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Comparison of 3D Deformable Models For in vivo Measurements of Mouse Embryo from 3D Ultrasound Images

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Abstract

In this paper, we are interested in the analysis of the 3D shape of mouse embryo from 3D ultrasound (US) images acquired using an experimental ultrasonic system. Two approaches for the 3D segmentation of mouse embryo are evaluated. The first one is the simplex-mesh deformable surface and the other one is a deformable elastic template. The methods are compared qualitatively by evaluating the segmentation results and quantitatively by computing shape parameters.

Keywords: 3D segmentation, deformable models, ultrasound imaging, small animals imaging

1 Introduction

As 3D ultrasound images become available, there is a growing need for the development of image analysis tools in order to automatically detect specific structures, extract their 3D shape and compute shape characteristics (volumes, dimensions...) [1],[6]. In the context of this paper, we are interested in the analysis of the 3D shape of mouse embryo from 3D ultrasound (US) images acquired using an experimental ultrasonic system. Automatic segmentation of structures from images is a difficult and still opened problem in the image processing field. Among the existing segmentation methods, the deformable model based techniques rely on a prior shape model of the structure. These methods have been widely used in various contexts, especially in medical image analysis. However, the methods have to be adapted according to the specific properties of ultrasound images (low signal to noise ratio, presence of speckle, not well defined structure boundaries). From the anatomical point of view, at least two structures (the embryo and the placenta plus amniotic liquid) can be extracted separately or even simultaneously by iteratively deforming the prior 3D anatomical model.

In this paper, we evaluate two deformable model techniques for the extraction of embryo in 3D US images. The first one is the simplex-mesh based deformable surface model (SMD method) [2],[5]. The second model is named the Deformable Elastic Template (DET method) [3],[4]. The main difference between the two methods is that in the SMD model, surface boundaries of structures

are represented by simplex meshes on which image forces are applied while in the DET method, both boundary and interior of the objects are modeled by an elastic material. Both methods rely on the minimization of energy functional but respective energy minimization schemes differ.

Methods comparison is conducted on 3D phantom images and 3D images from *in vivo* acquisitions of mouse tumor and embryo. The segmentation results obtained by the two methods are compared as well as shape characteristics such as volume and object principal dimensions.

2 Material and methods

2.1 Ultrasound data

- *Phantom image*

Acquisition has been conducted to image a 3D phantom containing an ellipsoid inclusion (egg shape). It is composed of a mixture of agar (4% by weight) and gelatin (8%). The acquisition was performed using a 5.5-8MHz US system with a 2D probe mounted on a motorized table to perform a 3D acquisition.

- *Mouse imaging*

Acquisition of a tumor and an embryo were performed *in vivo* on mice using a 20MHz transducer displaced mechanically. 3D data were acquired as a series of 2D images, incrementing the position of the probe using a step by step motor.

Some parameters of the raw data volumes are summarized in Tab. 1. As required by the segmentation methods presented in the next section, isotropic volumes were generated using a linear or cubic interpolation method. Dimension and resolution of these new volume data are given in the same table.

2.2 3D segmentation method

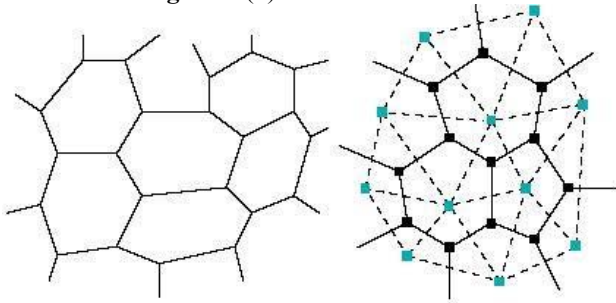
- **SMD method**

According to [2], SMD method belongs to the category of discrete deformable models as objects surface is represented by discrete simplex meshes. A simplex mesh [7] is a discrete model representation with a constant vertex connectivity. 2-simplex meshes are used to represent surfaces.

Tab. 1 Parameters of input data

3D Raw data				
-		egg-phantom	tumor	embryo
Dimension (pixel)	Nx	128	256	256
	Ny	128	256	256
	Nz	70	100	100
Resolution (mm/pixel)	Δx	0.46	0.054	0.054
	Δy	0.46	0.020	0.020
	Δz	0.46	0.160	0.160
3D Isotropic data				
Dimension (pixel)	Nx	128	296	296
	Ny	128	94	94
	Nz	70	295	295
Resolution (mm/pixel)	Δ	0.46	0.054	0.054

Each vertex of 2-simplex meshes is connected to three, and only three neighbors (as shown in **Figure 1 (a)**). 2-simplex meshes are topologically dual to triangulations as illustrated in **Figure 1 (b)**.



