INTRODUCTION

Analysis of microscopic images of blood smears for parasite-infected red blood cells can assist with screening and monitoring malarial infection. Millions of blood slides are manually screened for parasites every year, which is a time-consuming and subjective process. We have developed a smartphone-based software to perform this task using machine learning and image analysis methods for counting infected red blood cells automatically. The software runs on a standard Android smartphone attached to a microscope by a low-cost adapter. Images of thin blood smear slides are obtained through the eyepiece of the microscope using the smartphone’s built-in camera. The method we have implemented first needs to detect and segment red blood cells. However, the presence of white blood cells (WBCs) is adversely affecting the accuracy of red blood cell detection and segmentation since WBCs are often mistaken for red blood cells by current automatic cell detection methods. As a result, a pre-processing step for WBC elimination is necessary. Segmentation of WBCs is a complex process due to the morphological diversity of WBCs and staining differences. We propose a novel method for white blood cell segmentation in microscopic images of blood smears that combines a WBC detection method based on range filtering with a customized level-set algorithm to estimate the boundary of each WBC in an image.

WHITE BLOOD CELLS

- Also known as “Leukocytes”
- Nucleated cells in the blood that protect the body against infectious disease
- Five main types:
  - Neutrophils
  - Eosinophils
  - Basophils
  - Lymphocytes
  - Monocytes
- Segmentation challenges in blood smear images
  - Cell staining differences
  - Uneven illumination
  - Cell texture complexity
  - Shape diversities
  - Size variation

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METHOD

The proposed algorithm includes three main steps:
- Pre-processing
  - To extract region of interest (ROI) and perform illumination correction and color enhancement
- White Blood Cell Detection
  - To specify the location of any white blood cell present in an image using a range filtered version of the image
- White Blood Cell Segmentation
  - To segment the detected cells employing an accurate level-set algorithm

DATASET & ANNOTATION

- More than 1300 slide images containing about 1350 WBCs were acquired at Chittagong Medical College Hospital in Bangladesh
- Annotated manually by experienced professional slide reader using Firefly annotation tool (firefly.cs.missouri.edu)

RESULTS

- We evaluate the two processing steps of our algorithm, cell detection and cell segmentation, separately:
  - WBC Detection Step
    - Measure | Value
    - Precision | 96.37
    - Recall | 98.37
    - F1 Score | 97.36
  - WBC Segmentation Step
    - Measure | Value
    - Jaccard Index | 82.28
    - Dice Index | 78.33
- Illustrates, the output segmentation mask and ground truth mask are compared for three sample images. The green pixels represent the pixels falsely included in the segmented area (false positives) and the pink pixels show those falsely excluded from the segmented area (false negatives).

CONCLUSIONS

We introduced an algorithm that successfully detects white blood cells in microscopic images of thin blood smears following a range filtering approach. This method is capable of detecting different types of white blood cells with different shades of staining. The proposed algorithm accurately estimates the boundary of each cell utilizing a level-set algorithm. The results demonstrate that our algorithm provides promising segmentation outcomes. In future work, we will add this algorithm to our current smartphone application for malaria cell counting.

REFERENCES


White Blood Cell Detection and Segmentation in Microscopy Images of Thin Blood Smears

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