Glaucoma detection of retinal images based on boundary segmentation

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ABSTRACT

The rapid growth of technology makes it possible to implement in immediate diagnosis for patients using image processing. By using morphological processing and adaptive thresholding method for segmentation of optic disc and optic cup, various sizes of retinal fundus images captured through fundus camera from online databases can be processed. This paper explains the use of color channel separation method for pre-processing to remove noise for better optic disc and optic cup segmentation. Noise removal will improve image quality and in return help to increase segmentation standard. Then, morphological processing and adaptive thresholding method is used to extract out optic disc and optic cup from fundus image. The proposed method is tested on two publicly available online databases: RIM-ONE and DRIONS-DB. On RIM-ONE database, the average PSNR value acquired is 0.01891 and MSE is 65.62625. Meanwhile, for DRIONS-DB database, the best PSNR is 64.0928 and the MSE is 0.02647. In conclusion, the proposed method can successfully filter out any unwanted noise in the image and are able to help clearer optic disc and optic cup segmentation to be performed.

Keywords:
Adaptive Thresholding
Glaucoma
Morphological
Segmentation

1. INTRODUCTION

Glaucoma is one of the factors for irreversible blindness in the world. Visual field loss, optic disc cupping, and in most instances an increased intraocular pressure are the main features of glaucoma [1]. It is originated from increasing pressure called intraocular pressure (IOP). The pressure happens because of the leftover fluid that is gradually accumulated in the front part of the eye. The liquid cause increasing pressure within the eyeball and eventually lead to optic nerve damage.

If immediate treatment was not delivered, it could lead to permanent eye damage and untreatable vision loss. Figure 1 shows glaucoma progression from normal vision until extreme glaucoma.
However, not all high eye pressure caused glaucoma. Sometimes glaucoma can also happen to normal eye pressure as well. This depends on the level of pressure employed on the particular optic nerve. This pressure was usually measured using Goldmann Applanation Tonometry. Goldmann Applanation Tonometry was appraised as the main standard for IOP measurement [3].

Generally, there are two common glaucoma types which are primary open angle glaucoma (POAG) and angle closure glaucoma (ACG). About two thirds of glaucoma cases are POAG while the rest are ACG cases [1, 4-7]. Commonly, some well-developed health facility will provide free screening to detect glaucoma in patients [8]. This screening will be conducted using a fundus camera and performed by approved professionals and medical practices.

For glaucoma infected eyes, it can be seen from the size of the optic cup and optic disc as illustrated in Figure 2. Optic disc is the path where ganglion cell axons exit and leave the eye. Optic cup is situated in front of the optic disc. It is shaped like a small crater. Conventionally, the size of optic cup is smaller than optic disc. Only in glaucoma infected eye, the optic cup size increases and grow larger. This is due to elevation of IOP. Besides, blood flow to the optic nerve is loss and nerve fibers slowly dying. The normal size for optic cup is 0.3 to 0.4 while glaucomatous eye have cup size of 0.7 to 0.9 [9].

The algorithm for this software that will be developed can perform the image processing techniques needed for the detection of glaucoma. There are two important steps completed. The steps are pre-processing phase and segmentation phase shown in Figure 3.
2. RESEARCH METHOD

The images used in this project were obtained from two online databases which are RIM-ONE [11] and DRIONS-DB [12]. RIM-ONE database consists of 169 Optic Nerve Head (ONH) images retrieved from 169 full fundus images of different subjects. Figure 4 displayed examples of images from RIM-ONE. Domain experts have categorized the images into five different subsets:

a. 11 images of Ocular hypertension (OHT).
b. 14 images of Deep glaucoma.
c. 118 images of Normal eye (non-glaucomatous).
d. 12 images of Early glaucoma.
e. 14 images of Moderate glaucoma.

Meanwhile the next database which is DRIONS-DB demonstrates the segmentation of the optic nerve head in retinal fundus images. It was gathered by 2 medical experts. It has 2 ground truths for each set of 110 colour digital retinal images. Figure 5 illustrate the examples of images from this database.

