term preventing the curve to have unwanted wriggles and a distance term minimizing the distance to the attracting curve \( C_A \).

In variational design a physical model is used for the description of a "smooth" or a "fair" curve or surface [25, 36]. The most widely used fairness criteria originate from the observation that the shape of a thin elastic beam or a thin plate under deformation which minimizes the bending energy is always smooth, i.e. has a visual pleasing shape.

Since the bending energy for a parametric curve, \( E_R = \int \kappa^2(t) dt \), is a non-linear functional, it is common to use instead the linearized version [12, 4]:

\[
E_R = \int |c''(t)|^2 dt = \int x''(t)^2 + y''(t)^2 + z''(t)^2 dt .
\]

Both expressions are identical if \( |c'(t)| \equiv 1 \).

In the case of a piecewise linear curve, a discrete version is derived from a finite difference approximation of the derivatives:

\[
E(X, Y, Z) = \sum_{i=1}^{N-2} \frac{1}{2} (c_{i-1} - 2c_i + c_{i+1})^2 \]

\[
= \frac{1}{2} (X^T H X + Y^T H Y + Z^T H Z),
\]

where \( H \) is a banded matrix and \( X, Y, Z \) are the coordinate vectors of \( C \).

Besides we want the distance to \( C_A \) to be minimal. Hence we introduce a second term in the objective function, defined as the quadratic distance between \( C \) and \( C_A \):

\[
D(X, Y, Z) = ||X - X_A||^2 + ||Y - Y_A||^2 + ||Z - Z_A||^2 .
\]

The problem we aim to solve is now the following optimization problem:

\[
\arg \min \{ (1 - \beta)E(X, Y, Z) + \beta D(X, Y, Z) \} \quad \text{subject to } f_i(X, Y, Z) = 0 , \ i = 0, \ldots, N - 2 ,
\]

where the constraints \( f_i \) are defined in Eq. 2 (Section 2.1). \( 0 \leq \beta \leq 1 \) is a scalar value which balances between a smoother curve and a curve closer to \( C_A \). Similar energy functional have been used in [18].

With the technique of Lagrange multipliers [16], the minimization problem is restated to finding a stationary point of the following function:

\[
g(X, Y, Z, \Lambda) = (1 - \beta)E + \beta D + \sum_{i=0}^{N-2} \lambda_i f_i
\]

where \( \Lambda = (\lambda_0, \ldots, \lambda_{N-2})^T \) is the vector of Lagrange multipliers.

The length constraints are quadratic expressions. Since minimizing a quadratic cost function subject to quadratic constraints is costly and since one of our objectives is to provide a fast algorithm, let us approach the length constraints \( f_i \) by using linearized constraints \( \tilde{f}_i \) instead. \( \tilde{f}_i \) is the Taylor expansion of \( f_i \) with respect to \( \Delta x_i, \Delta y_i \) and \( \Delta z_i \) in the neighbourhood of \( \Delta x_i^A, \Delta y_i^A \) and \( \Delta z_i^A \). Hence one gets

\[
\tilde{f}_i = 2\Delta x_i^A \Delta x_i - (\Delta x_i^A)^2 + 2\Delta y_i^A \Delta y_i - (\Delta y_i^A)^2 + 2\Delta z_i^A \Delta z_i - (\Delta z_i^A)^2 - l_i^2 .
\]

Assuming that the approximation of \( f_i \) by \( \tilde{f}_i \) holds, the problem can be restated as solving:

\[
\overline{\nabla} g = 0 , \text{ where } \ g(X, Y, Z, \Lambda) = (1 - \beta)E + \beta D + \sum \lambda_i \tilde{f}_i .
\]

A symmetric square sparse system of linear equations has to be solved:

\[
\begin{align*}
(1 - \beta)H X + 2\beta X + 2\Delta X \Lambda &= 2\beta X_A \\
(1 - \beta)H Y + 2\beta Y + 2\Delta Y \Lambda &= 2\beta Y_A \\
(1 - \beta)H Z + 2\beta Z + 2\Delta Z \Lambda &= 2\beta Z_A \\
2\Delta X^2 X + 2\Delta Y^2 Y + 2\Delta Z^2 Z &= b
\end{align*}
\]
where \( b_i = l_i^2 + (\Delta x_i^A)^2 + (\Delta y_i^A)^2 + (\Delta z_i^A)^2 \) for \( i = 0 \ldots N - 2 \), and where \( \Delta X, \Delta Y \) and \( \Delta Z \) are 2-banded matrices of size \( N \times (N - 1) \) containing \( \Delta x_i^A, \Delta y_i^A \) and \( \Delta z_i^A \) (see [30] for details).

