

## GENERATION OF UNSTRUCTURED TETRAHEDRAL MESHES USING AN ADVANCING FRONT METHOD

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**Abstract.** *This paper presents an algorithm to generate uniform unstructured tetrahedral meshes based on advancing front method. The goal is to develop an efficient algorithm regardless the use of any external technique in order to improve convergence, such as backtracking or front controlling. Some examples show the ability of the algorithm to discretize arbitrary volumes, reaching reasonable distributions in terms of edge lengths, element volumes and other quality metrics.*

**Keywords:** *Mesh Generation, Unstructured Meshes, Advancing Front*

### 1. INTRODUCTION

Nowadays, various methods for generating unstructured tetrahedral meshes are available, but the most used are based on Delaunay triangulation (Delaunay, 1924), advancing front method (Lo, 1985), and recursive spatial decomposition (Meagher, 1982; Samet, 1989), in descending order of popularity. Advancing front based methods have the advantage that they produce boundary conforming meshes, and only one element is created per each node insertion, allowing total control when positioning the created nodes. However, the random way in which the front changes causes the generation of distorted polyhedra that usually break the discretization. In addition, it is necessary to use special data structures to make possible a robust code.

In this paper, an implementation of the advancing front method is proposed. This version is based on the algorithms conceived by Löhner and Parikh (1988) and Peraire and Peiró (1992). More algorithms such as Jin and Tanner (1993), Frykestig (1994), Dari (1994), Möller and Hansbo (1995), Rassineux (1998), Piteri (1998), Frey et al. (1998), and Cavalcante Neto et al. (2001) had also influenced it. Like in Golgolab (1989), Cavalcante Neto et al. (2001) and George and Seveno (1994), the domain surface mesh is assumed as an input data. This approach is interesting because it detaches the mesh generation process from the geometric modelling. Procedures to increase convergence, such as backtracking (Möller and Hansbo, 1995; Rassineux, 1998; Cavalcante Neto et al., 2001) or generation by layers (Piteri, 1998; Marcum, 2001; Ito et al., 2004), are not used. Only uniform meshes are generated, with element or edge sizes approximately constant throughout the domain.

## 2. PROPOSED METHOD

In the algorithm here proposed, the basic structure of a classic advancing front method is maintained. However, it was decomposed in two stages: generation of well shaped elements (Stage 1) and generation of elements to enforce convergence (Stage 2). During Stage 1, the maximum permitted number of elements with high quality are created, in order to avoid the propagation of distorted tetrahedra. Remaining void spaces after Stage 1 are discretized at Stage 2 with tetrahedra of inferior quality. Therefore, the larger is the number of the generated elements at Stage 2, the poorest is the quality of the mesh. In fact, it was developed one advancing front based algorithm for each particular stage.

Let  $\mathcal{M}_\Omega$  be the tetrahedral mesh of a domain  $\Omega$ ,  $\Omega \subset \mathbb{R}^3$ , with boundary denoted by  $\partial\Omega$ ,  $\partial\Omega \subset \mathbb{R}^3$ , which has its discretization represented by  $\mathcal{M}_{\partial\Omega}$ . Then,

**Definition 1.** A *front*  $\mathcal{F}_\Omega$  of a domain  $\Omega$ ,  $\Omega \subset \mathbb{R}^3$ , is a surface mesh whose faces  $F_i$ ,  $F_i \in \mathcal{F}_\Omega$ ,  $i = 1, 2, \dots, n_F$ , are used to create the tetrahedra that will compose  $\mathcal{M}_\Omega$ .

The main scheme of the proposed method is depicted in Alg. 1. The input parameters are an empty domain mesh  $\mathcal{M}_\Omega$  and its corresponding boundary discretization  $\mathcal{M}_{\partial\Omega}$ . Initially, the front  $\mathcal{F}_\Omega$  is equal to  $\mathcal{M}_{\partial\Omega}$ . After a number of elements is generated,  $\mathcal{F}_\Omega$  is updated and no longer represents  $\mathcal{M}_{\partial\Omega}$ . Here, the subscript  $\Omega$  in  $\mathcal{F}_\Omega$  specifies its original domain. Now, the faces can be classified in two types: *active* and *passive*. The following definition describes this notion.

**Definition 2.** A face  $F$ ,  $F \in \mathcal{M}_\Omega$ , is called *active*, if  $F \in \mathcal{F}_\Omega$ . Otherwise,  $F$  is *passive*.

A tetrahedron  $E$  is created by selecting an active face  $F_i$ ,  $F_i \in \mathcal{F}_\Omega$ ,  $i = 1, 2, \dots, n_F$ , and testing whether  $F_i$  generates a valid element. This particular selected active face is called *base face*. For this face, an “ideal” node is determined. This node is defined as follows.

**Definition 3.** A candidate node is said to be an *ideal node*, denoted by  $P_{ideal}$ , if its position is calculated in relation to a base face  $F_i = P_1P_2P_3$ ,  $F_i \in \mathcal{F}_\Omega$ ,  $i = 1, 2, \dots, n_F$ , such that it maximizes the adopted shape quality metric  $\mathcal{Q}_E$  of the element  $E = P_1P_3P_2P_{ideal}$ .

