

Comparative Study: Coronary Blood Vessel Segmentation in Angiogram Images

¹GaneshKumar .K and ²Dr. M. Mohamed Sathik,

¹PG Scholar Department of Computer Science, Sadakathullah Appa College, Tirunelveli, TamilNadu, India

²Principal, Sadakathullah Appa College, Tirunelveli, TamilNadu, India

Abstract—Coronary blood vessels can be visualized using various angiogram methods. It is difficult to analyze if images with low contrast and small blood vessels. So blood vessels are preprocessed and segmented for better medical analysis. In this paper we discuss about various techniques to segment coronary blood vessels.

Keywords—Coronary Blood Vessels, Angiography Images, Preprocessing, Segmentation

I. INTRODUCTION

World Health Organization informed, there are more than 12 million deaths due to cardiovascular disease, caused by coronary artery stenosis and blockage[12]. In medical research, angiography is a procedure to visualize blood vessels. There are many medical imaging methods like X-ray, Magnetic Resonance and Computed Tomography.

Interpretation to an angiogram is very difficult because of noise, foreshortening of vessels, small blood vessels and vessel overlap. Partitioning an angiogram image into vascular and background regions is the vital role of segmentation. There for developing accurate image segmentation methods for angiography are the challenges of research group.

There are various method to segment blood region from background such as thresholding, region based, clustering, level set, watershed, edge detection, centreline detection, match filter and classification.

II. METHODOLOGY

A. Pre-processing:

In Paper [10], image is pre-processed using Hessian matrix analysis and Frangi 2D filter. Coronary vessel structure found through Hessian matrix analysis. From Hessian matrix, Gaussian curvature is calculated by multiplying Eigen values. The Eigen vector shows the direction of curve. Frangi's 2D filter uses Eigen vectors to calculate likeness of region to vessel.

In paper [3], authors use Noise Adaptive Fuzzy Switching Median (NAFSM) filter to enhance angiographic images.

NAFSM [7] is used to remove salt and pepper noise from angiogram. Only noise pixel is identified by histogram in detection phase of NAFSM. In next phase it is replaced by its correlation term. That is from local information in the filter maximum absolute luminance difference is calculated. Fuzzy reasoning is applied to get maximum absolute luminance difference. This helps the filter to get accurate correlation term.

Based on Eigen vectors of Hessian matrix, Frangi's vessel detection technique is used to extract vessel branches.

In paper [4], low contrast images are preprocessed by using morphological top hat technique [5,13]. Top hat filtering technique computes the morphological open operation on image and then subtracts it from original image. It detects logical elevation on arbitrary background which can be used to enhance vessel part.

In paper [6], color image is convert to gray scale image. This gray scale image is feed into adaptive histogram equalization. Median filter with 75 x 75 is applied to subtract background from foreground.

B. Segmentation:

In paper [10], morphological operation is used to detect edges of vessel region of interest [2]. Firstly preprocessed image is converted into binary image then the image is filled, complemented and erode.

In paper [4], region growing is used to segment vessels from angiogram[8,9]. Authors have used input from Frangi's vessel detected image. They manually selected seed from desired region. They used region growing for its efficiency.

In paper [6], thresholding is applied to extract vessel from background. Authors have used Otsu's thresholding, because its complexity is low.

In paper [6], image segmentation is by Fuzzy C-Means (FCM)[1] algorithm. FCM computes cluster centres or centroids by minimizing the dissimilarity function with the help of iterative approach. By updating the cluster centres and the membership grades for individual pixel, FCM shifts the cluster centres to the "right" location within set of pixels.

III. EXPERIMENTED RESULT



Figure 1: Original Result

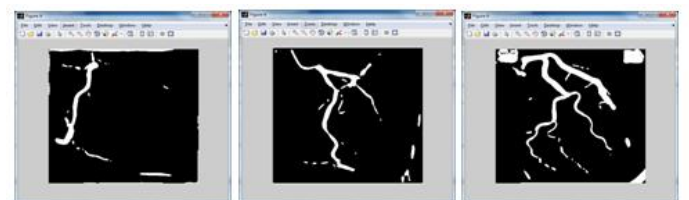


Figure 2: Result of paper [10]



Figure 3: Result of paper [3]

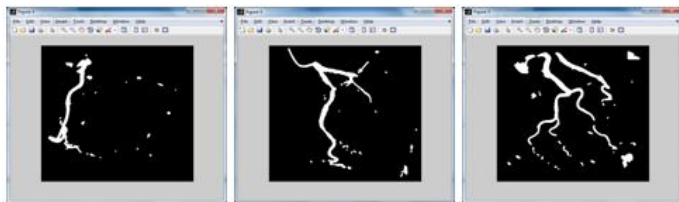


Figure 4: Result of paper [4]

