

Automated Detection of Hematological Disorders Using Blood Cell Images

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Abstract: *The project's represents it can automatically detect and analyze blood cells in microscopic photographs. Accurate identification of white blood cells (WBCs), red blood cells (RBCs), and platelets is critical for detecting a wide range of blood illnesses, including anemia, leukemia, and thrombocytopenia. Traditional blood analysis methods rely on manual observation, which is time-consuming and subject to human error. This project uses image processing and deep learning to automate the detection and segmentation of blood cells, delivering a quick, dependable, and efficient solution. The system leverages powerful Convolutional Neural Networks (CNNs) that have been trained on massive annotated blood picture datasets. The CNN model is intended to detect small changes in cell form, size, and structure, allowing for precise categorization and counting of WBCs, RBCs, and Platelets. Individual cells are isolated from the background using segmentation techniques, which improves detection accuracy and allows for detailed abnormality investigation. This automated technique not only saves time but also lowers the risk of misdiagnosis, allowing medical practitioners to make more informed judgments. The Python-based online application offers a user-friendly interface through which users may submit blood sample photos and acquire detailed data on cell counts and probable hematological illnesses. The system is intended to be scalable, dependable, and easily incorporated into clinical workflows, making it appropriate for hospitals, diagnostic labs, and research organizations. By merging deep learning with image processing, this study provides an efficient and practical solution for early diagnosis and monitoring of blood-related disorders, eventually contributing to better patient care and speedier diagnostic processes.*

Keywords: Blood-cell segmentation, Convolutional Neural Network (CNN), WBC detection, RBC detection, Platelet detection, Hematological disorder diagnosis, Image processing, Python web application, etc

I. INTRODUCTION

Hematological disorders, such as anemia, leukemia, and thrombocytopenia, are serious health issues that require timely diagnosis. Traditional methods for analyzing blood samples are time-consuming and often rely on the expertise of lab technicians, which may lead to human errors. With the growth of artificial intelligence and image processing techniques, it has become possible to automate blood-cell detection for faster and more accurate diagnosis.

This project aims to design a web application that takes microscopic blood images as input and automatically detects different types of blood cells. By applying image segmentation and training a model with Convolutional Neural Networks (CNNs), the system can identify WBCs, RBCs, and Platelets efficiently. The segmentation process isolates



each cell type, allowing precise counting and assessment of abnormalities, which can indicate the presence of specific hematological disorders.

The Python-based web platform provides an interactive and user-friendly interface for medical professionals and researchers. It can serve as a reliable diagnostic tool, enabling quicker analysis of blood samples and aiding in the early detection of diseases. Implementing such an intelligent system not only improves diagnostic accuracy but also supports large-scale medical screenings with minimal human intervention.

II. PROBLEM STATEMENT

Detecting and diagnosing blood-related disorders requires accurate identification of blood cells like RBCs, WBCs, and platelets. Manual methods are time-consuming, prone to human error, and need expert supervision. There is a lack of easily accessible, automated tools for quick analysis. Hospitals and labs face challenges in early diagnosis due to limited tools. Therefore, there's a need for a web-based, automated solution. The system should accurately segment blood cells and help in early detection. An efficient and intelligent system can improve patient care and save time.