The ideal node  $P_{ideal}$  is calculated by iterations as in Frey et al. (1998) so that the resulting tetrahedron is almost equilateral. To check the quality of an element, it was used a tetrahedron shape quality metric ( $\mathcal{Q}_E$ ) defined by

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### Algorithm 1 Main scheme

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- 1:  $\mathcal{F}_\Omega \leftarrow \mathcal{M}_{\partial\Omega}$ .
  - 2: Define element size  $h_E$ .
  - 3: Stage 1 – Generation of well shaped elements.
  - 4: Stage 2 – Generation of elements to enforce convergence.
  - 5: If  $\mathcal{F}_\Omega = \emptyset$  then
  - 6:      $\mathcal{M}_\Omega$  was successfully generated.
  - 7: Else
  - 8:     Convergence deficiency.
  - 9: End If
-

**Definition 4** (Dompierre et al. (1998)). A *shape quality metric* of a tetrahedron  $E$ , or simply a *shape metric* of  $E$ , is a continuous scalar function  $\mathcal{Q}_E : \mathbb{R}^3 \rightarrow \mathbb{R}$ , satisfying the following conditions:

- S1.  $\mathcal{Q}_E$  is invariant under translation, rotation, reflection and uniform scaling of  $E$ ;
- S2.  $0 \leq \mathcal{Q}_E \leq 1, \forall E$ ;
- S3.  $\mathcal{Q}_E = 1 \Leftrightarrow E$  is a regular tetrahedron;
- S4.  $\mathcal{Q}_E = 0 \Leftrightarrow E$  is degenerated.

In this algorithm,  $\mathcal{Q}_E$  is equal to the minimum solid angle of a tetrahedron. Considering an element  $E = P_1P_2P_3P_4$ , the solid angle ( $\theta_i$ ) in each vertex  $P_i, i = 1, 2, 3, 4$ , is equal to the surface area of a spherical triangle delimited by the intersection between  $E$  and an unit sphere centering in  $P_i$ . For details about the minimum solid angle and other shape metrics see Parthasarathy et al. (1993), Liu and Joe (1994) and Naylor (1999).

Whenever an element  $E$  is created,  $\mathcal{F}_\Omega$  advances into  $\Omega$ . The domain  $\Omega$  will be totally discretized and the mesh  $\mathcal{M}_\Omega$  completely defined, when  $\mathcal{F}_\Omega = \emptyset$ . At this moment, Alg. 1 returns  $\mathcal{M}_\Omega$  and the discretization process finalizes. If any triangular face is left in  $\mathcal{F}_\Omega$  after Stage 2, the discretization process will fail. This is an indicative of deficient convergence of advancing front based algorithms. Convergence problems are solved by restarting the entire process with a new  $\mathcal{M}_{\partial\Omega}$  or applying a perturbation  $\delta$  over the element size ( $h_E$ ), i.e.,  $h_E = (1 + \delta) h_E$ . Caution is necessary in the application of  $\delta$ , because it affects the distribution of  $h_E$  throughout the domain. In this work,  $|\delta| \leq 0.15$  was used.

## 2.1 Stage 1 – Well Shaped Elements

The algorithm of this stage is similar to that of Löhner and Parikh (1988), and Peraire and Peiró (1992), but with the addition of validity tests analogous to those of Jin and Tanner (1993). However, the algorithm here proposed has four levels of quality, similarly to the multifrontal scheme presented by Rassineux (1998).

The steps of Stage 1 are shown in Alg. 2.  $\mathcal{F}_{Rej}$  is a list created to store the faces from  $\mathcal{F}_\Omega$  that did not generate tetrahedra satisfying the imposed quality restrictions, i.e., it stores rejected faces. The advancing front algorithm of Stage 1 is executed for each of the four quality levels. After each execution,  $\mathcal{F}_\Omega$  must be empty. Then, rejected faces are set as actives, and the process restarts only if some element was created. When it is not possible to generate elements satisfying the imposed restrictions, a new quality level is created by relaxing the restrictions, and the discretization process is restarted.

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### Algorithm 2 Stage 1 – Generation of well shaped elements

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- 1: Define minimum quality values.
  - 2: For each quality level, do:
  - 3:     Repeat
  - 4:         Execute the advancing front algorithm of Stage 1;
  - 5:          $\mathcal{F}_\Omega \leftarrow \mathcal{F}_{Rej}$ ;
  - 6:     While the elements have been generated.
  - 7: End For
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The advancing front algorithm applied at Stage 1 is shown in Alg. 3. While there are faces in  $\mathcal{F}_\Omega$ , the steps from 2 to 8 will be executed. For each base face  $F = P_i P_j P_k$ ,  $F \in \mathcal{F}_\Omega$ , an ideal node  $P_{ideal}$  is determined. Initially,  $P_{ideal}$  is the center of a spherical region  $S$ , with radius  $r_S = \max(\ell_{P_i P_{ideal}}, \ell_{P_j P_{ideal}}, \ell_{P_k P_{ideal}})$ , where  $\ell_{P_i P_j}$  is the length of the edge  $L = P_i P_j$ , used to search preexisting nodes near  $F$ . Like in Löhner and Parikh (1988), an *octree* data structure was used to speed up this operation. Every node  $P$ ,  $P \in S$ , is said to be in a list  $L_P$  of preexisting candidate nodes to form a valid tetrahedron with base  $F$ . The candidate node  $P$ ,  $P \in L_P$ , that results in a tetrahedron  $E = P_i P_k P_j P$  with greatest shape quality metric  $Q_E$ , satisfying the validity and quality tests, is selected. Thus, a new element is created, and  $\mathcal{F}_\Omega$  and  $\mathcal{M}_\Omega$  must be updated. If none of the nodes  $P$ ,  $P \in L_P$ , results in a valid element, then the ideal node  $P_{ideal}$  is verified. In case the tetrahedron  $E = P_i P_k P_j P_{ideal}$  satisfies the validity and quality tests, a new element and a new node ( $P_{ideal}$ ) are created. Then,  $\mathcal{F}_\Omega$  and  $\mathcal{M}_\Omega$  are updated. If it is not possible to generate an element with this base face,  $F$  is moved from  $\mathcal{F}_\Omega$  to  $\mathcal{F}_{Rej}$ .