![Image](i) (ii) (iii) (iv) (v)

Figure 4. The examples of images from RIM-ONE database for (i) Ocular hypertension (OHT) (ii) Deep glaucoma (iii) Normal eye (non-glaucomatous) (iv) Early glaucoma and (v) Moderate glaucoma

![Image](i) (ii) (iii) (iv) (v)

Figure 5. Samples of images from DRIONS-DB database

2.1. Pre-Processing

Pre-processing image is a common early method to obtain improved and enhanced image and to draw out useful information from it. Pre-processing usually involve images with the lowest level of abstraction. The aim of pre-processing is to remove noise, correct some of the degradation in image, suppresses unwanted distortions and amplifies image features important for further processing.

For pre-processing, the method used is color channel separation method [13]. The fundus images obtained from databases are split into three color channels. Normally used color channels are red, green, and blue color channels (RGB). The RGB components usually regarded as three sub-sources of single color source [14]. 24-bit RGB image can give storage pixels with conventional brightness intensities of 0 until 255 [15]. It first started with black color and the light emission increase as correspond to added colors. Figure 6 shows examples of images obtained as the outcome of color channel separation method.

One of the color channels is chosen and then processed further into grayscale. For this system, red color channel selected and undergoes the next process.
2.2. Segmentation Phase

The next process are segmenting and extracting important features from the grayscale image obtained from pre-processing. Segmentation phase will separate an image into pixels with similar attributes. The pixels then accumulated into its distinguishable regions. This distinguishable regions need to have large relation to characteristic of interest for an accurate image interpretation. Typically for glaucoma, the important features that need to be extracted are blood vessels and optic disc.

In this phase, the obtained grayscale image is converted into binary disc image. Binary disc image demonstrate raw disk images after conversion from grayscale image.

After that, a further processed detailed image almost similar to binary disc image acquired using Morphological Processing method [16]. This image acquired is more refined that previous binary disc image existed as it improves and enhance image so that further processing can be made with ease. Generally, there are two formulas used in Morphological Processing method [16]. Each one represents dilation operation and erosion operation. Dilation operation is a type of operation that increase object in a binary image whether in terms of size or thickness meanwhile erosion operation is the operation that does the opposite of dilation method for example the operation reduce object’s size or thickness in binary image. These two operations are very important in increasing the accuracy of Morphological Processing method. It is expressed as:

\[ X \oplus B = X + b = \{ x + b : x \in X, b \in B \} \] (1)

Where \( X \) is any grayscale shape and \( B \) is symmetric structuring element.

\[ X \ominus B = X - b = \{ z : (B + z) \subseteq X \} \] (2)

Symmetric structuring element is contained inside input set \( X \) as output of erosion is also translated as set of translation points. Figure 7 shows the example of dilation operation while Figure 8 represents erosion operation.

![Figure 7. Dilation operation (i) before operation (ii) after operation](image)
Furthermore, the segmented disk image showing optic disc using Adaptive Thresholding method is secured. By using this method, optic disc can be carefully and clearly be extracted out. Optic disc is the most important characteristic in diagnosing glaucoma [17]. This owe to the fact that glaucoma is very hard to be roughly detected with only our eyes. It needs proper eye examination with suitable machine. Optic disc can show one of the obvious sign in diagnosing glaucoma.

Lastly, the segmentation was made to identify optic cup as well. Optic cup is also very important characteristics in glaucoma detection. Its size can determine whether an eye is healthy or glaucomatous. From images obtained, it is proven that the optic cup of glaucomatous eye is bigger than optic cup of normal eye [18-20]. All the images for each process can be viewed in Table 1.

Circular Hough Transform (CHT) method was commonly implemented as segmentation of optic disc [21]. By comparing to the proposed method, it can be seen that Adaptive Thresholding and Morphology method is more accurate as it also extract out optic cup and optic disc of retina fundus image. The CHT method only capable to show and illustrates countless circles as in shown in Figure 9 (i) while in Figure 9 (ii), one circle to detect optic disc was picked as maximum energy in the vicinity.

<table>
<thead>
<tr>
<th>Table 1. Images Obtained after Each Process</th>
</tr>
</thead>
<tbody>
<tr>
<td>RIM-ONE</td>
</tr>
<tr>
<td>Original image</td>
</tr>
<tr>
<td>Grayscale image</td>
</tr>
<tr>
<td>Binary disc image</td>
</tr>
<tr>
<td>Segmented disc image</td>
</tr>
<tr>
<td>Binary cup image</td>
</tr>
<tr>
<td>Segmented cup image</td>
</tr>
</tbody>
</table>

the best approximate circle (maximum energy in the vicinity) [21]
2.3. Performance Metrics

There are two parameters calculated for performance metrics for pre-processing. The Peak Signal to Noise Ratio (PSNR) and the Mean Square Error (MSE) are the two error metrics used to compare image compression quality. The calculation for PSNR and MSE is completed to compare the accuracy of the pre-processing method used [22-23].