III. LITERATURE SURVEY

- J. Alsamri, H. Alqahtani, A. M. Al-Sharafi et al. proposed a computer-aided diagnosis system for hematologic disorder detection based on spatial feature learning networks using microscopic blood cell images. Their study, published in Scientific Reports, focused on leveraging deep spatial feature extraction to enhance the detection accuracy of abnormal blood cells. The proposed model effectively captured both local and global spatial dependencies among blood cell features, improving classification precision for various hematologic conditions. Experimental results demonstrated superior accuracy compared to conventional CNN-based models, highlighting the potential of spatial feature learning for reliable, automated hematologic diagnosis.
- S. Choudhary, S. Kumar, and P. S. Siddharth et al. conducted a comprehensive performance evaluation of state-of-the-art deep learning architectures for blood cell detection and classification in their work published in BMC Medical Informatics and Decision Making. The study compared several CNN variants, including DenseNet, EfficientNet, and ResNet, across different blood cell datasets. The results indicated that modern architectures, particularly EfficientNet, achieved higher accuracy and faster convergence, making them suitable for medical image analysis. The research provided valuable insights into model selection and optimization strategies for automated hematologic analysis.
- X. Wang, G. Pan, and Z. Hu et al. introduced a two-stage algorithm combining an improved YOLOv7 detector and EfficientNetv2 classifier for accurate blood cell detection and classification, as reported in Scientific Reports. The proposed hybrid system first localized blood cells using an optimized YOLOv7 model and then classified them using EfficientNetv2 to achieve robust performance under varying image conditions. The method addressed challenges such as overlapping cells and uneven illumination, outperforming existing single-stage models in terms of precision and recall. Their approach demonstrates the effectiveness of integrating detection and classification networks for biomedical imaging.
- In their paper published in PLOS ONE, Z. Jan, M. Shabir, H. Farman, A. Rahman, and M. Nasrallah proposed a deep learning-based semantic segmentation model to identify leukemia-affected white blood cells. Their approach utilized a U-Net-inspired architecture to segment and analyze morphological changes in WBCs for early leukemia detection. The system demonstrated remarkable segmentation accuracy, particularly in differentiating healthy and abnormal leukocytes. The study emphasized the significance of semantic segmentation in medical diagnostics, offering a scalable solution for detecting hematological malignancies like leukemia.
- H. Sazak and colleagues developed an automated blood cell detection and classification framework using the YOLOv10 object detection algorithm. Their preprint work demonstrated that YOLOv10 could efficiently identify and categorize RBCs, WBCs, and Platelets from microscopic images with minimal preprocessing. The model achieved faster inference speed and higher detection accuracy than earlier YOLO variants, making it suitable for real-time



laboratory applications. The study showcased the growing relevance of cutting-edge deep learning detectors in clinical diagnostic automation.

IV. SYSTEM OVERVIEW

The proposed system is designed to automate the process of identifying and classifying blood cells from microscopic images using advanced deep learning techniques. The system employs a Convolutional Neural Network (CNN)-based architecture trained on a labeled image dataset containing samples of White Blood Cells (WBCs), Red Blood Cells (RBCs), and Platelets. The goal of this system is to enhance diagnostic accuracy, reduce human dependency, and speed up the detection of hematological abnormalities that may indicate diseases such as anemia, leukemia, or thrombocytopenia. The system's workflow involves image acquisition, preprocessing, segmentation, feature extraction, classification, and result visualization through a user-friendly web interface. This ensures that medical professionals can easily upload blood sample images and receive accurate analysis results within seconds.

The backend of the system is implemented using Python, leveraging powerful libraries such as OpenCV for image processing, TensorFlow/Keras for deep learning, and NumPy/Pandas for efficient data handling. The web-based front-end interface is developed using Flask, enabling smooth communication between the user and the model. The CNN model is trained on an image dataset that undergoes preprocessing techniques such as normalization, noise reduction, and contrast enhancement to improve segmentation quality. Once processed, the system classifies the cells based on learned patterns and displays the detected cell types along with segmentation boundaries. The overall system architecture emphasizes scalability, modularity, and high performance, making it adaptable to new datasets and capable of real-time or near-real-time analysis. This intelligent system not only reduces diagnostic workload but also contributes to more accurate and consistent results, thus offering a valuable tool for hematological research and medical diagnosis.

V. PROPOSED SYSTEM

The proposed system aims to design and develop a Python-based intelligent web application for automatic blood cell segmentation and classification to assist in the detection of hematological disorders. The system uses an advanced Convolutional Neural Network (CNN) model trained on a large dataset of microscopic blood images. The CNN architecture enables the system to automatically extract and learn deep visual features from the input images, distinguishing between different types of blood cells such as White Blood Cells (WBCs), Red Blood Cells (RBCs), and Platelets with high accuracy. By automating the cell detection and classification process, the system minimizes manual intervention, reduces diagnostic time, and ensures consistent and reliable results that can aid pathologists in making quicker clinical decisions.

