



Segment of Blood Vessels and Optic Disc in the Fundus Retinal Images

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ABSTRACT: In this paper, we present a segment of blood vessels and optic disc in the fundus retinal images. The two main steps in this proposed method are: Our method takes as first step the extraction of the retina vascular tree using the graph cut technique Markov Random Field (MRF) image reconstruction method segments the optic disc by eliminating vessels from the optic disc region and second step as Compensation Factor (CF) method segments the optic disc using prior local intensity knowledge of the vessels. Those alternative methods are show that our method achieved outstanding performance in segmenting the blood vessel and optic disc. After completing the segmentation using input image as retinal image then we have to find the retinal diseases such as diabetic retinopathy, glaucoma and hypertension.

KEYWORDS: Retinal images, vessel segmentation, optic disc segmentation, graph cut segmentation.

I. INTRODUCTION

The segmentation of fundus retinal image fundus features a nice interest as a result of it might be used as a non-intrusive identification in fashionable medicine. The morphological of the retinal vessel and also the optic disk is a very important structural indicator for evaluating the presence and extremity of retinal diseases like diabetic disease, high blood pressure, glaucoma, haemorrhages, vein occlusion and neo-vesselcularisation. though the valuate of diameter and curvity of retinal blood vessel or the form of the optic disk, human planimetric has unremarkably been employed by specialist, that results in an individual's error, if vessel fundus ar unclear or a large variety of images are non-inheritable to be examined by hand. Therefore, a real machine-controlled methodology for retinal vessel and optic disk segmentation, that causes numerous vessel and optic disk characteristics is enticing in pc aided-diagnosis.

An automated segmentation and examination of fundus retinal blood vessel features such as diameter, colour and curvity as well as the optic disk morphology allows ophthalmologist and oculist to perform mass vision screening exams for early detection of retinal diseases and treatment evaluation. This could stave off and decrease vision impairments; growth related diseases and many cardiovascular diseases as well as decreasing the cost of the screening.

In past years, many segmentation techniques has been utilized for the segmentation of retinal fundus like blood vessels and optic disk and diseases in fundus retinal images. but the acquisition of fundus retinal images underneath completely different conditions of lighting, resolution and Field Of Read (FOV) and the combination tissue in the tissue layer cause a vital humiliation to the presentation of machine-driven vessel and optic disk segmentations. Thus, there's a desire for a dependable technique for retinal tube tree enlargement and optic disk detection, that causes numerous vessel and optic disk shapes. within the following section, we tend to studied in previous section on blood vessel segmentation and optic disk segmentation severally.

II. RELATED WORK

There are two different approaches has been deployed to segment the blood vessels of the fundus retina: the pixel processing based methods and tracking based methods [2].The pixel processing based approach performs the vessel segmentation in a two-pass operation. First the appearance of the blood vessel is enlarge using detection process such as morphological pre-processing techniques and adaptive filtering. The second operation is the recognition of the

International Journal of Innovative Research in Computer and Communication Engineering

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 3, March 2016

blood vessel fundus using thinning or thickening the branching point operations to identify a pixel as a blood vessel. These approaches process every pixel in the image and apply multiple operations on each pixel. There are some pixel processing methods used for neural networks and frequency analysis to define pixels in the retinal image as well as blood vessel pixels and branching pixels. Typical pixel processing operations are shown in Hoover et al. [1], Mendoca et al. [5], Soares et al. [6], Staal et al. [3], Chaudhuri et al. [4] and Zana et al. [7].

The second set of approaches to blood vessel segmentation are referred to as vessel tracking, vectorial tracking [2]. In contrast to the pixel processing based approaches, the tracking methods to find first initial vessel branch points, and then track the rest of the vessel pixels through the image by measuring the continuity properties of the blood vessels. This technique is used as a single pass operation, where the detection of the vessel fundus and the recognition of the fundus are alternatively performed.

