A Review on Segmentation Algorithms of Oral Cancer Detection

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Abstract: Oral cancer is one of the complex diseases in the world. The rate of the death is increasing all over the world due to this type of malignancy. Early diagnosis of malignancy is essential to avoid risk of disease. Earlier detection of oral lesions aids surgeons to provide proper treatment to save lives. An oral tumor can be Malignant or benign. Benign tumor of the mouth is growth that does not spread (metastasize) to other parts of the body, malignant tumors are cancerous and they rapidly grows and metastasize to other parts of the body. Oral cancer detection and segmentation is one of the most challenging and time consuming task in medical image processing. This research is focused towards the comparative study of earlier proposed segmentation techniques discussed in the literature.

Keywords: Oral cancer, Medical Image Processing, preprocessing, Segmentation

INTRODUCTION:

Oral cancer constitutes 3rd common malignancy in developing countries. Dental technology offers improved solution to traditional dental problems and helps detecting oral problems at an early stage. Oral cancer is also known as mouth cancer ,it may arise as a primary lesion originating in any of the tissues in the mouth, there are several types of oral cancer ,but around 90% are squamous cell carcinomas, originating in the tissues that line the mouth and lips. oral or mouth cancer most commonly involves the tongue, it may also occur on the floor of the mouth, cheek lining ,gums, lips or hard and soft palate(roof of the mouth).most oral cancers look very similar under the microscope and called as squamous cell carcinomas, but less commonly other types of oral cancer occur such as melanoma.

Oral tumors are mainly classified as Benign and Malignant. Benign tumor of the mouth is growth that does not spread (metastasize) to other parts of the body, it is not cancerous and life threatening, where as malignant tumors are cancerous and they rapidly grows and metastasize to other parts of the body.

There are several image processing technique such as histogram equalization, image segmentation, image enhancement, morphological operation, feature extraction and classification.

II. LITERATURE REVIEW:

Nurtanio et al. [1], presented a segmentation system for cyst or tumor cases from dental X-ray images.In preprocessing, the colored images are transformed into gray images, and then apply Gaussian filter to smooth images then segmentation using the active contour models (snake) provides a solution for number of visual problems, which includes detection of lines, edges and subjective contours(snakes),those model were used to handle ROI, not the entire image .validation using receiver operating characteristic(ROC)analysis as a Performance measure.

Adv: snakes reduced the need for edge linking compared to traditional edge based segmentation

Anuradha and Sankaranarayanan [2], proposed a technique for mouth cancers detection utilizing a morphological mathematical watershed algorithm, this approach preserves the edge details, the marker controlled watershed is used for tumor segmentation purpose, and then the marker controlled watershed

segmentation algorithm was compared with the watershed segmentation, and the marker controlled watershed has better results. Preprocessing is carried out by linear contrast stretching of input(x ray) image, then marker-controlled segmentation is applied.

F., Abdolali, R. A., Zoroofi, Y., Otake, & Y., Sato [3], proposed automatic segmentation algorithm based on the symmetric axis analysis has proposed, which is an analysis of the left and right sides of each image. In this, automatic segmentation of tumor of CBCT images is used. In preprocessing, diffusion filtering is used to image enhancement and smoothing .for symmetry axis detection, an efficient method with modern feature points is used. Multi resolution hierarchical FFD is used in multi-scale registration with fastness and robustness. Using this approach high TP and low FP is obtained. Dice's coefficients, jaccard index and hausdorff distance are performance metrics of this automatic segmentation.

ADV: Multi resolution hierarchical FFD is used in multi-scale registration with fastness and robustness.

DisadV:FFD registration is computing Inverse FFD

Limitation of this method is incapability of handling cysts and lesion without specific boundary and cases with large asymmetric variations.

Maghsoudi et al. [4], proposed an intelligent system based on artificial neural networks. This proposed system used 120 patient images, 120 for training process and 30 for testing and validation, and four features were extracted and used for diagnosis of mouth legions. This work indicated that artificial neural network when employed in the diagnostic systems provided a powerful approach for prediction and diagnosis of dental and oral diseases. In this, diagnosis and prediction of lichen planus, leukoplakia and oral squamous cell carcinoma diseases have been studied using artificial neural network.

Anuradha and Sankaranarayanan [5], proposed an oral cancer detection system; it works by preprocessing images using Linear Contrast Stretching for noise removal, and then Watershed Segmentation is utilized. Because of Watershed segmentation problems, Marker Controlled Watershed segmentation is utilized, and Marker Controlled watershed segmentation decreases the over segmentation problem. The Improved Marker Controlled Watershed Segmentation algorithm obtains better segmentation result (higher speed) with or without linear contrast stretching of the image. The Processing Speed is calculated before stretching and after stretching.