## 2.2 Stage 2 – Enforcing Convergence

It is said that the convergence is enforced at this stage because the quality restrictions are relaxed. Therefore, it is interesting to control the quantity of generated elements at this stage so that it does not decrease significantly the global mesh quality. Ideally, the entire domain is discretized at Stage 1 resulting in a mesh with maximum quality, considering the maximum value obtained by the current algorithm. However, the random evolution of  $\mathcal{F}_\Omega$  originates well distorted polyhedra of hard decomposition, even allowing the insertion of Steiner nodes (see Schönhardt (1928); Senechal (1981); Chazelle (1984); Ruppert and Seidel (1992)). Thus, it is clear that Stage 2 is fundamental to ensure the convergence of this method.

The algorithm for Stage 2 are similar to Alg. 2. However, the initial quality conditions are less restrictive than in Stage 1. As in Stage 1, while there are faces in  $\mathcal{F}_\Omega$  the advancing front algorithm (see Alg. 4) is traversed. A face  $F = P_i P_j P_k$ ,  $F \in \mathcal{F}_\Omega$ , is selected and the position of  $P_{ideal}$  is calculated. The list  $L_P$  of preexisting nodes inside a spherical region  $S$ , with center  $P_{ideal}$  and radius  $r_S$ , is constructed. If a node  $P$ ,  $P \in L_P$ , forms an element

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### Algorithm 3 Advancing front algorithm of Stage 1

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- 1: While  $\mathcal{F}_\Omega \neq \emptyset$ , do:
  - 2:   Select a face  $F = P_i P_j P_k$ ,  $F \in \mathcal{F}_\Omega$ .
  - 3:   Calculate  $P_{ideal}$  with relation to  $F$ .
  - 4:   Mount a list  $L_P$  of preexisting candidate nodes.
  - 5:   Select the node  $P$ ,  $P \in L_P$ , that forms an element  $E = P_i P_k P_j P$  satisfying validity and quality tests, and with the greatest value of  $Q_E$  among all  $P \in L_P$ . If such a node exists, continue from step 8.
  - 6:   If  $P_{ideal}$  forms a tetrahedron  $E = P_i P_k P_j P_{ideal}$  satisfying validity and quality tests for ideal node, continue from step 8.
  - 7:   Move  $F$  from  $\mathcal{F}_\Omega$  to  $\mathcal{F}_{Rej}$  and continue from step 9.
  - 8:   Update  $\mathcal{F}_\Omega$  and  $\mathcal{M}_\Omega$ ;
  - 9: End While
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**Algorithm 4** Advancing front algorithm of Stage 2

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- 1: While  $\mathcal{F}_\Omega \neq \emptyset$ , do:
  - 2:   Select a face  $F = P_i P_j P_k$ ,  $F \in \mathcal{F}_\Omega$ .
  - 3:   Calculate  $P_{ideal}$  with relation to  $F$ .
  - 4:   Mount a list  $\mathbf{L}_P$  of preexisting candidate nodes.
  - 5:   Select the node  $P$ ,  $P \in \mathbf{L}_P$ , that forms an element  $E = P_i P_k P_j P$  satisfying validity and quality tests, and with the greatest value  $Q_E$  among all  $P \in \mathbf{L}_P$ . If such a node exists, continue from step 8.
  - 6:   If  $P_{ideal}$  forms a tetrahedron  $E = P_i P_k P_j P_{ideal}$  satisfying validity and quality tests for the ideal node, continue from step 8. Otherwise, decrease the height of  $P_{ideal}$  with relation to  $F$  and restart step 6.
  - 7:   Remove  $F$  from  $\mathcal{F}_\Omega$ , store it in  $\mathcal{F}_{Rej}$  and continue from step 9.
  - 8:   Update  $\mathcal{F}_\Omega$  and  $\mathcal{M}_\Omega$ ;
  - 9: End While
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$E = P_i P_k P_j P$  satisfying validity and quality tests, resulting in the largest value of  $Q_E$  among all candidate nodes in  $\mathbf{L}_P$ , this node is selected. Consequently, a new element is created, and  $\mathcal{F}_\Omega$  and  $\mathcal{M}_\Omega$  must be updated. Otherwise,  $P_{ideal}$  is used as a candidate node to generate a tetrahedron  $E = P_i P_k P_j P_{ideal}$ . If  $E$  satisfies validity and quality tests, an element and a node ( $P_{ideal}$ ) are created. After this,  $\mathcal{F}_\Omega$  and  $\mathcal{M}_\Omega$  are updated.

If  $E = P_i P_k P_j P_{ideal}$  do not pass the tests, the height of  $P_{ideal}$  in relation to face  $F$  is decreased by a factor of 20% and step 6 of Alg. 4 is restarted. If the minimum allowed height of  $P_{ideal}$  is reached (e.g. 20% of the original height) and the element do not pass the tests, face  $F$  is removed from  $\mathcal{F}_\Omega$  and stored in  $\mathcal{F}_{Rej}$ . Some authors (Ruud and Wille, 1990; Peraire and Peiró, 1992; Frykestig, 1994) treat these nodes as members of a separated list called *escape nodes*.

### 2.3 Validity and Quality Tests

An element can be accepted as valid only if it satisfies validity and quality tests. These tests are just verifications of validity and quality criteria to ensure mesh consistence and to avoid the propagation of distorted elements into the domain. The validity criterion applied is the following.