The tracking based approaches are included in semi automated tracking and automated tracking methods. In the semi automated tracking methods, the user manually selects the initial vessel branch point. Those methods are generally used in Quantitative Coronary Angiography (QCA) for a reduction in the coronary artery luminal cross-sectional diameter and generally those are segmentation of the blood vessels. In fully automated tracking, the algorithms automatically select the initial blood vessel points and most methods used as Gaussian functions to identify a vessel like model, which locates a vessel points for the vessel tracking. They are efficient and more suitable for fundus retinal image processing. Examples of the tracking based approaches are presented in Xu et al. [8], Maritiner-perez et al. [9], Staal et al. [3], Zhou et al. [10].

These techniques are performed using morphological operations to removing the blood vessels from the fundus retinal image. However, the application of morphological operations can define the image by degrading some useful information.

III. BLOOD VESSELS SEGMENTATION

Blood vessels may be seen as skinny elongated fundus within the tissue layer, with variation in breadth and length. In order to section the vessel from the anatomical fundus retinal image, we've enforced a pre-processing technique, that consists of effective reconciling bar graph feat (AHE) and strong distance remodel. This operation improves the strength and also the accuracy of the graph cut algorithmic rule. Fig.1 shows the illustration of the vessel segmentation algorithmic rule.

1. Vessel Segmentation:

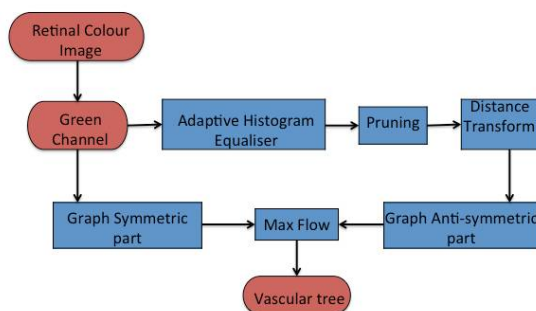


Fig .1 vessel segmentation

Detection of the papilla region and vessel detection on images of the tissue layer are issues that may be resolved with pattern recognition techniques. geography images, as provided e.g. by the hormone-replacement therapy device, yet as body fundus images is used as supply for the detection. it's of diagnostic importance to separate vessels within the

International Journal of Innovative Research in Computer and Communication Engineering

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 3, March 2016

papilla space from those outside this area. Therefore, detection of the papilla is vital additionally for vessel segmentation. during this contribution we tend to gift state of the art strategies for automatic disk segmentation and compare their results. Vessels detected with matched filters (wavelets, derivatives of the mathematician, etc.) are shown yet as vessel segmentation using image morphology. we tend to gift our own methodology for vessel segmentation supported a special matched filter followed by image morphology. Duringthis contribution we tend to argue for a brand new matched filter that's fitted to giant vessel in HRT images.

2. Optic Disc Segmentation:

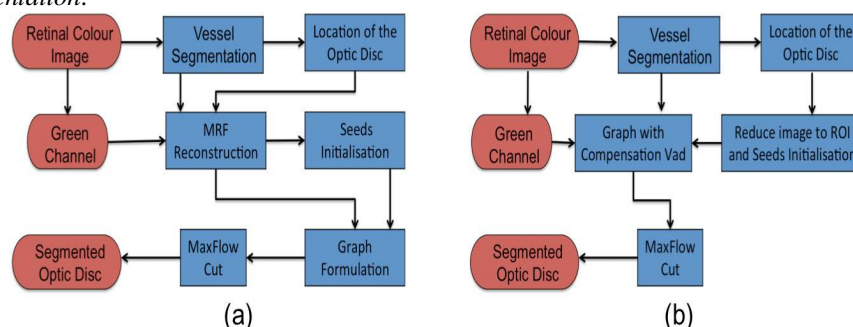


Fig .2 (a) Markov Random Field Image Reconstruction method diagram, (b) Compensation Factor method diagram

Detection of optic disk space is advanced because it's settled in a vicinity that's thought-about as pathological blood vessels once in segmentation and therefore require a technique to discover the realm of the optic disk, this paper projected the optic disk segmentation employing a method that has not been used before, and this technique is very easy, K-means clump may be a projected technique in this paper to discover the optic disk space with form reducing adaptative morphology. This paper with success detect optic disk space quickly and segmental blood vessels a lot of quickly.

