

Modified technique for volumetric brain tumor measurements

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ABSTRACT

Quantitative measurements of tumor response rate in three dimensions (3D) become more realistic with the use of advanced technology imaging during therapy, especially when the tumor morphological changes remain subtle, irregular and difficult to assess by clinical examination. These quantitative measurements depend strongly on the accuracy of the segmentations methods used. Improvements on such methods yield to increase the accuracy of the segmentation process. Recently, the essential modification in the Traditional Region Growing (T-RG) method has been developed and a “Modified Region Growing Method” (MRGM) has been presented and gives more accurate boundary detection and holes filling after segmentation. In this paper, the new automatic calculation of the volumetric size of brain tumor has been implemented based on Modified Region Growing Method. A comparative study and statistical analysis performed in this work show that the modified method gives more accurate and better performance for 3D volume measurements. The method was tested by 7 fully investigated patients of different tumor type and shape, and better accurate results were reported using MRGM.

Keywords: Region Growing, Modified Region growing, and Volumetric Brain Tumor Measurements

1. INTRODUCTION

More recent studies have shown that 3D quantitative imaging-based method of tumor size assessment using MRI is highly accurate in determining actual tumor size [1,2] and may be superior to clinical palpation in predicting local tumor control [3,4,5].

Manual region of interest (ROI) volumetry method is a standard approach of 3D quantitative measurements which is very precisely to detect tumor contours and it is considered to be the “gold standard” because region of

interest (ROI) is segmented manually by the expert radiologists. The disadvantage of this method is, it requires intensive time because of its dependency on manual segmentation process. Segmentation of ROI in volumetric medical images is still a challenging problem, and solutions usually have been based on either model-based deformation of templates or intensity thresholding such as region growing method [6,7]. Recent studies prove that the region growing is an effective approach and less computation intensive for image segmentation especially for the homogenous regions [8,9,10,7,11]. The primary disadvantage of region growing method is the partial volume effect [12,13]. Partial volume effect limits the accuracy of MR brain image segmentation. It blurs the intensity distinction between tissue classes at the border of the two tissues types, because the voxel may represent more than one kind of tissue types [14,15]. S. Lakare [12] *et. al*, developed effective modifications in region growing technique. This modification called Modified region growing method (MRGM) used to remove the partial volume effects and to incorporate gradient information for more accurate boundary detection and filling holes occurred after segmentation.

The software implemented in this paper involves the proposed quantitative measurement of brain tumors based on MRGM segmentation and the visualization tool to monitor and reconstruct the brain tumor in 2D and 3D space. For testing and validation, the proposed MRGM method has been compared with traditional region growing method against experts’ manual trace method, and the statistical analysis was performed to evaluate the proposed method against TRG and golden tracing method by experts.

In this paper, section 2 describes the settings of MR image acquisition, the details of patient population and noise reduction technique as data pre-processing. In section 3, we present the segmentation and calculation used for assessment of the brain tumor measurements, also this section describes the statistical and data analyses used to evaluate and validate the proposed method. Section 4 describes the results of brain tumor visualization in 2D and 3D spaces, the brain volume calculation using the proposed method compared with other method, and the result of statistical and data analysis. Section 5 presents the merits and demerits of the proposed method

compared to each others and concludes all over the work done in this study.

2. MATERIALS

2.1. MRI Image Acquisitions

MRI images were acquired on a 1.5T using T1-weighted contrast images. A resolution of 256x256x60 with a voxel resolution of 0.93mm x0.93mm with slice thickness of 3mm was set.

2.2. Patient Population

The study group consists of seven patients scanned with 228 MRI axial slices with biopsy histologically proved Glioblastoma Multiforme (GBM) and Low Grade Astrocytoma brain tumors types.