**Definition 5** (Validity Criterion). Let be  $\mathcal{F}_\Omega$  a front of a domain  $\Omega$ ,  $\Omega \subset \mathbb{R}^3$ , with tetrahedral mesh  $\mathcal{M}_\Omega$ . An element  $E = P_i P_j P_k P_l$  is said to be *valid* if it satisfies the following conditions:

- V1.  $V_E > 0$ , and;
- V2. If  $E \cap F \neq \emptyset$ ,  $\forall F \in \mathcal{F}_\Omega$ , then  $E \cap F \prec E$  and  $E \cap F \prec F$ .

where  $V_E$  denotes the volume of  $E$ .

Condition V1 of Def. 5 prevents the creation of degenerated tetrahedra, and condition V2 requires that the intersection between a tetrahedron  $E$  and every face  $F$ ,  $F \in \mathcal{F}_\Omega$ ,

must be empty, a vertex or an edge of either, or only equal to  $F$ . In this work, the intersection between two triangular faces is checked based on orientation predicates similar to the Guigue and Devillers (2003).

The quality criterion tries to guarantee local (i.e. element) and global (i.e. with relation to active faces) quality. It is used to check if an element is suitable to be inserted in  $\mathcal{M}_\Omega$ . Its definition is as follows.

**Definition 6** (Quality Criterion). Let be  $\mathcal{F}_\Omega$  a front of a domain  $\Omega$ ,  $\Omega \subset \mathbb{R}^3$ , with tetrahedral mesh  $\mathcal{M}_\Omega$ . An element  $E = P_i P_j P_k P_l$  is said to be *suitable* if it satisfies the following conditions:

- Q1.  $\mathcal{Q}_E \geq \min_{\mathcal{Q}_E}$ .
- Q2. The minimum distance between new faces of  $E$  and neighbor nodes in  $\mathcal{F}_\Omega$ ,  $d_{fp}$ , is greater than or equal to  $\min_{d_{fp}}$ .
- Q3. The minimum distance between new edges of  $E$  and neighbor nodes in  $\mathcal{F}_\Omega$ ,  $d_{lp}$ , is greater than or equal to  $\min_{d_{lp}}$ .
- Q4. The minimum distance between new edges of  $E$  and neighbor edges in  $\mathcal{F}_\Omega$ ,  $d_{ll}$ , is greater than or equal to  $\min_{d_{ll}}$ .
- Q5. If  $P_l$  is a new node, the minimum distance between it and neighbor faces in  $\mathcal{F}_\Omega$ ,  $d_{pf}$ , is greater than or equal to  $\min_{d_{pf}}$ .
- Q6. The angle between new faces of  $E$  and incident edges,  $\theta_{fl}$ , is greater than or equal to  $\min_{\theta_{fl}}$ .
- Q7. The angle between new faces of  $E$  and adjacent faces,  $\theta_{ff}$ , is greater than or equal to  $\min_{\theta_{ff}}$ .

Condition Q1 from Def. 6 ensures the minimum element quality levels. On the current implementation, Stage 1 initializes with  $\min_{\mathcal{Q}_E} = 0.60$ , which is maintained until the second loop over  $\mathcal{F}_\Omega$ . During the last two loops at Stage 1,  $\min_{\mathcal{Q}_E}$  is equal to 0.40. From the first to the third loop of Stage 2,  $\min_{\mathcal{Q}_E}$  is equal to 0.40, 0.30 and 0.20, respectively. And if there are faces in  $\mathcal{F}_\Omega$ , the fourth loop is traversed with  $\min_{\mathcal{Q}_E} \approx 0.0$  to enforce convergence. Conditions Q2 – Q7 from Def. 6 are posed to maintain a minimum distance between active faces and new nodes, edges or faces. This is necessary to preserve an empty region between opposed front faces that can be discretized.

The minimum values of distances and angles were obtained based on a unit regular tetrahedron  $E$ . For such an element, it is observed the following properties:

- The minimum distance between any vertex  $P$ ,  $P \in E$ , and its corresponding opposed face is equal to  $0.81h_E$ .
- The minimum distance between any vertex  $P$ ,  $P \in E$ , and a non-incident edge  $L$ ,  $L \in E$ ,  $P \notin L$ , is equal to  $0.86h_E$ .

- The minimum distance between any edge  $L$ ,  $L \in E$ , and its corresponding opposed edge is equal to  $0.70h_E$ .
- The angle between any face  $F$ ,  $F \in E$ , and an incident edge to  $F$  is equal to  $54.7^\circ$ .
- The angle between any face  $F$ ,  $F \in E$ , and an adjacent face to  $F$  is equal to  $70.5^\circ$ .

From these properties, initial minimum values were established:  $\min_{d_{fp}} = 0.67h_E$ ,  $\min_{d_{ip}} = 0.67h_E$ ,  $\min_{d_{il}} = 0.60h_E$ ,  $\min_{\theta_{fi}} = 45^\circ$ , and  $\min_{\theta_{ff}} = 60^\circ$ . After each loop over  $\mathcal{F}_\Omega$ , they are decreased about a factor of 15%. At the end, they become close to zero. In this case, an element is judged valid only checking the validity criterion of Def. 5.

New nodes that results in valid elements must satisfy an additional quality criterion as described in condition Q5 from Def. 6. This condition avoids the insertion of nodes that are close to the active faces. During Stage 1,  $\min_{d_{pf}}$  is made equal to  $0.67h_E$  to guarantee that new nodes are located in an optimum distance from  $\mathcal{F}_\Omega$ . At the begin of Stage 2,  $\min_{d_{pf}}$  is maintained equal to  $0.67h_E$ . However,  $\min_{d_{pf}}$  is decreased about a factor of 50% after each loop over  $\mathcal{F}_\Omega$  until  $\min_{d_{pf}}$  is approximately equal to zero in the last loop.