The proposed system follows a structured workflow that includes several important stages — image acquisition, preprocessing, segmentation, feature extraction, and classification. Initially, the microscopic blood smear images are collected and preprocessed using techniques such as noise removal, contrast enhancement, and image normalization to improve image clarity. Then, the CNN-based segmentation module identifies and separates different types of cells by learning distinct features from the training dataset. Once the cells are segmented, the classification module determines the type of each cell and provides statistical insights into their count and distribution, which helps in identifying possible abnormalities linked to specific hematological disorders.



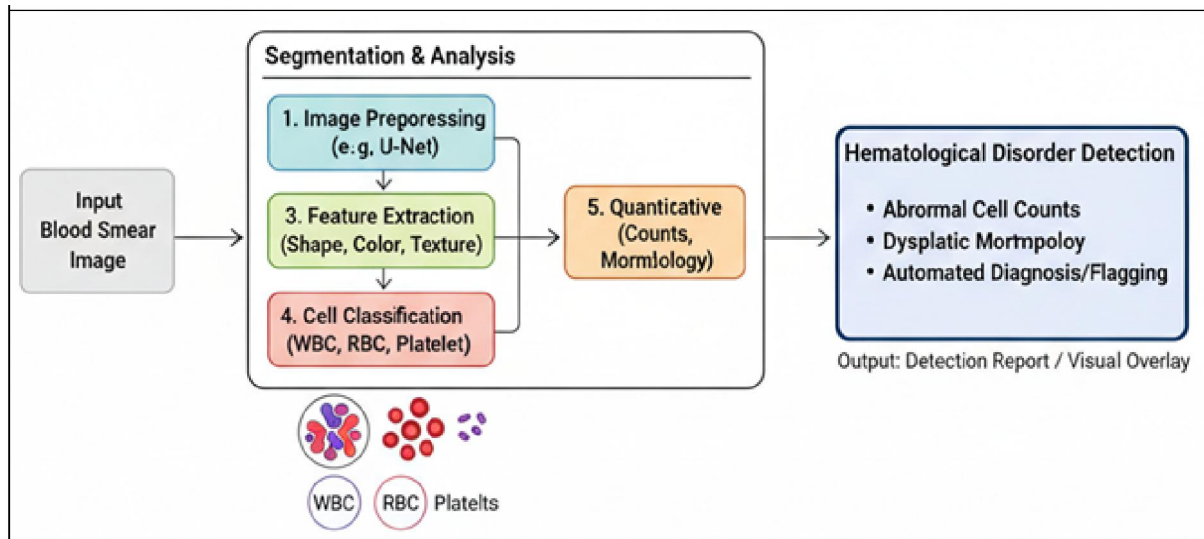


Fig. 1: System Architecture Design

Furthermore, the entire model is integrated into a user-friendly web interface built using Flask. This allows users, such as lab technicians or healthcare professionals, to upload blood sample images and instantly receive analyzed results with visual segmentation output. The proposed system is designed for efficiency, scalability, and ease of use, making it suitable for both research and clinical applications. By combining the power of deep learning with accessible web-based technology, this system represents a significant step toward intelligent and automated blood diagnostics that can improve early disease detection and enhance overall healthcare outcomes.

VI. CONCLUSION

The project successfully demonstrates how modern image processing and deep learning technologies can automate and enhance the accuracy of medical diagnostics. The system was designed to automatically detect, segment, and classify blood cells—specifically White Blood Cells (WBCs), Red Blood Cells (RBCs), and Platelets—from microscopic images using an advanced Convolutional Neural Network (CNN). By automating this process, the system addresses the limitations of traditional manual analysis, which is often time-consuming, labor-intensive, and susceptible to human error. The integration of a Python-based web interface further ensures usability and accessibility, allowing healthcare professionals to analyze blood samples efficiently and with minimal technical expertise. The use of preprocessing and post-processing techniques enhanced image clarity and boundary refinement, while the web application facilitated smooth user interaction and visualization of diagnostic results. These combined innovations make the proposed system an efficient, scalable, and practical tool for supporting hematological diagnosis and research.

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