

A multiresolution wavelet scheme for irregularly subdivided 3D triangular mesh

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Abstract

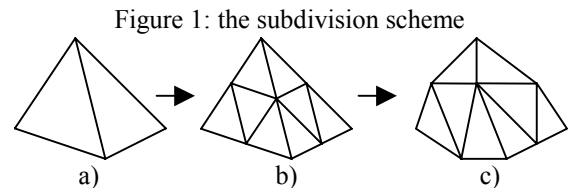
We propose a new subdivision scheme derived from the Lounsbery's regular 1:4 face split, allowing multiresolution analysis of irregularly subdivided triangular meshes by the wavelet transforms. Some experimental results on real medical meshes prove the efficiency of this approach in multiresolution schemes. In addition we show the effectiveness of the proposed algorithm for lossless compression.

1. Introduction

Multiresolution analysis of 3D objects is receiving a lot of attention nowadays, due to the practical interest of 3D modelling in a wider and wider range of applications. Multiresolution analysis of these objects gives some useful features : several levels of details can be built for these objects, accelerating the rendering when there is no need for sharp details, and allowing progressive transmission. Another feature is that multiresolution analysis can be an efficient way for data compression. A survey of the existing methods used to simplify meshes which is the first step for processing multiresolution analysis, like vertex decimation [2], edge contraction [3] and wavelet surfaces [4], was reported in [1]. We put our attention on the third method, because wavelets are well-suited for multiresolution analysis. In section 2, we will shortly explain multiresolution analysis of meshes [3], and show its drawbacks in practical implementation, which we improved, as described in section 3. In section 4, we show why our proposal is suitable for compression. The next part (section 5) gives the results obtained with this new scheme.

2. Lounsbery's wavelets based multiresolution scheme

In wavelets decomposition, a mesh (for example a tetrahedron, see figure 1.a) is quaternary subdivided (figure 1.b) and deformed (figure 1.c), to make it fit the surface to approximate. Subdividing the mesh consists in splitting each triangular face into four faces. These steps can be processed depending on the required resolution levels.



Multiresolution analysis is computed with two analysis filters A^j and B^j for each resolution level j . Reconstruction is done with two synthesis filters P^j and Q^j . These filters are represented with matrix notation and, to ensure exact reconstruction, must satisfy the following constraint:

$$\begin{bmatrix} A^j \\ B^j \end{bmatrix} = \begin{bmatrix} P^j & Q^j \end{bmatrix}^{-1} \quad (1)$$

Let us call C^j the $N(j) \times 3$ matrix giving the coordinates of each vertex of the mesh at the resolution level j . Then we have the relations:

$$C^j = A^{j+1} \cdot C^{j+1} \quad (2)$$

$$D^j = B^{j+1} \cdot C^{j+1} \quad (3)$$

$$C^{j+1} = P^j \cdot C^j + Q^j \cdot D^j \quad (4)$$

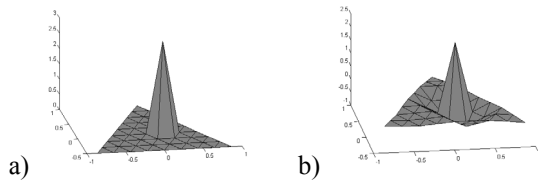
D^j represents the wavelet coefficients of the mesh, necessary to reconstruct C^{j+1} from C^j . From a theoretical point of view, each column of the P^j matrix (respectively the Q^j matrix) represents a scaling function (respectively a wavelet function). These functions are defined on a 3D space fixed by the mesh topology.

We apply the lifting scheme [6] which consists here in constructing wavelet functions (starting from the hat function, figure 2.a) orthogonal to the scaling functions (which are hat functions too, but with a twice wider support). Without the lifting scheme, Lounsbery's multiresolution analysis would simply consist in subsampling the mesh, but with the lifting, the mesh at resolution level j is ensured to be the best approximation in the mean square sense for the mesh at level $j+1$. The main material for the lifting is the inner product between two functions defined by Lounsbery as:

$$\langle f, g \rangle = \sum_{\tau \in \Delta(M)} \left(\frac{K_j}{\text{area}(\tau)} \int_{s \in \tau} f(s).g(s).ds \right) \quad (5)$$

$\Delta(M)$ is the set of triangles τ of the mesh and K_j is a constant for a given resolution level j . ($K_j=4^{-j}$). Note that in this inner product it is assumed that the triangular faces of the mesh have the same area. The consequence of this assumption is that a mesh at resolution level j will effectively be the best approximation of the mesh at level $j+1$ only if this constraint is fulfilled. We can see in figure 2.b) the effects of the lifting scheme on the hat function showed in figure 2.a).