Let us note that the result is not exact for the length constraints due to the linearization of the length constraints.

In order to increase the precision, the solving is encapsulated in a loop which iterates the system solving and replaces \( (X_A, Y_A, Z_A) \) by \( (X, Y, Z) \) at each loop. Since the system is very sparse iterative methods are efficient.

In order to validate the approximation of \( f_i \) by the linearized expression \( \tilde{f}_i \) we have to ensure that \( \Delta x_i, \Delta y_i, \Delta z_i \) are close to \( \Delta x_i^A, \Delta y_i^A, \Delta z_i^A \). In other words the curve \( C_A \) must nearly satisfy the length constraints. Hence the choice of \( C_A \) is crucial. We explain in the next section how to find a curve satisfying these conditions.

### 2.4 Explicit length preservation step for 3D curves

In this section we present a method computing in linear time an attracting curve suitable for the optimization step since it sufficiently approximates the length constraints. Though based on the same idea as the 2D curve [30] we will see that the 3D solution is more complex, involving the intersection of two spheres and a plane instead of two circles.

There are three different curves involved in this section: \( C_R \), the reference curve, \( C_D \), the curve deformed by the user, and \( C_A \), the resulting curve that nearly satisfies the length constraint and that therefore is a good starting curve for the optimization of Section 2.3.

Suppose the user has modified the initial multiresolution curve \( C_R \) by displacing one control point at resolution level \( e, 0 \leq e < n \), leading to \( C_D \) which differs from \( C_R \) only in a localized portion. This classical multiresolution editing operation generally doesn’t preserve the length of \( C_R \). But in case of keeping the curve length constant one desired effect would be the generation of wrinkles at some scale \( w \) depending on the object. In order to approximate the reference length, the idea is to compute \( C_A \) from \( C_D \) by computing new detail coefficients of level \( w \). All other detail coefficients of level \( w + 1 \) up to \( n \) and all coarse coefficients (if possible) of level \( w \) are kept fixed.

Let us explain this particular choice. The detail coefficients at level \( w \) are used because they encode the geometric information at that scale. Assume \( c_i^w \) and \( c_{i+1}^w \) (see Fig. 2) have been moved closer by the user. A new detail \( d_i^w \) is computed such as the lengths \( l_{2i} \) and \( l_{2i+1} \) equal the reference lengths. Hence the norm of \( d_i^w \) increases, creating a wrinkle at scale \( w \). Through the choice of \( w \) the user controls explicitly the wrinkling scale.

Fixing all other detail coefficients ensures that fine details of the initial curve are preserved. To not modify the coarse coefficients of \( C_D \) at level \( w \) ensures that the global shape of the resulting curve \( C_A \) differs as less as possible from \( C_D \), i.e. it respects as much as possible the user’s original deformation.
epliants match whenever possible, except the details at level \( w \) (level of expected wrinkles). That is to say the curve \( C_A \) follows as much as possible the modification applied by the user.

- Let us recall the main condition for using the optimization method efficiently: a starting curve nearly satisfying the length constraints. Thanks to the length preservation at the scale \( w + 1 \), the reconstructed curve \( C_A \) (i.e. after adding back the fine details) has almost the same length as the initial curve. Hence it can be used as a good starting point for the second step (see Section 2.3) in order to precisely approach the length constraint.

- Following this method, a main role of the previous optimization method, whose cost dramatically drops, is to smooth the curve. It is useful since \( C_A \) generally may have sharp features.