2.3. Pre-processing

Noise presented in the image can reduce the capacity of the region growing filter to grow large regions, or may result in a fault edges. When faced with noisy images, it is usually convenient to pre-process and enhance the image by using a noise reduction filter. Gaussian smoothing filter [16] is commonly used as an approach for noise reduction. The size of the neighborhood mask can be set by the user. The quality of the enhanced Gaussian filtered images is much better as the contrast between tumor and surrounding tissue is high as well tumors studied are of homogenous borders (regular convex shapes).

3. METHODS

3.1. Traditional Region Growing Segmentation Method

The Traditional region growing algorithm based on extraction of a connected set of pixels whose pixel intensities are consistent with the pixel statistics of a seed point. The mean and variance across 8-connected neighborhood are calculated for a seed point [16,17].

3.2. Modified Region Growing Segmentation Methods (MRGM)

To understand the basic principles behind MRGM, we first reviewed the S. Lakare, *et al.* [12]. In their work, MRGM provided for object segmentation has been implemented. This implementation allowed stable boundary detection when the gradients suffer from intersection variations and gaps.

3.3. Volume Calculation

3.3.1. Manual ROI Volumetry

Figure 1 shows the area inside the outline that was manually segmented, labeled, calculated, and multiplied by the MR slice thickness plus the interslice gap to calculate a per-slice tumor volume. The total tumor volume was obtained by summing the volume calculations for all slices.

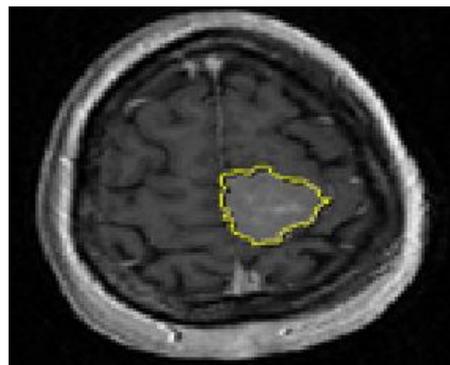


Figure 1. Manual trace method

3.3.2. TRG and MRGM Calculations for Growing Tumor

After tumor region has been segmented using both T-RG and MRGM segmentation techniques, the tumor volume calculations are performed in this segmented region. To calculate the volume of segmented tumor region, the automatic labeling of the entire volumetric tumor region has been done slice by slice and by calculating the total number of pixels into the labeled regions. Areas of the labeled region were calculated and multiplied by the MR slice thickness plus the interslice gap to obtain a per-slice tumor volume. The total tumor volume was then obtained by summing the tumor-bearing slices.

3.4. Statistical Consideration and Data Analysis

The comparative study has been done using T-RG, MRGM and Manual ROI volumetry methods. Two observers (expert radiologist and a none radiologist) independently rated each MR image twice by using manual and the two automated measurement methods. Observers have performed the comparison between inter and intra-observer reliability and image processing computational times for both methods were reported. To compare the intra and inter-observer reliability of the three measurement methods, we used the following agreement index [21], AI as follow:

$$AI=1 - \frac{|X_a - X_b|}{(X_a + X_b)/2} \quad (1)$$

For inter-observer agreement calculation, X_a was the measurement obtained by observer 1 and X_b , the measurement obtained by observer 2 with the same technique on the same case. For intra-observer agreement calculation, X_a is the measurement made during the first trial, and X_b is the measurement made during the second trail by the same observer with the same technique on the same case. Intra-observer and inter-observer agreement indexes were calculated for each image, to increase sensitivity performance. A value of "1" indicates the perfect agreement and value of "0" indicates no agreement.

4. RESULTS

Results showed that the proposed quantitative measure-

ment of brain tumors based on MRGM has a higher accuracy and precision against traditional region growing method compared to expert manual computation. This yields to better effect in the assessment of brain tumor measurements.

4.1. PC Based Package

We improve our previous work [10] PC based software package implemented using three programming development environment, as VTK[18], ITK [20] and Visual C++, to segment, visualize and calculate the tumor volume at different instants of tumor growing or shrinking. **Figure 2** shows the result of tumor segmentation using T-RG and MRGM. **Figure 3** shows the 3D reconstruction of segmented tumor region for MRGM method, using surface reconstruction algorithm [19].