IV. PROPOSED METHOD

In this paper ,two proposed methods are involved they are our technique takes as initiative the extraction of the membrane vascular tree mistreatment the graph cut technique. The vessel data is then accustomed estimate the placement of the optic disc. The optic disc segmentation is performed mistreatment two different methods.

The Markov Random Field (MRF) image reconstruction methodology segments the optic disc by removing vessels from the optic disc region.

The Compensation Factor (CF) technique segments the optic disc mistreatment previous native intensity information of the vessels.

V. EXPERIMENTAL RESULTS

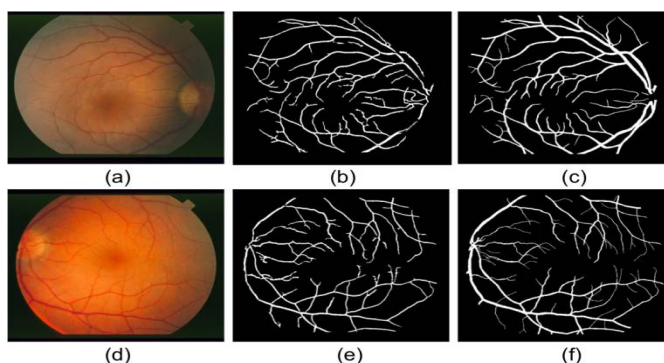


Fig. 3. The STARE dataset: a) and d) retinal images, b) and e) our segmentation results, and c) and e) manually labelled results.



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To facilitate the performance comparison between our method and different retinal blood vessels segmentation approaches, parameters like verity positive rate (TPR), the false positive rate (FPR) and also the accuracy rate (ACC) are derived to live the performance of the segmentation [5]. The accuracy rate is defined because the total of verity positives (pixels properly classified as vessel points) and the true negatives (non-vessel pixels properly identified as non vessel points), divided by the whole variety of pel in the pictures. True Positive Rate (TPR) is defined because the total variety of true positives, divided by the amount of blood vessel pel marked within the ground true image. False Positive Rate (FPR) is calculated because the total variety of false positives divided by the amount of pixels marked as non-vessel within the ground true image. it's price mentioning that an ideal segmentation would have a FPR of zero and a TPR of 1. Our technique and every one the choice ways used the first knowledgeable hand tagged pictures as performance reference.

VI. CONCLUSION AND FUTURE WORK

We have conferred a unique approach for blood vessels and optic disk segmentation in retinal images by integration the mechanism of flux, MRF image reconstruction and compensation issue into the graph cut technique. the method conjointly involves distinction sweetening, accommodative bar graph feat, binary gap and distance remodel for pre-processing. we've got evaluated the performance of vessel segmentation against ten different strategies together with human manual labeling on the STARE data set and fifteen different strategies together with human manual labelling on the DRIVE dataset. For the optic disk segmentation, we've got evaluated the performance of our technique against 3 different strategies on the DRIVE and DIARETDB1 datasets. Further a lot of, the projected technique addresses one among the most problems in medical image analysis, "the overlapping tissue segmentation". Since the blood vessels converse into the optic disk space and misguide the graph cut rule through a brief path, breaking the optic disk boundary. to attain a decent segmentation results, the MRF image reconstruction rule eliminates vessels within the optic disk space with none modification of the image fundus before segmenting the optic disk. On the opposite hand the compensation issue incorporates vessels victimisation native intensity characteristic to perform the optic disk segmentation. so our technique is applied indifferent medical image analysis applications to beat "the overlapping tissue segmentation." Our future analysis are going to be supported the segmentation of membranel diseases (lesions) referred to as "exudates" victimization the divided fundus of the retina (blood vessels and optic disc). Thus a background guide is created victimization these fundus. Then this guide is accustomed perform the detection of suspicious areas (lesions) within the retinal images.

ACKNOWLEDGMENT

The authors would really like to impart V. Kolmogorov for providing the computer code MaxFlow-v3.01 to cipher the graph cut.

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