Figure 2: wavelets in 3D

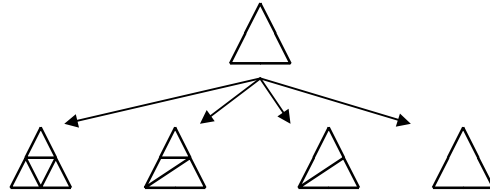


Wavelet surfaces give a powerful tool for multiresolution analysis. However, in the simplification process, the major drawback is that faces are always merged four to one to have a simpler mesh. If the mesh does not respect this connectivity constraint, one has to process a resampling of the mesh, which results in a mesh having more faces than the original, as explained in details in [5]. The aim of this work is to solve this problem by improving the subdivision process, as described in the next section.

3. A proposal for irregular subdivision

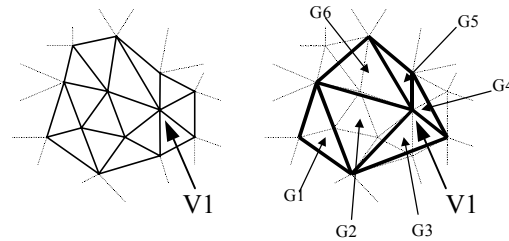
In the proposed scheme, the subdivision process is changed : each face of the mesh to subdivide is no more systematically split into four faces, but can also be split into three or two faces or remain unchanged. As an example, four different cases of subdivision are shown in figure 3. This approach allows to simplify meshes even if some faces cannot be merged four to one.

Figure 3: some possible cases of subdivision



The simplification is done with an algorithm that merges the faces of a mesh, considering rules established by the subdivision process. Figure 4 shows an example, where 15 faces are reduced to 6, resulting from merging 4 :1 faces for G2, 3 :1 faces for G3 and G6, 2 :1 faces for G1 and G4 and keeping one face unchanged for G5.

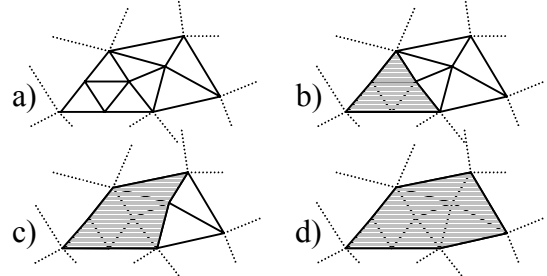
Figure 4: an example of surface simplification



One important consequence of the simplification rules is that a vertex can be removed during the simplification process only if its valence is equal to 4, 5 or 6. Clearly the efficiency of our algorithm depends on the number of removable vertices in the mesh. Fortunately, meshes usually have such vertices, and during the simplification, the valence of the vertices tends to decrease. As an example, the vertex V1 shown in figure 4 has a valence of 7, which makes it impossible to remove, but after one simplification step, its valence is reduced to 5, and the algorithm will be able to remove it in a further step.

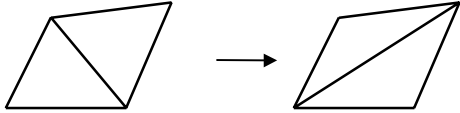
Briefly, the simplification algorithm starts by merging 4 faces to 1, building a set of merged faces, and tries to expand this set by merging faces around it. Figure 5 shows the beginning of the expansion of the merged faces set (in gray), merging sequentially 4:1 faces, 3:1 faces and 2:1 faces.

Figure 5: expansion of the simplified mesh set



The algorithm stops when no more faces have to be simplified. In order to prevent the algorithm from being unable to simplify some faces with respect to the subdivision rules, a modification of the mesh is allowed. It consists in an edge permutation between two neighbour faces, as shown in figure 6. Of course this modification has to be stored, to recover the original mesh after subdivision and guarantee the reversibility of the simplification process.