4.2. Tumor Volume Measurement Accuracy

The Relative Error (RE) for tumor volume can be calculated as Where P_q tumor volume using 3D quantitative methods (Traditional and modified region growing methods), P_m is tumor volume calculated using expert

$$RE(\%) = \left(\frac{P_q - P_m}{P_m} \right) \times 100 \quad (2)$$

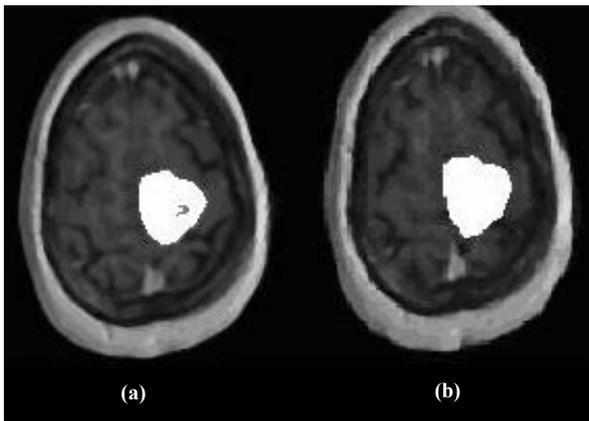


Figure 2. (a) Results of T-RG segmentation; (b) Results of MRGM segmentation

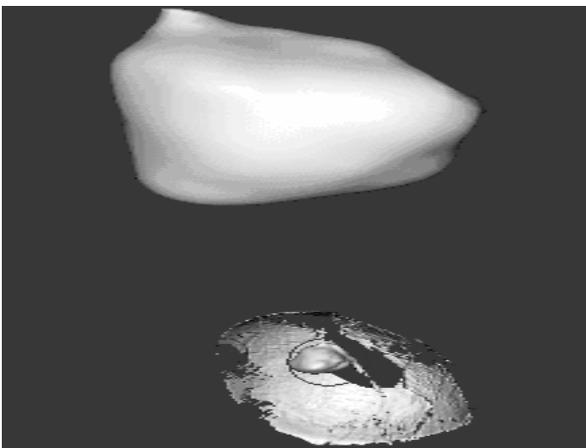


Figure 3. Extracted tumor in 3D space

manual tracing method. **Table 1** shows calculation results and their relative errors for the different quantitative methods compared with the gold standard manual method. These results had been obtained by *observer 1* and **Figure 4** summarizes these relative errors in chart.

Also, **Table 2** shows calculation results and their relative errors for the different methods compared with the standard manual method. These results obtained by *observer 2* and **Figure 5** summarizes these relative errors in chart.

4.3. Observer Agreements

The intra and inter-observer agreement indexes for the two observers are summarized in **Table 3**. As shown in the table, there are no significant differences in mean intra and inter-observer agreement between the manual method, traditional and modified region growing methods.

Table 1. Volumetric calculation for brain tumor in cm³ using the different calculation methods and the relative errors for each method compared with manual segmentation method (data rated using first observer)

Cases	Volume in cm ³			Relative Error%	
	T-RG	MRGM	Manual	T-RG	MRGM
Case#1	5.587	6.281	6.125	8.7836	2.546
Case#2	8.782	9.845	10.234	14.188	3.801
Case#3	11.406	12.066	11.234	1.5310	7.406
Case#4	13.687	12.755	12.987	5.3900	1.786
Case#5	14.014	14.284	14.123	0.7717	1.139
Case#6	15.223	16.116	16.987	10.384	5.127
Case#7	16.122	16.695	16.354	1.4186	2.0851