M. K. Alsmadi, [6], proposed The hybrid Fuzzy C-Means and Neutrosophic approach which is used for segmenting jaw image and detecting the jaw lesion region in panoramic X-ray images which may help in diagnosing jaw lesions. Area error metrics are used to assess the performance and efficiency of the proposed approach from different aspects. Both efficiency and accuracy are analyzed. Specificity, sensitivity and similarity analyses are conducted to assess the robustness of the proposed approach. Comparing the proposed approach with the Hybrid Firefly Algorithm with the Fuzzy C-Means, and the Artificial Bee Colony with the Fuzzy C-Means algorithm, the proposed approach produces the most identical lesion region to the manual delineation by the Oral Pathologist and shows better performance.

Zhalong Hu, Abeer Alsadoon, Paul Manoranjan 1, P.W.C. Prasad, Salih Ali, A. Elchouemic[7], proposed system focused on image pre-processing and segmentation steps, using anisotropic diffusion and Fuzzy C-Means to enhance the quality of the image, then improve the accuracy of tumor detection and classification. The proposed algorithm can remove noise and improve image quality by keeping all the edges clearly visible. In the segmentation stage, the use of the fuzzy c-means is aimed at improving the accuracy and specificity which would affect results.

Sr	Authors	Segmentation	Results	Future enhancement
no		algorithms used		
		(input image)		
1	Nurtanio I,	Active contour	It guarantees continuous	
	Purnama IKE,	model(snake)	and smooth lesion	
	Hariadi M,	(dental	boundaries, it reduced the	
	Purnomo MH	panoramic(x-ray)	need for edge linking	
		image)	compared to conventional	
			segmentation methods,	
			accuracy value is 99.67%	
2	Anuradha K,	Marker-controlled	Overcome the problem of	To increase accuracy
	Sankaranarayanan	watershed	Over segmentation.	and speed
	К	Segmentation	speed obtained is 92.55%	Up to 99%.
		(dental x-ray	&Accuracy obtained is	
		images)	90.25%	
3	F., Abdolali, R.	automatic	High TP and low FP is	Develop methods to
	A., Zoroofi, Y.,	segmentation	obtained. it	handle cysts and
	Otake, & Y.,	algorithm based on	is effective for	lesions without
	Sato,	the symmetric axis	segmentation of various	specific boundary and
		analysis	cysts.	cases with large
		(CBCT image)		asymmetric variations.
				Future works will
				investigate other
				methods to reduce
				talsely classified
4				healthy tissues.
4	Maghsoudi R,	Artificial neural	Using ANN, oral and	
	Bagneri A,	network	dental diseases are	
	Magnsoudi MI		diagnosed and predicted	
5	Anuradha K	Improved marker	Speed of algorithm is	To increase speed and
5	Allulaulia K, Sankaranarayanan	controlled	calculated before	accuracy up to 90%
	K	Controlled	calculated before	accuracy up to 7770
		Negmentation	Stretching is 90.5% and	
	IX	algorithms	Stretching is 90.5% and after stretching 92.6%	
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		algorithms (dental x-ray images)	Stretching is 90.5% and after stretching 92.6%	
6	M. K. Alsmadi.	Segmentation algorithms (dental x-ray images) A hybrid fuzzy c-	Stretching is 90.5% and after stretching 92.6%	To develop a classifier
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6	M. K. Alsmadi,	Segmentation algorithms (dental x-ray images) A hybrid fuzzy c- means & Neutrosophic approach	Stretching is 90.5% and after stretching 92.6% It can be used to segment an find lesion accurately even in the low contrast and complicated jaw x-r	To develop a classifier to distinguish benign or malignant tumor of jaw in feature
6	M. K. Alsmadi,	Segmentation algorithms (dental x-ray images) A hybrid fuzzy c- means & Neutrosophic approach (dental x-ray	Stretching is 90.5% and after stretching 92.6% It can be used to segment an find lesion accurately even in the low contrast and complicated jaw x-r av images, it reduced	To develop a classifier to distinguish benign or malignant tumor of jaw in feature extraction
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III. COMPARISON OF VARIOUS SEGMENTATION ALGORITHMS:

7	Zhalong Hu,	Fuzzy c-means	Proposed method	To increase accuracy,
	Abeer Alsadoon,	segmentation	achieved	specificity, sensitivity
	Paul Manoranjan	(CT image)	accuracy	
	1, P.W.C. Prasad,		90.11%, specificity	
	Salih Ali, A.		87.5% &sensitivity	
	Elchouemic		92.15%	

IV. CONCLUSION:

In this paper the survey study of various oral cancer detection techniques in the medical image processing is done. The advanced Medical image processing techniques and algorithms are reviewd. A comparative study is made on various techniques. After evaluation of well-known technique it is clearly shown the various methods which can detect the tumor efficiently and provide accurate result. This work will be extended for new algorithm for Oral tumor detection which will provide more efficient result in terms of accuracy and speed than the existing methods in future.

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