(a) 2-simplex mesh

(b) triangulation dual to a simplex mesh

Figure 1 Simplex mesh

The surface representation affects the possible evolution of a model. As with most deformable models, the surface evolution in the SMD method is driven by the minimization of an energy functional E , which is composed of 2 terms including an intrinsic regularizing term E_{reg} and a data term E_{ext} . The energy of a surface S is defined by:

$$E(S) = E_{reg}(S) + E_{ext}(S) \quad (1)$$

If v_i denotes one vertex of S , equation (1) derives in:

$$\frac{d^2 v_i}{dt^2} = \gamma \frac{dv_i}{dt} + \alpha f_{int}(v_i) + \beta f_{ext}(v_i) \quad (2)$$

In this case, the internal force is a curvature regularity term computed by averaging the simplex angle of the surface at each vertex over a neighborhood. The simplex angle is a geometric parameter intrinsic to the simplex mesh representation that is related to the notion of curvature. It ensures that the modeled object will have a smooth surface and no sharp edges.

The external force is computed at each vertex by tracking the highest gradient value detected in the image along the surface normal. Thus, it is proportional to the distance to image contour points. The image gradient is estimated by a simple Sobel convolution operator.

The weights α and β tune the relative strength of the internal and the external forces while the damping parameter γ is used to optimize the numerical scheme convergence speed.

• DET method

In the DET method [1],[3],[4], the object is described by an elastic body defined by Young modulus (E) and Poisson coefficient (μ).

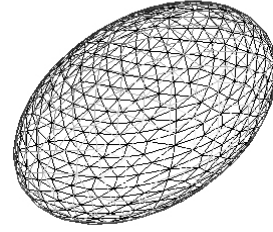


Figure 2 Initial template: a 3D ellipsoid

In the context of this study, a 3D ellipsoid (see **Figure 2**) is taken as the prior model while being considered as a linear elastic body. The equilibrium of the model is obtained through the minimization of the following global energy functional:

$$E = E_{elastic} + E_{data} \quad (3)$$

In which, $E_{elastic}$ represents the elastic deformation energy of the model (the 3D ellipsoid), and E_{data} is the energy due to the external image forces. Their respective expressions are:

$$E_{elastic}(u) = \frac{1}{2} \int_{\Omega} (Su)^T \mathbf{D}(Su) d\Omega \quad (4)$$

$$E_{data}(u) = - \int_{\Gamma} t \cdot u d\Gamma \quad (5)$$

In which, S is a differential operator, \mathbf{D} is the elasticity matrix, and u is the displacement vector, with Γ the border of the object domain Ω . The boundary force field \mathbf{t} is either derived from the gradient of a potential function P ($\mathbf{t} = -grad(P)$) computed from an edge map, or a specific force field called gradient vector flow (GVF) [8] which is sometimes more efficient regarding the initialization and the convergence to edges.

The minimization process which drives the initial ellipsoid to coincide with the data relies on the Finite Element Method.

2.3 Method Comparison

The shared characteristics by the SMD and DET methods are:

- 1) Both methods rely on the minimization of an energy function to reconstruct the object as the segmentation result;
- 2) An initial shape (surface or ellipsoid) is needed and manually positioned in the data/image;
- 3) Boundary force field is needed to deform the model.

The differences are:

- 1) SMD is a surface model, which means only the surface of the object is constructed; while DET is a volume model, including the object's interior in addition to the surface;
- 2) Based on 3D elasticity theory framework, DET has some physical parameters like E , μ to describe the elastic body;
- 3) Minimization process is different. DET has clearer physical meaning.

3 Experimental results

Results on Egg phantom

Results obtained on egg phantom data are shown in Figure 3, including two different orthogonal projection images and 3D presentation of the final results.

Parameters of SMD are often changed during processing by manual interaction. They concern the types of transformation, locality parameter, threshold for force field map, etc.; sensitive parameters of DET are adjusted according to successive tests. For all the experiments, a canny operator was used to compute the potential image. Force factor, which weights the external energy part, is set to 50 and displacement threshold (stopping criteria) is set to 0.010 (voxel). Quantitative evaluation is shown in Tab. 2. The results obtained by both methods are close to the true sizes.

Results on Mouse Tumor

Results obtained with tumor data are shown in Figure 4. With the DET method, force factor is set here to 100. Displacement threshold is set to 0.005.

As shown in Tab. 3, the quantitative evaluation presents significant differences between the two methods. DET method tends to give smaller measures than the SMD model. The true tumor's sizes lie in between the two results considering the experts' opinions.

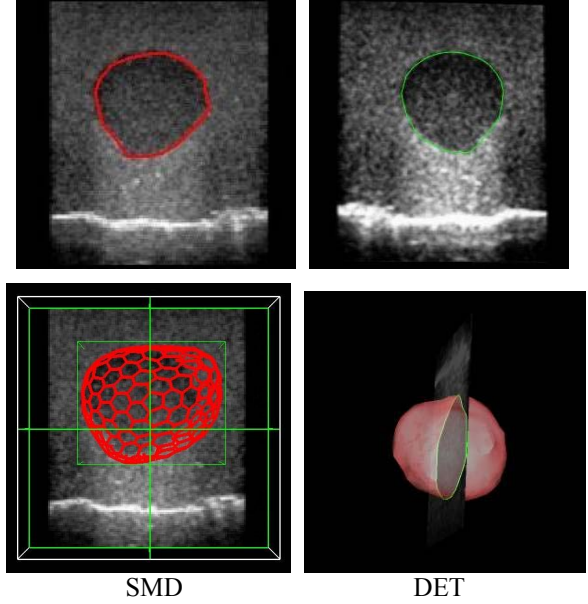


Figure 3 Results on egg phantom data

Tab. 2 Quantitative comparison with phantom image

Method	Volume (cm ³)	X (mm)	Y (mm)	Z (mm)
Reference	10.60	33	-	26
SMD	11.31	33.28	27.75	27.39
DET	10.90	31.76	28.16	25.92