3 Surfaces

Modelling inelastic soft tissues is a hard task because it involves area preserving, length preserving and minimizing some energies on the surface. In this section we show how the length preserving curve deformations presented in the previous section can be used to mimic the dynamic behaviour of materials like skin or cloths. We take advantage of the computational efficiency of the curve model in order to generate wrinkles dynamically with an easy control on the frequency and the extent.

The main idea of our proposal is to extract a curve on the surface that will be deformed following the algorithm of Section 2 and to re-inject it into the surface. The wrinkles are then propagated in a pre-defined neighbourhood of this curve. The wrinkles are smoothly attenuated inside the neighbourhood. The propagation is obtained by displacing each vertex of the neighbourhood, depending on its distance to the curve and on the curve’s nearest points. Though all examples are given with manifold triangular meshes it can be applied to any manifold mesh.

3.1 Curve extracting

The first step consists in extracting a curve defined as a sequence of edges of the mesh. Hence we get a piecewise linear 3D curve whose control points correspond to the vertices of the mesh. Depending on the application several extracting methods are possible, including user defined curve, shortest path between specified ends [24], intersection with other objects, extracting ridges [26], etc.

Then the curve is deformed at constant length using the MR method presented in Section 2. The deformed curve thus defines a displacement of the corresponding mesh vertices. In order to create a surfacic deformation we have to propagate these displacements on the mesh in a predefined neighbourhood.

3.2 Propagation on the surface

Firstly we specify an area of influence defining the part of the mesh that will be modified. In the literature, different approaches to solve this problem have been used. In [20] they use a geodesic distance on the surface. This geometry-based approach is probably the most accurate but it is much time consuming. Another possibility [21] is to project the mesh on the tangent plane and to use the euclidian distance, but this approach fails in high-curvature regions. This is the reason why we choose the following topology-based approach. Herein, we choose the region of extend for the wrinkle propagation by selecting the vertices at a uniformly bounded topological distance of the extracted curve (Fig. 3). The topological distance between two vertices is the minimal number of edges for all paths connecting this two vertices. We define the area of influence as the set of mesh triangles whose vertices have a topological distance to a curve’s vertex less or equal than a specified number. The choice of using topological distance is made for efficiency reasons. Highly non-uniform meshes would however require using geodesic distance instead of a topological distance.

Figure 3. Propagation of the deformation.

Let us now explain how to compute the new position of all vertices belonging to the area of influence so that the new shape of the extracted 3D curve is propagated smoothly inside that area. Let \( v \) be a vertex (see Fig. 3) inside the area. The corresponding vertex \( \tilde{v} \) on the deformed surface is obtained by \( \tilde{v} = v + \delta_v \). The displacement \( \delta_v \) has to be defined. This vertex may have several closest points on the curve (named \( v_1, \ldots, v_k \)) with respect to the topological distance. It means that there is a path from \( v \) to each of them with topological length equal to the minimal topological distance \( l_v \) from \( v \) to the curve. Remember we know the displacement vector of each curve point. The displacement \( \delta_v \) is defined as an average of \( \delta_{v_1}, \ldots, \delta_{v_k} \). But
the edges of the mesh may have different lengths involving important differences between the geometric distances $\|v - v_1\|, \ldots, \|v - v_k\|$. In order to overcome that problem the average is weighted by the inverse of the geometric distance between the points and $v$:

$$\delta_v = a(l_v) \left( \sum_{i=1}^{k} \frac{1}{\|v - v_i\|} \right)^{-1} \sum_{i=1}^{k} \frac{1}{\|v - v_i\|} \delta_{v_i}, \quad (6)$$

where $a(l)$ is a transverse attenuation function making the deformation more realistic. It must decrease from 1 for $l_v = 0$ (i.e. $v$ belongs to the curve) to 0 for $l_v$ maximum (i.e. when $v$ is on the boundary of the selected area). In our examples $a$ is made up of a step $a(l) = 1$ for $l$ small, followed by a smooth cubic polynomial join.