Figure 6: an edge permutation between two faces



We notice that this modification will introduce a quality loss during multiresolution analysis, but the difference between the original mesh and the altered mesh is small and experimental results show that this local error is negligible compared to the approximation error. Finally, the algorithm is very efficient for simplifying a large set of meshes.

The last thing to do is to compute the approximation of the high resolution mesh with the simplified one that is to calculate the analysis filters A^j and B^j . This can be done with Lounsbery's scheme. A difference has to be noticed, due to the change of the subdivision process. The inner product (5) has to be reformulated and becomes:

$$\langle f, g \rangle = \sum_{\tau \in \Delta(M)} \left(\frac{K_j(\tau)}{area(\tau)} \int_{s \in \tau} f(s).g(s).ds \right) \quad (6)$$

$K_j(\tau)$ is no longer a constant and changes with each face of the mesh. For example, a face in a low resolution mesh that will split in 3 faces will have $K_j(\tau)=3$ and the three resulting faces will have $K_{j+1}(\tau)=1$, taking into account the differences between the triangle areas : the first face cited above will approximately be three times larger than the three last.

4. Compression

The proposed method has powerful features for compressing meshes, for two reasons:

- The wavelet decomposition, used to compute the vertices coordinates, transforms coordinates into wavelet coefficients which histogram is concentrated around the zero value, making them well suited for entropy coding.

- Starting from the lowest resolution level, there is no need to store or transmit the faces descriptions to reconstruct higher levels, only the subdivisions have to be, which lets the amount of information needed to reconstruct the connectivity of the mesh close to 3 bits per face.

In the experimental results section, the lowest resolution mesh is coded using the algorithm described in [8].

5. Results







Table 1 shows the results on a heart mesh simplification using the proposed method. The high resolution mesh is a regular tessellation similar to that proposed in [7]. Its vertices coordinates are coded with a 7 bits/vertex precision. The number of bytes noted in table 1 is the amount of information needed to reconstruct the mesh to the concerned resolution level starting from the nearest lower resolution level.

Table 2 gives some more results on different 3D meshes: brain, body and lung. We notice that at the middle resolution levels, used in practical implementation, the approximation quality remains rather good in terms of local shape and size.

Table 1: multiresolution representation of a 3D left ventricle of a heart mesh

<p>Level 6</p> <p>1008 faces 413 bytes</p>	<p>Level 5</p> <p>446 faces 226 bytes</p>	<p>Level 4</p> <p>208 faces 123 bytes</p>
<p>Level 3</p> <p>94 faces 65 bytes</p>	<p>Level 2</p> <p>46 faces 36 bytes</p>	<p>Level 1</p> <p>22 faces 22 bytes</p>
<p>Level 0</p> <p>8 faces 40 bytes</p>	<p>Summary: Original file: 4730 bytes Compressed files: 925 bytes</p> <p>Ratio : 5.1 : 1</p>	

Table 2: complementary results

<p>Original mesh</p>  <p>3584 faces 8 levels</p>	<p>Level 4</p>  <p>524 faces</p>	<p>Original file: 19493 bytes</p> <p>Compressed file: 3246 bytes</p> <p>Ratio: 6.1 : 1</p>
<p>Original mesh</p>  <p>4454 faces 10 levels</p>	<p>Level 5</p>  <p>306 faces</p>	<p>Original file : 25894 bytes</p> <p>Compressed file: 4062 bytes</p> <p>Ratio : 6.4 : 1</p>
<p>Original mesh</p>  <p>1916 faces 10 levels</p>	<p>Level 5</p>  <p>168 faces</p>	<p>Original file: 9705 bytes</p> <p>Compressed file : 1801 bytes</p> <p>Ratio : 5.4 : 1</p>

6. Conclusion

We proposed a new scheme allowing to process multiresolution analysis on arbitrary meshes. In sharp contrast with [4] where a resampling of the original mesh is necessary, our scheme processes directly on the original mesh. The proposed method has many potential applications such as mesh compression, progressive transmission and fast rendering of 3D images.

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