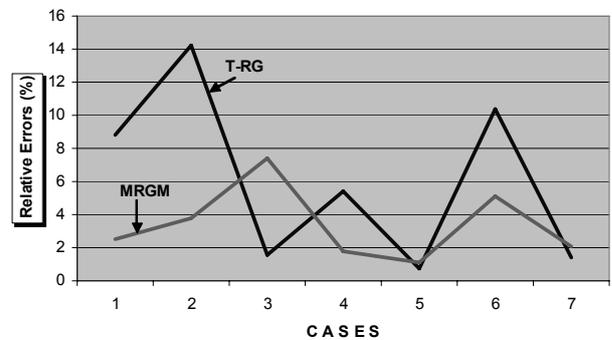


Figure 4. Relative errors for T-RG and MRGM compared with manual segmentation method (data rated using first observer)

Table 2. Volumetric calculation for brain tumor in cm³ using the different calculation methods and the relative errors for each method compared with manual segmentation method (data rated using second observer)

Cases	Volume in cm ³			Relative Error%	
	T-RG	MRGM	Manual	T-RG	MRGM
Case#1	5.763	6.223	5.987	3.7414	3.941
Case#2	9.088	9.637	10.345	12.150	6.843
Case#3	10.836	12.963	11.897	8.9182	8.960
Case#4	13.962	12.597	12.235	14.115	2.958
Case#5	13.509	14.108	14.678	7.9643	3.883
Case#6	14.73	16.122	16.544	10.964	2.550
Case#7	14.831	16.363	16.786	11.646	2.519

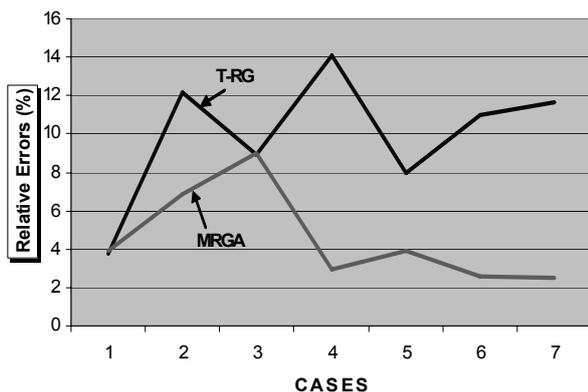


Figure 5. Relative errors for T-RG and MRGM compared with manual segmentation method (data rated using second observer)

Table 3. Results of intra- and inter observer agreement index

Index	Observer	MRGM	T-RG	Manual
Intra-Observer Agreement	1	0.9899 ±0.019	0.9374 ±0.058	0.9665 ±0.0183
	2	0.9773 ±0.026	0.9728 ±0.0215	0.9744 ±0.0122
Inter-Observer Agreement	1,2	0.9689 ±0.0342	0.9587 ±0.0207	0.9559 ±0.0237

5. DISCUSSION AND CONCLUSION

Recent attention has been given to improve the segmentation methods in order to increase brain tumor measurements accuracy [8]. In this paper the essential modifications has been implemented in the region growing algorithm and presented as “Modified Region Growing Method” (MRGM). These modifications overcome the partial volume effect artifacts. Hence, applying MRGM to segment brain tumors will increase the accuracy of the volumetric measurements of the brain tumors. A volumetric measurement based on T-RG, MRGM and manual segmented ROI volumetry method have been applied in MR volumetric data total of 228 MRI slices of Glioblastoma Multiforme (GBM) and Low Grade Astrocytoma brain tumors scanned from 7 patients. The results of comparative study showed that the MRGM produces lower relative errors than T-RG method. Also, it has a no significant differences in inter- and intra-observer agreement index. These results ensure that the accuracy of the volumetric measurements of the brain tumor have been improved using MRGM which yields to great effects in many applications in tumor prognosis and therapy such as early signs of treatment failure in radiotherapy and chemotherapy to avoid unneeded higher doses of radiation to patient, tumor growth rate and early detection of tiny changes in tumor size in which it is difficult to be detected in traditional visual metric and clinical examination measurements.

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