A hard task when deforming surfaces is to prevent self-intersections. Creating wrinkles may cause some when the user prescribes a large compression. A simple heuristics proved in practice to prevent most of the intersections between neighbouring wrinkles. It consists in slightly modifying the explicit length preserving step (see Section 2.4): the radii-equalizing technique is extended to hazardous cases, i.e. when $\alpha < 0$ or $\alpha > 1$ (see Appendix B).

Furthermore, one could consider some alternatives to the propagation around one single reference curve. One of them consists in applying the method of Section 2 on several regularly spaced curves (according to the topological distance) but it turns out to lack robustness.

### 3.3 Results

We present here examples of mesh editing that illustrates the process previously presented. Once the mesh is loaded and the deformation area is defined, the deformation process works dynamically.

Figure 4 shows the wrinkling of a mesh. The initial mesh (left) has 50,000 triangles and the selected area (blue/dark area) has 1,400 triangles. The extracted curve (red) is pinched by the user and the length is enforced following the process of Sections 2 and 3. Two different values for $w$ (see Section 2.2) are used: $w = 4$ (middle) and $w = 3$ (right), providing small and large wrinkles on the back of the hand. It illustrates the control of the wrinkling scale thanks to the MR representation. The propagation process automatically creates realistic surface wrinkles.

Figure 5 shows the superposition of three deformations (lower row). The initial surface (upper left) has 93,000 triangles but only 13,000 triangles have been modified, creating several wrinkles around the cow’s neck (upper right). This example illustrates the ability of our model to be integrated into a surface designing tool.

### 4 Conclusion

We have first presented a method for length preserving deformation of multiresolution 3D curves. It combines a step of length approximating and a smoothing step via constrained optimization. The MR representation allows easy editing and control of the deformation scale. This method is then used for mesh wrinkling by extracting a deformation profile which is propagated on the surface. The whole process works dynamically. It is illustrated on triangular meshes.

Future works concern the adaptation of the method to adaptative meshes. Refining the mesh on the wrinkling area while keeping a coarse sampling on flat areas could improve the results with minor computation overloading.

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Figure 5. Wrinkles around the neck. The initial mesh (upper row left) is wrinkled (upper row right) through three deformations (lower row).

References


A. Multiresolution curves

Let us briefly sketch the notation of the wavelet based multiresolution analysis used here. For more details see [23], [13], and [31]. Suppose we have a certain functional space $E$ and some nested linear approximation spaces $V^j \subset E$ with $V^0 \subset V^1 \subset \cdots \subset V^n$. Let $V^n$ be of dimension $N$. Since we are dealing with finite curves, these spaces have finite dimension. Let $V^j$ be spanned by a set of basis functions $\varphi^j = [\varphi^j_1, \ldots, \varphi^j_m]^T$, called scaling functions. A space $W^j$ being the complement of $V^j$ in $V^{j+1}$ is called the detail space. Its basis functions $\psi^j = [\psi^j_1, \ldots, \psi^j_{N-j}]^T$ are such that together with $\varphi^j$ they form a basis of $V^{j+1}$. The functions $\psi^j$ are called wavelets.

The space $V^n$ can therefore be decomposed as follows:

$$V^n = V^{n-1} \oplus W^{n-1} = V^{n-2} \oplus \cdots \oplus W^2 \oplus W^1 = \cdots = V^0 \oplus \cdots \oplus W^j.$$  

(7)

A multiresolution curve is then defined as a parametric curve $c(t) = (c^n(t))^T (c^n)$, element of $V^n$, where $c^n$ is a column of its $N$ control points in $\mathbb{R}^3$. Due to property 7 the same curve can be expressed in terms of the basis functions of the different decompositions of $V^n$, each of it corresponding to a certain resolution of the curve. The multiresolution curve at any level of resolution $\epsilon \in [0, n]$, i.e. element of $V_\epsilon \oplus \cdots \oplus W_j$ is then given by some coarse control points $c^\epsilon$ that form approximations of the initial control polyline and by the detail coefficients $d^\epsilon, \ldots, d^{n-1}$ as follows:
\[ c(t) = (c^e)^T(\varphi^e) + (d^e)^T(\psi^e) + \ldots + (d^{n-1})^T(\psi^{n-1}). \]  

The filter bank algorithm [23, 13] is used to compute the coefficients of all levels of resolutions from the initial coefficients \( c^o \) and vice versa.