## 2.4 Front Management

Active faces are ordered using a heap  $\mathcal{H}_\Omega$  (Löhner, 1988). The area of each face  $F \in \mathcal{F}_\Omega$  was used as the key ordering. Thus, a face with smallest area among all in  $\mathcal{F}_\Omega$  will be always at the root of  $\mathcal{H}_\Omega$ . When an active face is selected from  $\mathcal{F}_\Omega$ , the number of the face in the root of  $\mathcal{H}_\Omega$  is returned. Although  $h_E$  is constant throughout  $\Omega$ , this strategy is used to prevent large elements going into zones of small elements. The rejected faces list  $\mathcal{F}_{Rej}$  is ordered in a similar way, using another heap  $\mathcal{H}_{Rej}$ .

New front faces can arise after each created element. In most of algorithms, these faces are stored in  $\mathcal{F}_\Omega$  and are inserted in  $\mathcal{H}_\Omega$ . Nevertheless, in the actual implementation, they are stored in  $\mathcal{F}_{Rej}$  and are inserted in its corresponding heap  $\mathcal{H}_{Rej}$ , in which the respective areas are the value of key. Also, when a base face  $F$  is rejected from  $\mathcal{F}_\Omega$ , the face  $F$  is moved to  $\mathcal{F}_{Rej}$  and inserted in  $\mathcal{F}_{Rej}$ , but with a null value of key instead of its area. This scheme keeps oldest faces located before newest faces in  $\mathcal{F}_{Rej}$ . Then, when the faces in  $\mathcal{F}_{Rej}$  become actives,  $\mathcal{F}_\Omega$  will advance from outer to the inner part of the subdomain delimited by the current front. This is used to decrease the effects of the random way the front develops without significant changes of the advancing front algorithm.

## 3. RESULTS AND DISCUSSIONS

In this section, a classical benchmark, the unit cube, and two other examples of arbitrary geometries are presented. The quality parameters of the generated meshes and the corresponding computational cost are shown, allowing a full analysis of the proposed algorithm.

All results were obtained using C programming language and Microsoft® Visual C++® 6.0 on a AMD Athlon™ XP1800+ processor (512MB of RAM, 256KB of cache, and Windows® XP Professional SP2 as operational system).

### 3.1 Unit Cube

The input data for this example is a surface mesh with 1341 vertices and 2678 faces (see Fig. 1(a)). The final tetrahedral mesh, whose cross section is shown in Fig. 1(b), consists of 3723 vertices, 23222 edges, 37661 faces and 18161 elements. Approximately 72% of the elements was generated in Stage 1 and 28% in Stage 2. The total CPU time was 35.75 seconds (31.47s in Stage 1 and 4.28s in Stage 2).

The relation between maximum and minimum edge lengths ( $\ell$ ) is around 6 (six). The minimum volume ( $V$ ) is equal to  $4.25 \times 10^{-8}$  (see Tab. 1) indicating the presence of flat elements mostly generated at Stage 2. In spite of this, it is noted in Fig. 1(b) an uniform distribution of element sizes  $h_E$ . Figure 2(a) shows a quasi-Gaussian distribution of the minimum solid angle ( $\sigma_{min}$ ), characteristic of  $\sigma_{min}$  distributions as observed by Dompierre et al. (1998). About 66%, 96% and 99% of the elements has  $\sigma_{min}$ , radius ratios ( $\rho$ ), and mean ratios ( $\eta$ ), respectively, greater than 0.50. The distributions of Fig. 2(b) show the ability of the algorithm to reach the required element size. In addition, the mean value of  $h_E$  was equal to 0.0833 (14% greater than  $h_E$  in  $\mathcal{M}_{\partial\Omega}$ ).

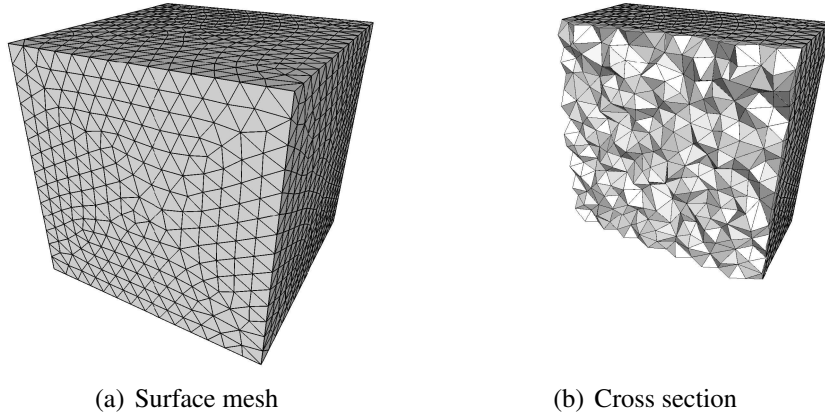


Figure 1: Unit cube.

