An Automated Solution for Extracting and Counting of White Blood Cells in a Blood Smear Images

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ABSTRACT

There are various tools present commercially for automatic counting of Blood Cells. These tools are used for counting and finding the different types of Blood Cells present in the Blood Smear. The count of White Blood Cells is important for detecting various diseases as well as to follow the accurate treatment for the diseases like Anemia, Leukemia, Inflammation, Systemic illness, Allergy and Burn-Induced etc. The White Blood Cell count gives the vital information about the disease's which is used for diagnosing the patients. The old conventional method of counting White Blood Cells under microscope gives an inaccurate and unreliable results depending upon how the laboratory technician works on it. Another method for White Blood Cell counting is done through cell counter machine, which is very expensive and cannot be afforded to remote/rural areas, hence to overcome such problems this paper proposes a cost-effective approach, i.e, automated solution using smart phone based solution for extracting and counting of White Blood Cells.

KEYWORDS: Blood Smear Image, Contour, OPENCV, Smart Phone, White Blood Cell

I. INTRODUCTION

The analysis of microscopic images is acutely important in both the medical and the computer/information science fields. Many research issues are related to the analysis of microscopic images, such as Complete Blood Cell (CBC) tests and the analysis of blood smears, which is considered the first step in detecting and diagnosing Anemia, Leukemia, and Malaria. Also, during a complete physical exam a series of tests are performed. One of these tests is the CBC, which is in used to evaluate the composition and concentration of the cellular blood components. The CBC determines red blood cells (RBC) counts, white blood cell (WBC) counts, platelet counts, hemoglobin (HB) measurements, and mean red blood cell volume. In this paper, we have focused on White Blood Cell as these cells provide major defense system against infections. The normal WBC count ranges from 4,500 to 11,000 WBC's per cubic millimeter of blood. Higher WBC counts (above 30000 cells per cubic millimeter) indicate an infection, systemic illness, inflammation, leukemia, or burninduced tissue injury. However, manual or visual quantification of parasitemia in the thin blood films and WBCs in leukemia is an extremely tiresome, time-consuming, and subjective task with a high probability of counting error. An accurate segmenting mechanism and counting mechanism that gathers information about the distribution of microscopic particles may help diagnose abnormalities during clinical analysis.

WBCs count is very important in the diagnosis of several diseases, hence this paper has acutely worked on the issues of extracting and counting of White Blood Cells in the images captured from the Electronic microscope.

The proposed algorithm separates WBCs from other blood cells in the image and counts the total number of WBCs present in the images. The proposed work has been done using OPENCV 2.4.10 library and ANDROID 4.2.2 (Jellybean) platform. Finally the count is obtained in per cubic millimeter (cu mm), which is standard practice of a medical practitioner.

II. LITERATURE REVIEW

In [1] provide the very fast, easy and non-expensive segmentation process through thresholding. With the aid of thresholding methods differentiate the grey level in the image from the background level and also the images are do not touch each other, it is one of the significant role of the segmentation by thresholding. One more image segmentation method for white blood cells was implemented in [2], but this also based upon thresholding but that thresholding is depending upon several gray level thresholding. Filters are implemented in this method. Low pass filters are used in this method with window size of 5 x 5. In [3] performed comparisons between nine image segmentation which is gray level thresholding pattern matching, filtering operators, morphological operators, edge detection operators, gradient-in method, RGB color thresholding, color matching and HSL (hue, saturation, lightness) and color thresholding techniques on RBC and concluded that there is no single method can be considered good for RBC segmentation [4]. These are two difficult issues in image segmentation where common segmentation algorithms cannot solve this problem [5]. Besides that, there are staining and illumination inconsistencies also acts as uncertainty to the image Sabino and Costa [7] used the Green channel of the RGB model to segment WBC. On the other hand, Westpfalz applied HSI color model to separate WBC from background and de-cluster the clustered WBCYang, Foran and Meer [9] improved the algorithm of IGDS to better segment WBC from other BCs presented in the smear image. Sinha and Ramakrishnan [10] used color used HIS equivalent of the WBC image, K-Means clustering followed by EMalgorithm to segment WBC along with the cytoplasm nucleus. In [11] used fuzzy patch labeling to segment WBC from other blood elements. The main drawback of the above systems is that they have not concentrated on doing segmentation process in the Android Smart phone. Hence those systems cannot be afforded to remote/rural areas. The proposed system has focused on this aspect. The proposed system will run on the Android Smart phone and it has used HSV color model in order to segment the blood smear images collected from the different Blood samples

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III. METHODOLOGY

This paper introduces an automatic WBC count using computer vision which helps to perform the counts accurately using image based analysis from which the blood smear taken by the electronic microscopic. Steps involved in the process of estimating the White Blood Cells include: Image Acquisition, Image Preprocessing, Image Segmentation, Image Post processing and WBC counting. **Figure 1** shows the overall flow of the proposed system.