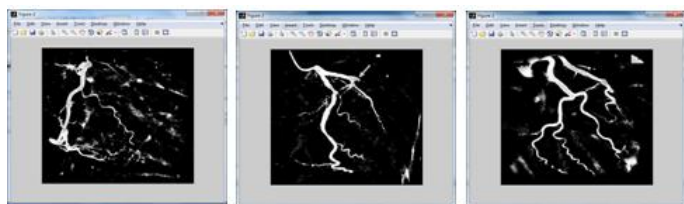


Figure 5: Result of paper [6]

Table 1: Result of Jaccard Coefficients of Various Segmentation Techniques

Image Sample	Paper[10] Jaccard coeff.	Paper[3] Jaccard coeff.	Paper[4] Jaccard coeff.	Paper[6] Jaccard coeff.
1	1	0.919	1	0.9208
2	0.9713	0.9073	0.991	0.9893
3	0.9826	0.8207	0.9911	0.9881
4	0.9987	0.9064	1	0.9986
5	0.9752	0.9192	0.9475	0.4125
6	0.9838	0.837	0.9983	0.662
7	0.9944	0.8313	0.9992	0.9986
8	0.9654	0.7826	0.993	0.9845
9	0.9605	0.8764	0.9816	0.9773
10	0.9479	0.8862	0.9578	0.8069
11	0.9995	0.8615	1	0.904
12	0.9755	0.7909	1	0.9729
13	0.9783	0.7601	0.9732	0.9743
14	0.9816	0.8473	0.979	0.52
15	0.975	0.9151	0.9928	0.9797
16	0.9727	0.9422	0.9896	0.9919
17	0.963	0.8749	0.9941	0.6194
18	0.9959	0.838	0.986	1
19	0.9913	0.7023	1	0.9953
20	0.9855	0.7472	0.9972	0.9896
21	0.991	0.8875	1	1

CONCLUSIONS

Jaccard similarity coefficient is the one among method to calculate segmentation performance. Jaccard similarity coefficient lies between value 0 and 1. A value of 0 indicates

that there is no similarity whereas a value of 1 indicates a similarity. For each image, jaccard coefficient is calculated. We know that, highest jaccard coefficient indicates better segmentation.

References

- [1] Akara Sopharak, Bunyarit Uyyanonvara, Sarah Barman, "Automatic Exudate Detection from Non-dilated Diabetic Retinopathy Retinal Images Using Fuzzy C-means Clustering" Sensors 2009
- [2] Eiho.S and Qian.Y, "Detection of Coronary artery tree using morphological operator", Computers in Cardiology, pp 525-528, 1997.
- [3] Geethu sasidharan and Angitha George, "Frangi's Vessel Detection Approach for Coronary Angiogram Segmentation", International Journal of Engineering Trends and Technology (IJETT) – Volume 13 Number 5 – Jul 2014
- [4] Harsha P. Jawale and .K.S.Bhagat" Blood vessel segmentation in coronary angiogram Image", International Journal of Engineering Development and Research, Volume 4, Issue 3, 2016
- [5] Jean Serra, "Image Analysis and Mathematical Morphology", Volume 1,1982;Volume2,1988, Academic Press.
- [6] Nilanjan Dey, 2Anamitra Bardhan Roy, et al, "FCM Based Blood Vessel Segmentation Method for Retinal Images", International Journal of Computer Science and Network (IJCSN), Volume 1, Issue 3, June 2012
- [7] Noise Adaptive Fuzzy Switching Median Filter for salt-and pepper noise reduction, Kenny Kal Vin Toh and Nor Ashidi Mat Isa, IEEE Signal Processing. Letters, vol. 17, pp. 281-284, March 2010.
- [8] P.J. Yim, P.L. Choyke, and R.M. Summers. Gray-scale skeletonization of small vessels in magnetic resonance angiography. IEEE Transactions on Medical Imaging, 19:568–576, 2000.
- [9] S. Eiho, H. Sekiguchi, N. Sugimoto, T. Hanakawa, and Urayama S. Branch-based region growing method for blood vessel segmentation. In Proceedings of Conference on International Society for Photogrammetry and Remote Sensing, pages 796–801, 2004.
- [10] S. Sathish Kumar and R. Amutha "Automated Segmentation of Angio Vessels" Advanced Computing and Communication Technologies for HPC Applications-ACCTHPCA, June 2012
- [11] T. Chanwimaluang and G. Fan. An efficient blood vessel detection algorithm for retinal images using local entropy thresholding. In Proc. of the IEEE Intl. Symp. on Circuits and Systems, 2003.
- [12] World Health Organization, "Cardiovascular diseases (cvds), fact sheet n317," September 2012.
- [13] Y Qian, S Eiho, N Sugimoto, M Fujita," Automatic Extraction of Coronary Artery Tree on Coronary Angiograms by Morphological Operators",IEEE conference on computers in cardiology 1998.