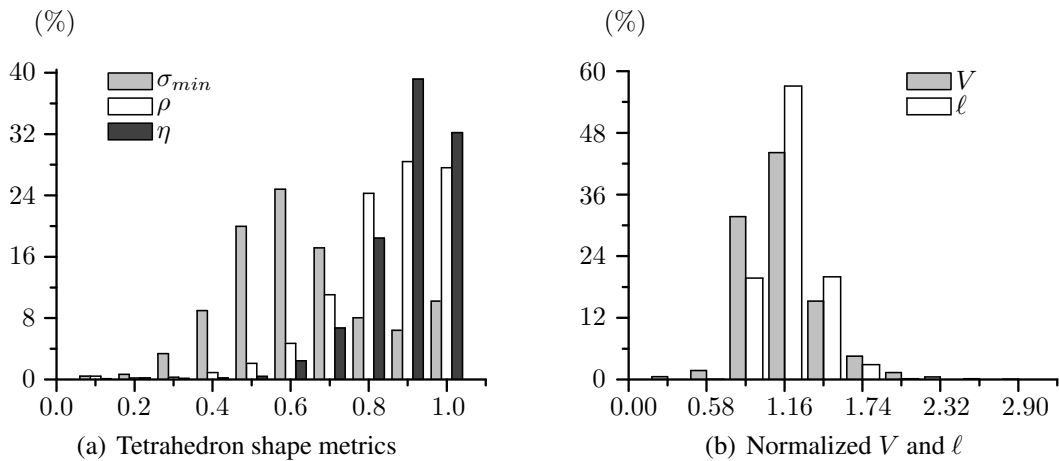


Figure 2: Distributions of tetrahedron shape metrics ( $\sigma_{min}$ ,  $\rho$ , and  $\eta$ ) and normalized volumes ( $V$ ) and edge lengths ( $\ell$ ) corresponding to the tetrahedral mesh of the unit cube.



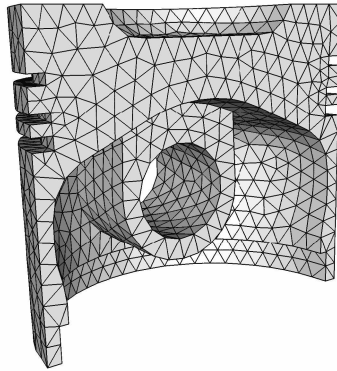
Table 1: Statistical data from the tetrahedral mesh of the unit cube.

	Non-Normalized				Normalized		
	min	$\mu$	max	$\sigma$	min	max	$\sigma$
$\ell$	$2.50 \times 10^{-2}$	$8.33 \times 10^{-2}$	$1.56 \times 10^{-1}$	$1.67 \times 10^{-2}$	$3.00 \times 10^{-1}$	1.881	$2.00 \times 10^{-1}$
$V$	$4.25 \times 10^{-8}$	$5.50 \times 10^{-5}$	$1.56 \times 10^{-4}$	$1.53 \times 10^{-5}$	$7.72 \times 10^{-4}$	2.834	$2.79 \times 10^{-1}$
$\sigma_{min}$	$8.49 \times 10^{-4}$	$5.92 \times 10^{-1}$	1.000	$1.92 \times 10^{-1}$	$1.43 \times 10^{-3}$	1.688	$3.25 \times 10^{-1}$
$\rho$	$4.57 \times 10^{-6}$	$8.03 \times 10^{-1}$	1.000	$1.52 \times 10^{-1}$	$5.69 \times 10^{-6}$	1.244	$1.89 \times 10^{-1}$
$\eta$	$9.70 \times 10^{-3}$	$8.45 \times 10^{-1}$	1.000	$1.20 \times 10^{-1}$	$1.14 \times 10^{-2}$	1.183	$1.42 \times 10^{-1}$

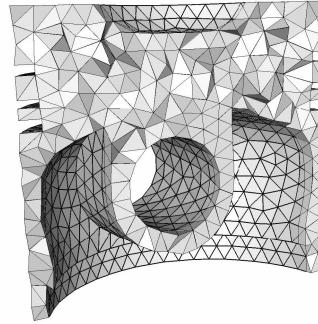
### 3.2 Piston

Figure 3(a) illustrates the surface mesh of a piston constituted of 1497 vertices and 2994 faces. The CPU time to generate the entire tetrahedral mesh was 13.36 seconds, where 11.09 seconds were spent at Stage 1 and 2.27 seconds at Stage 2. The final mesh consists of 1986 vertices, 10648 edges, 15827 faces and 7165 elements. Stages 1 and 2 have generated 65% and 35% of the mesh elements, respectively. Figure 3(b) illustrates a cross section of the tetrahedral mesh. In this figure, it may be observed an uniform distribution of  $h_E$ .

The ratio between maximum and minimum values of  $\ell$  was about 5 (five), cf. Tab. 2. As in the first example, flat elements were generated to ensure the convergence of the algorithm (see minimum volume in Tab. 2). Approximately 58%, 95% and 98% of the elements presented  $\sigma_{min}$ ,  $\rho$ , and  $\eta$ , respectively, greater than 0.50. Nearly 70% of the edges has normalized edge lengths  $\ell$  ranging from 0.56 to 1.12 (see Fig. 4(b)).



(a) Surface mesh



(b) Cross section

Figure 3: Piston (courtesy of Drexel Repository).

Table 2: Statistical data from the tetrahedral mesh of the piston.

	Non-Normalized				Normalized		
	min	$\mu$	max	$\sigma$	min	max	$\sigma$
$\ell$	$2.29 \times 10^{-3}$	$6.27 \times 10^{-3}$	$1.19 \times 10^{-2}$	$1.32 \times 10^{-3}$	$3.66 \times 10^{-1}$	1.896	$2.11 \times 10^{-1}$
$V$	$1.91 \times 10^{-13}$	$2.38 \times 10^{-8}$	$6.56 \times 10^{-8}$	$7.37 \times 10^{-9}$	$8.03 \times 10^{-6}$	2.757	$3.09 \times 10^{-1}$
$\sigma_{min}$	$1.47 \times 10^{-5}$	$5.41 \times 10^{-1}$	1.000	$1.75 \times 10^{-1}$	$2.73 \times 10^{-5}$	1.846	$3.24 \times 10^{-1}$
$\rho$	$2.51 \times 10^{-9}$	$7.74 \times 10^{-1}$	1.000	$1.55 \times 10^{-1}$	$3.24 \times 10^{-9}$	1.291	$2.00 \times 10^{-1}$
$\eta$	$5.84 \times 10^{-4}$	$8.18 \times 10^{-1}$	1.000	$1.25 \times 10^{-1}$	$7.13 \times 10^{-4}$	1.222	$1.53 \times 10^{-1}$