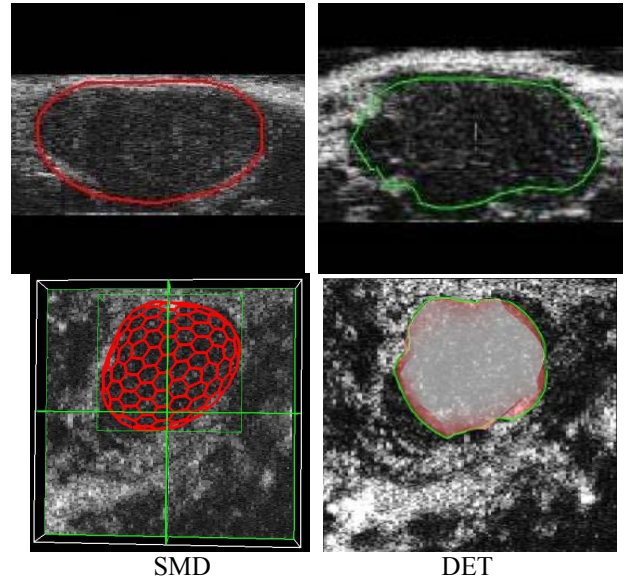


Figure 4 Results on tumor data

Tab. 3 Quantitative comparison with tumor data

Method	Volume (mm ³)	X (mm)	Y (mm)	Z (mm)
SMD	173.25	8.34	4.33	8.73
DET	133.25	8.45	3.85	8.16

Results on mouse embryo

Results with embryo data are shown in Figure 5. In this test, images are incomplete and only part of the embryo is visible. So neither SMD nor DET can obtain good results except for some partial borders. With the DET method, force factor is set to 150 and displacement threshold is set to 0.005.

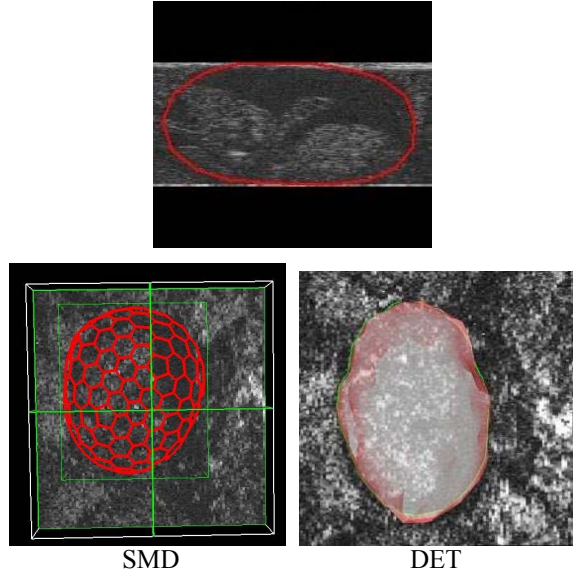


Figure 5 Results on embryo data

Tab. 4 Quantitative comparison with embryo image

Method	Volume (mm ³)	X (mm)	Y (mm)	Z (mm)
SMD	309.34	11.00	5.00	9.57
DET	246.75	12.80	4.65	8.81

4 Discussion and Conclusions

There is a clear difference in image quality between egg phantom images, tumor and especially embryo images. The contours of the embryo are in fact very difficult to identify even by eye.

Results on phantom images are pretty similar with the two methods. One can say that generally DET method tends to give smaller measures than the SMD model and is slightly closer to the reference values in this particular test (10.9cm³ vs. 11.31cm³ compared with 10.60cm³).

The same tendency is observed with mouse tumor and embryo, but the differences are more significant especially for volume estimation with 133.25 cm³ (DET) vs. 173.25 cm³ (SMD) for tumor, and 246.75cm³ (DET) vs. 309.34cm³ (SMD) for embryo. Both methods have the initialization and convergence problems with in vivo images.

Manual interactions are usually required with the SMD model and made very efficient within a clever and intuitive user-friendly interface. As no interaction was made possible here, DET method automatically performed the segmentation after a manual initialization. Indeed, corrections are still required with in vivo data

with both methods, unless images of better quality are available, and US specific force field is proposed. Also, a judicious combination of the two approaches could certainly allow to obtain accurate segmentation results within a reduced time. This will be the part of our future work.

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