Figure.3, shows the original blood smear microscopic image



Figure.3: Original blood smear microscopic image

B. Image Pre-Processing

HSV image.

The input for this step is a blood smear image and this image is enhanced and filtered to get a good quality image for the next steps. The higher resolution image is decoded using Bitmap decode in order to reduce the resolution of the image.

Further the image is converted to HSV image, which clearly defines the bright objects such as WBC. The pre-processed image is sent for Image Segmentation. Figure. 4, shows the

A. Image Acquisition

This is the first method which is performed to capture the lopment image from the prepared blood smear image using Electronic microscope. The blood smear slide is prepared in 2456-647 content://media/external/images/media/34149 the form of tongue shape, and then the images are captured from the feather edge of the blood smear as shown in *Figure2.* The images are captured from the slides by an Olympus BX50 microscope equipped with Moti camera.

For the proposed method the blood smear images of the two patients are considered, for each patient a single slide is prepared from which ten sample images are captured at different angles. These images are stored in JPEG format. The dataset used in this project consists of actual microscopic images of blood samples, captured in the Laboratories.



Figure 2: Blood smear in tongue shape

The images were taken by placing the prepared blood smear under the electronic microscope at a magnification factor of 40x. The image thus captured is provided as an input for the preprocessing stage.



Figure.4: HSV image.

C. Image Segmentation

The input for this step is the pre-processed image. Here two threshold values have been considered namely lower and upper threshold values. Where the lower threshold value is set to (120,100,100) and the upper threshold value is set to

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(179,255,255), in order to segment the image. i.e, these 🗊 SMB threshold values are applied to extract the saturation component say S, because this S in the image show clearly Click Me the White blood cells. Figure 5, shows the segmented image. T SMB content://media/external/images/media/34149 Click Me content://media/external/images/media/34149 1 ŵ Figure.7: Extraction of contours Figure.5: Segmented image SMB

D. Image Post-Processing

After segmentation the noise present in the segmented Click Me image will be removed using Gaussian blur method of kernel nal Journ content://media/external/images/media/34149 size 17 x 17. A Contour can be explained simply as a curve joining continuous points, having same color or intensity. in Scientifi The proposed method extracts the contours of the cells in arch and the segmented image and then the area of the each contour will be found. From the experimental results we found that if the contour area lies between 5,400 to 6,000 pixels, then those contours are considered as WBCs contours. For highlighting WBCs area draw contour method is used. In this case, we obtain three different images for each method. Figure 6, shows the image after applying Guassian blur. Figure 7, shows the image after extracting the contours. Figure 8, shows the image after applying Draw contour.



Figure.6: After applying Guassian Blur





Figure.8: After applying draw contour

E. Counting

After the application of Draw contour method only the WBCs cells will be highlighted, which will be used for counting purposes. However, blood count in medical terms means the number of blood cells in a cubic millimeter of blood volume. Hence we have deduced a formula (1) to calculate the number of white blood cells per cubic millimeter, based on the number of cells in the area of the given image of the blood sample. We have assumed that the thickness of the blood sample film is 0.1 mm, which is the normal medical practice. This allows for an overlapping of maximum two layers in thickness, which is the common trend in the images provided. This formula requires number of white blood cells

[8]

obtained from each blood smear image, where the number of blood smear images per patient is considered to be ten.

There is a standard value which is suggested by the pathology technician, which is a constant value to be multiplied with the result obtained from dividing the number of White blood cells obtained with the number of input images per patient i.e., ten. The overall result is measured in cubic millimeter (cu mm). Considering these factors the formula for the WBCs count becomes:

$$N = \left[\frac{C}{10} \times 2000\right]_{(1)}$$

Where,

N = the actual WBC count per Cu mm. C = WBC count obtained from images.

RESULTS AND DISCUSSION IV.

Here in this paper we have considered five patient's blood smear images for our study. Each of these images is preprocessed by the above mentioned techniques. Finally the number of White blood cells was extracted and counted in each of these images bases on the contours area. The results of our study have been discussed in Table - 1.

Table1. Comparison of the results between the	e proposed C
method and the Manual method	no

Patient ID	Number of Images Taken per patient	Blood count Manually per Cu mm (in millions)	Blood count by proposed method per Cu mm (in millions)	Sf iona d in a
1	10	4990	4800	calu
2	10	5020	4780 Dev	elop
3	10	4870	4950	245
4	10	4980 🔨	4875	243
5	10	5180 🗸	5030	

V. CONCLUSION

This paper presents a methodology to achieve an automated extraction and counting of White blood cells in microscopic images using Contours. This technique is affordable to remote/rural area, independent of specialist to generate patient WBC report. Results indicate that the proposed method for White Blood Cell counting in blood smear images offers a remarkable accuracy of 96.6% by restricting the input images to ten in number per patient. The proposed method is also very cost-effective and can be easily implemented in medical facilities anywhere with minimal investment in infrastructure. It is also very time effective as manual counting is a very tedious job and time consuming.

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