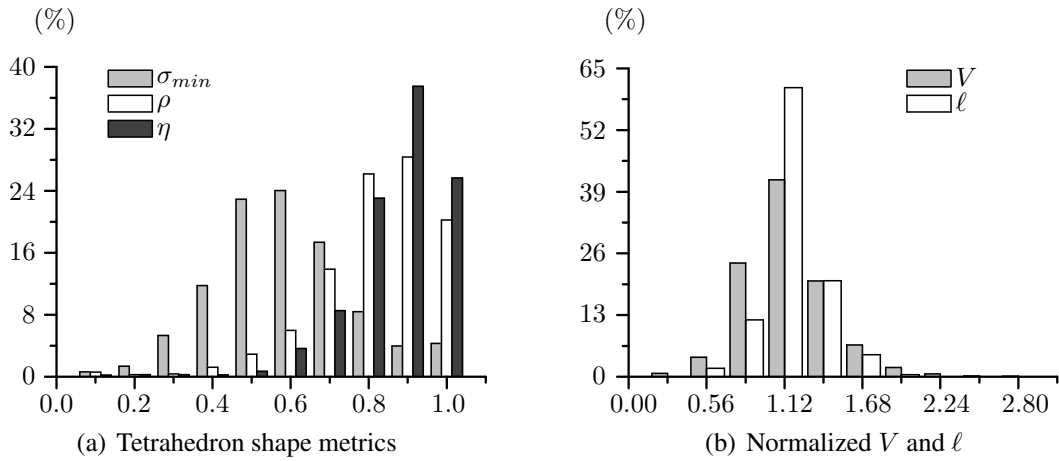


Figure 4: Distributions of tetrahedron shape metrics ( $\sigma_{min}$ ,  $\rho$ , and  $\eta$ ) and normalized volumes ( $V$ ) and edge lengths ( $\ell$ ) corresponding to the tetrahedral mesh of the piston.

### 3.3 LH Mount

The surface mesh of a LH mount, shown in Fig. 5, has 1781 vertices and 3578 faces. The total CPU time to generate the entire mesh was 8.32 seconds, where 4.26 seconds were spent at Stage 1 and 4.06 seconds at Stage 2. The final tetrahedral mesh has 1939 vertices, 9848 edges, 14021 faces and 6116 elements, from which 46% were created at Stage 1 and 54% at Stage 2. Obviously, the increase in the number of generated elements at Stage 2, in relation to the other examples, has reduced the mesh quality.

As it can be inferred from Tab. 3, the ratio between the maximum and minimum values  $\ell$  was approximately 14 (fourteen). Mean values of  $\sigma_{min}$ ,  $\rho$  and  $\eta$  were 0.50, 0.74 and 0.79, respectively (see Tab. 3). These values are less than most of the values previously shown.

About 50%, 92% and 96% of the tetrahedra has values of  $\sigma_{min}$ ,  $\rho$  and  $\eta$ , respectively, greater than 0.5 (cf. Fig. 6(a)). Nearly 63% of the elements and 80% of the edges have normalized volumes and edge lengths between 0.59 e 1.18 (see Fig. 6(b)).

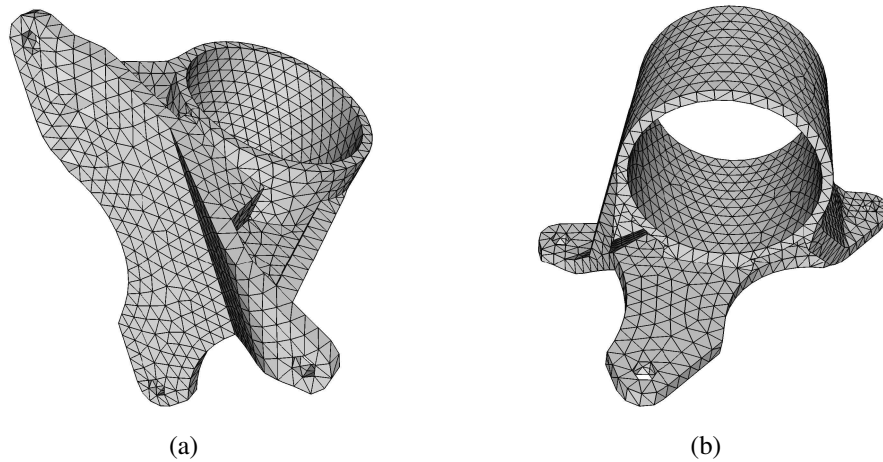
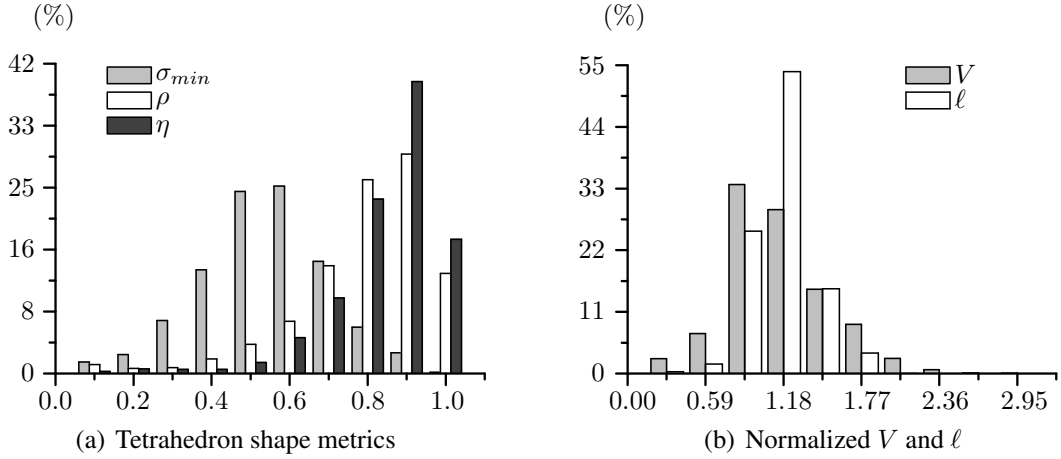


Figure 5: LH mount (courtesy of Eric Saltel).