The function so represented can be edited in an intuitive way: editing the coarse coefficients modifies the global shape without affecting the details; in contrast editing the details modifies the character of the object without changing its overall shape.

**B Intersecting 2 spheres and a half-plane**

![Figure 6. Intersecting two spheres.](image)

Figure 6. Intersecting two spheres.

The problem of section 2.4 can be stated as finding the intersections of (see Fig. 6):

1. the sphere \( S_1(O_1, r_1) \),
2. the sphere \( S_2(O_2, r_2) \),
3. the plane \( P : (O_1, O_2, \delta) \),

instantiated with \( O_1 = e_i^w \), \( O_2 = e_{i+1}^w \), \( r_1 = \tilde{t}_1 \), \( r_2 = \tilde{t}_{i+1} \) and \( \delta = d_i^w \) (detail of the reference curve \( C_R \)).

Let us define \( d = \text{dist}(O_1, O_2) = \|\overrightarrow{O_1O_2}\| \).

Assuming the intersection \( S_1 \cap \mathcal{S}_2 \neq \emptyset \) (i.e. \( r_1 + r_2 \leq d \) and \( |r_1 - r_2| \leq d \)), that intersection is a circle \( \mathcal{C}(O, r) \):

1. lying in a plane orthogonal to \( \overrightarrow{O_1O_2} \) (then also to \( P \)),
2. with center \( O \in (O_1O_2) \),
3. with radius \( r \leq \min \{r_1, r_2\} \).

Let \( (1 - \alpha, \alpha) \) be the homogenous barycentric coordinates of \( O \) with respect to \( \{O_1, O_2\} \) i.e. \( \overrightarrow{O_1O} = \alpha \overrightarrow{O_1O_2} \) (see Fig. 7).

Applying Pythagorean theorem in the triangles \( (O_1, O, F) \) and \( (O_2, O, F) \) with \( F \) belonging to the circle, we deduce \( O \) and \( r \):

\[
\begin{aligned}
\{ & r^2 + (1 - \alpha)^2 d^2 = r_2^2 \\
& r^2 + \alpha^2 d^2 = r_1^2 \\
\Rightarrow & \left\{ \begin{array}{l}
\alpha = \frac{(d^2 + r_1^2 - r_2^2)}{2d^2} \\
r^2 = \frac{r_1^2}{2} - \alpha^2 d^2
\end{array} \right.
\end{aligned}
\]

Hence \( S_1 \cap \mathcal{S}_2 \cap \mathcal{P} = \mathcal{C}(O, r) \cap \mathcal{P} \) is the set of 2 points \( \{F, F'\} \) such as:

1. \( \overrightarrow{OF} = \overrightarrow{OF'} \perp \overrightarrow{O_1O_2} \),
2. \( F \) and \( F' \in \mathcal{P} \), i.e. \( \overrightarrow{OF} \) and \( \overrightarrow{OF'} \perp \overrightarrow{\delta} \times \overrightarrow{O_1O_2} \),
3. \( \|\overrightarrow{OF}\| = \|\overrightarrow{OF'}\| = r \),
4. \( \overrightarrow{OF} = -\overrightarrow{OF'} \).

Then \( \overrightarrow{OF} \) and \( \overrightarrow{OF'} \) are the only 2 vectors \textit{colinear} with \( \overrightarrow{O_1O_2} \times (\overrightarrow{\delta} \times \overrightarrow{O_1O_2}) \) whose norm equals \( r \). The one (let’s say \( \overrightarrow{OF} \)) lying in the same half-plane as \( \overrightarrow{\delta} \) has the same direction as \( \overrightarrow{O_1O_2} \times (\overrightarrow{\delta} \times \overrightarrow{O_1O_2}) \). The point \( F \) is our choice for the head of the new detail vector \( d_i^w \).

![Figure 7. The intersection circle.](image)

(a) \( \alpha > 0 \)

(b) \( \alpha < 0 \)