Peak signal-to-noise ratio (PSNR) shows the ratio of power limit to the power of same image with noise. The PSNR calculate the peak signal-to-noise ratio between two images. It is normally regarded in logarithmic decibel scale (dB). However, it also can be calculated in percent (%). MAXI (maximum possible pixel value) of the image used is 255 because the image is uint8 [24].

As shown in (3) below is the formula for calculating PSNR and (4) is the formula to calculate MSE.

\[
\text{PSNR} = 10 \times \log_{10} \left( \frac{\text{MAXI}^2}{\text{MSE}} \right) \quad (3)
\]

\[
\text{MSE} = \frac{1}{mn} \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} \left( \text{Image}(i,j) - \text{NoisyImage}(i,j) \right)^2 \quad (4)
\]

3. RESULTS AND ANALYSIS

It can be concluded that for RIM-ONE database, the PSNR is 0.01891 and MSE is 65.62625. Meanwhile, for DRIONS-DB database, the PSNR is 64.0928 and the MSE is 0.02647. The results obtained clearly stated that the PSNR value obtained from after pre-processing is higher than the original image. The higher the PSNR value, the higher the level of audio signal than level of noise [25]. So, it is proven that the image quality after pre-processing is better than original image as noise was removed [26]. Table 2 indicates the PSNR value of original image before and after pre-processing was made.

<table>
<thead>
<tr>
<th>No.</th>
<th>Image sample from database</th>
<th>MSE value (%)</th>
<th>PSNR value (%)</th>
<th>Image sample from database</th>
<th>MSE value (%)</th>
<th>PSNR value (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>N</td>
<td>Im133</td>
<td>0.0158</td>
<td>66.1523</td>
<td>image_001</td>
<td>0.0366</td>
<td>62.4979</td>
</tr>
<tr>
<td>m</td>
<td>Im152</td>
<td>0.0196</td>
<td>65.2033</td>
<td>image_002</td>
<td>0.0314</td>
<td>63.1555</td>
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<tr>
<td>a</td>
<td>Im160</td>
<td>0.0197</td>
<td>65.1949</td>
<td>image_003</td>
<td>0.0348</td>
<td>62.7135</td>
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<tr>
<td>l</td>
<td>Im161</td>
<td>0.0246</td>
<td>64.2238</td>
<td>image_031</td>
<td>0.0204</td>
<td>65.0375</td>
</tr>
<tr>
<td>G</td>
<td>Im169</td>
<td>0.0278</td>
<td>63.6834</td>
<td>image_062</td>
<td>0.0113</td>
<td>67.6144</td>
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<td>0.0103</td>
<td>68.0137</td>
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<tr>
<td>u</td>
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<td>67.3244</td>
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<td>0.0275</td>
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<tr>
<td>c</td>
<td>Im038</td>
<td>0.0151</td>
<td>66.3510</td>
<td>image_024</td>
<td>0.0239</td>
<td>64.3449</td>
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<tr>
<td>o</td>
<td>Im103</td>
<td>0.0309</td>
<td>63.2242</td>
<td>image_047</td>
<td>0.0246</td>
<td>64.2258</td>
</tr>
<tr>
<td>m</td>
<td>Im120</td>
<td>0.0133</td>
<td>66.8915</td>
<td>image_092</td>
<td>0.0259</td>
<td>63.9897</td>
</tr>
<tr>
<td>a</td>
<td></td>
<td>Average (%)</td>
<td>0.01891</td>
<td>65.62625</td>
<td>0.02647</td>
<td>64.0928</td>
</tr>
</tbody>
</table>

4. CONCLUSION

As conclusion, the segmentation and detection of glaucoma is successfully completed. The methods used for preprocessing and segmentation works perfectly and are able to give satisfactory result. For future
work, the next process must be conducted. For instances, next process should involve calculating the size of optic cup and optic disc then proceed to classification into two common different types of glaucoma. Besides, a MATLAB GUI can also be developed in the future for a more efficient and accurate glaucoma detection system.

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Glaucoma detection of retinal images based on boundary segmentation (Norainia Alia Zainudin)


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