Table 3: Statistical data from the tetrahedral mesh of the LH mount.

	Non-Normalized				Normalized		
	min	$\mu$	max	$\sigma$	min	max	$\sigma$
$\ell$	$5.30 \times 10^{-3}$	$4.09 \times 10^{-2}$	$7.41 \times 10^{-2}$	$9.10 \times 10^{-3}$	$1.29 \times 10^{-1}$	1.810	$2.22 \times 10^{-1}$
$V$	$1.44 \times 10^{-9}$	$6.40 \times 10^{-6}$	$1.87 \times 10^{-5}$	$2.40 \times 10^{-6}$	$2.25 \times 10^{-4}$	2.930	$3.75 \times 10^{-1}$
$\sigma_{min}$	$1.16 \times 10^{-4}$	$4.94 \times 10^{-1}$	$9.29 \times 10^{-1}$	$1.60 \times 10^{-1}$	$2.36 \times 10^{-4}$	1.881	$3.25 \times 10^{-1}$
$\rho$	$1.54 \times 10^{-7}$	$7.42 \times 10^{-1}$	$9.95 \times 10^{-1}$	$1.72 \times 10^{-1}$	$2.07 \times 10^{-7}$	1.342	$2.32 \times 10^{-1}$
$\eta$	$3.91 \times 10^{-3}$	$7.90 \times 10^{-1}$	$9.96 \times 10^{-1}$	$1.39 \times 10^{-1}$	$4.95 \times 10^{-3}$	1.261	$1.76 \times 10^{-1}$


 Figure 6: Distributions of tetrahedron shape metrics ( $\sigma_{min}$ ,  $\rho$ , and  $\eta$ ) and normalized volumes ( $V$ ) and edge lengths ( $\ell$ ) corresponding to the tetrahedral mesh of the LH mount.

### 3.4 Computational Cost

Early researches on advancing front based algorithms for generation of unstructured tetrahedral meshes (Löhner, 1988; Peraire and Peiró, 1992) report a computational cost  $\mathcal{O}(N \log(N))$ ,  $N$  being the number of generated elements, independently of the domain geometry. Jin and Tanner (1993) developed an algorithm whose computational cost changes according to different domain discretizations. They observed  $\mathcal{O}(N^{1.15} \log(N))$  for a unit cube, and  $\mathcal{O}(N \log(N))$  for an aircraft discretization. Later, Möller and Hansbo (1995) concluded that  $\mathcal{O}(N^p \log(N))$ , with  $p$  slightly greater larger than 1.0, was a more realistic measure of the computational cost for this type of algorithm.

In fact, algorithms that employ special data structures to speed up search operations, such as *heap* (Löhner, 1988), *octree* (Löhner, 1988), and *alternating digital tree* (Bonet and Peraire, 1991), are strongly dependent on the way the generation front advances (Möller and Hansbo, 1995). In addition, intersection tests between triangular faces also contribute to the increase of the total cost (Löhner and Parikh, 1988; Löhner, 2001).

The obtained computational cost of the proposed algorithm was evaluated considering the CPU time spent in ten different discretization levels of a unit cube ( $h_E = 1/5, 1/6, \dots, 1/14$ ). Figure 7 shows the CPU times for generation of each discretization level. The computational cost of the proposed algorithm for the unit cube was  $\mathcal{O}(N^{1.18} \log(N))$ , which is more close to that predicted by Möller and Hansbo (1995).

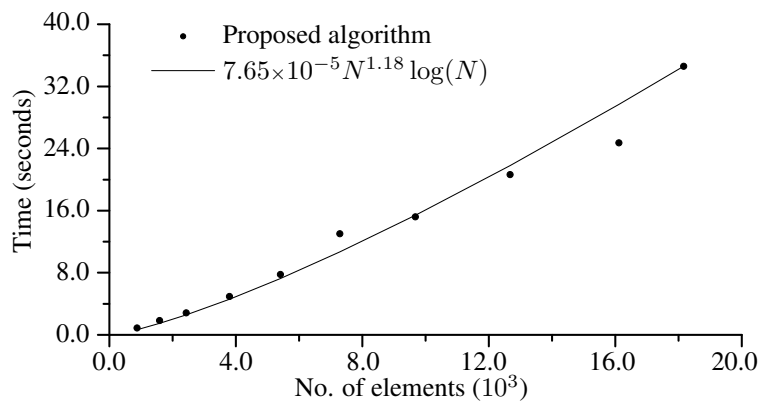


Figure 7: CPU times corresponding to ten levels of discretization of a unit cube ( $h_E = 1/5, 1/6, \dots, 1/14$ ).

#### 4. CONCLUSION

An advancing front based algorithm to generate unstructured tetrahedral meshes was described. The goal of this algorithm is to generate as much well shaped elements as possible, enforcing convergence by filling up distorted void spaces with poorly shaped elements. The presented examples show that the algorithm reached a reasonable degree of stability. Good quality meshes were obtained, even without the use of any external technique to improve convergence. Moreover, the ability of the present algorithm in dealing with arbitrary geometries was clearly observed. The computational cost was also satisfactory, in accordance with previously observed results. Therefore, it may be concluded that this algorithm can be used as a core of more complex mesh generators based on the advancing front method.

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