



Acute Myeloid Leukemia Detection in Blood Microscopic Image by Using PNN

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ABSTRACT: Acute Myeloid Leukemia is a type of cancer that affects the blood cells growth and bone marrow where WBCs (leukocytes) born. This trade name of contagion extend shortly if it is moan diagnosed in beginning age. There are several types of insightful leukemia: sever Acute Lymphoblastic Leukemia (ALL) and piercing Acute Myelogenous Leukemia (AML). In this formula we analyse AML. AML is also called as Acute Myeloblastic Leukemia, Acute Granulocytic Leukemia and Acute Nonlymphocytic Leukemia. Dangerous Acute Myelogenous Leukemia (AML) is solid in loosely adults and divergent in children. There is primary need for automation of leukemia detection because it affect in the body organ. In this paper, a simple technique that automatically detects and segments nucleus from WBCs in blood smears and diagnose the cancer. The propositional make advances deem: the minute core bod is RGB looks which vary into gray height apropos clustering smoke K-means algorithm, texture and shape based feature extraction and classification by using PNN classifier.

KEYWORDS: Acute Myelogenous Leukemia (AML) web image, preprocessing, segmentation, feature extraction, classification.

I. INTRODUCTION

White blood cells (WBCs) or leukocytes plays role in the not only diagnosis of different diseases but also fight the infections i.e. acts as soldier as a result, extracting information about them is valuable for hematologists. It refers to the cancer of the blood or the bone pit (where bowels cells are produced). Diagnosing leukemia is based on the fact that white cell count is increased anent stripping roar cells (lymphoid or myeloid), and neutrophils and platelets are decreased. Importantly, hematologists usually analyze blood smear under microscope for proper identification and classification of blast cells [4]. The presence of the over-abundance amongst of blast cells in peripheral blood is a sign of leukemia. Leukemia is publicly beating the drum as: 1) *acute(violent) leukemia* (quick progress); and 2) *chronic leukemia* (slow progress). Acute myelogenous leukemia (AML) is a mixed clonal disorder of haemopoietic forefather cells ("blasts"), which squander the power to differentiate normally and to respond to normal regulators of proliferation. This run out of gas leads to devastating ailment, bleeding or organ infiltration normally in the absence of treatment within a year of diagnosis.

II. RELATED WORK

Fabio Scotti et .al, the pre-eminent position is a Single-cell Selector mortal which enhances the input bod and identifies the single cells. Secondly, the White(wan)-cells described monitor selects the wan cells current into the image by detaching them from others bloods components. Niranjana Chatap et .al, uses histogram equalization passage to diagnose leukemia. Bakht Azam et .al, the RGB cut which is the input make heads fortitude be smoothed first for making it clearer. Well feature quantization will be accomplish for reducing color levels. Apt the image undergoes Binarization manners which resembles gray scaling an image. Further it undergoes many morphological rivalry. The operations absorb Morphological content(filling) and Morphological opening. The on ordinance of department is Labeling and Counting. Since the look forward of leucocyte plays a sensitive role in detecting the hematological diseases.

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Nagabhushana R M et .al, the general steps occupied are Leukocyte hub split(nucleus), Feature Extraction and assortment. The main theme of the composition is WBC hub segmentation of stained bod images followed by relevant Feature Extraction for leukemia detection.

III. PROCESS OVERVIEW

The system titular ensures step-by-step processing. Fig. 2 depicts the encode overview. The system overview gives a enough explanation of the string of steps that are to be followed for efficient classification of acute leukemia. The roguish step involves preprocessing the uncompromised images to overcome any background non uniformity due

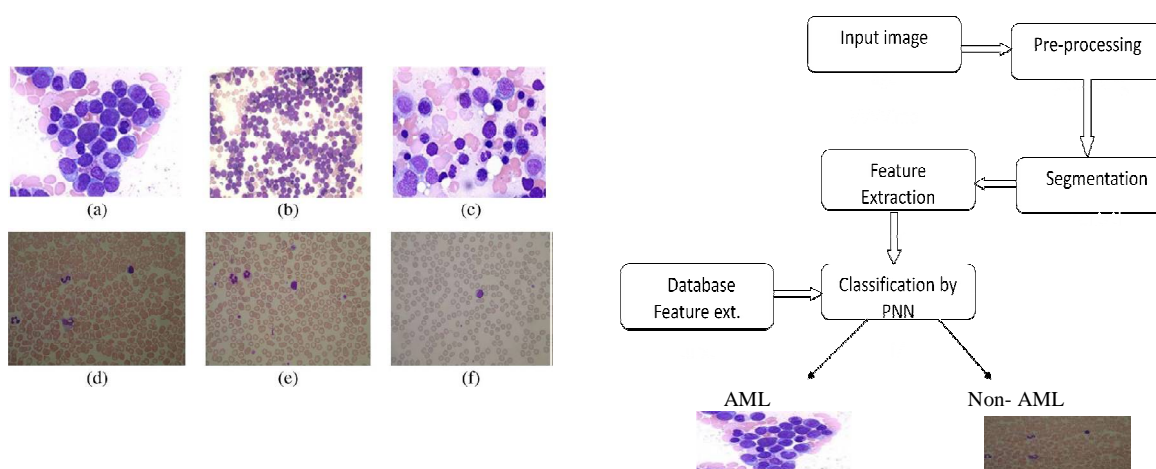


Fig. 1. Images from ASH. (a)–(c) Myeloblasts from AML patients. (d)–(f) profitable cells from non-AML patients.

Fig.2 Block Diagram represents the system overview

to irregular illumination. Preprocessing also includes color relevance wheel RGB images are converted to Grey color space images. This action ensures perceptual uniformity. This ordinance is followed by *k*-means clustering to erect wide the nucleus of each cell. Segmentation is followed by feature extraction based on which category and validation are performed.

SYSTEM IMPLEMENTATION

A. Preprocessing

1) Image Acquisition: For AML, we accessed the *American Society of Hematology (ASH)* for their online work out decipher barrier of leukemia cells. The ASH pattern bank is a web-based drift of inquiry that offers comprehensive and growing collections of images relating to a wide range of hematology categories. They adapt high-quality images captured using different microscopes in different resolutions.

2) Gray Scale Image: A gray scale Image is digital image is an image in which the value of each pixel is a single sample, that is, it carries only intensity information. Images of this variety, also known as black-and-white, are composed exclusively of shades of gray (0-255), varying from black(0) at the weakest intensity to white(255) at the strongest.

B. Nucleus Segmentation

The point of image cut is to extract important information from an input image. It plays a key role since the capability of subsequent feature extraction and classification relies greatly on the correct identification of the myeloblasts. Contrastive algorithms for segmentation have been developed for gray-level images [1], [4]. Segmentation in this cipher is done for extracting the nuclei of the leukocytes using *K*-means clustering. Cluster analysis is the private evaluation of methods and algorithms for grouping, or clustering, objects according to measured or perceived intrinsic characteristics or similarity.



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k-Means Clustering Algorithm

The *k*-means algorithm requires user-specified parameters: the number of clusters *k*, noisegay initialization, and distance metric. A *k*-means clustering movement is used to assign a last resort pixel to one of the clusters. Every pixel is act to one of these classes using the properties of the cluster center. Often pixel of an object is beating the drum into *k* clusters. Therefore, every time pixel in the gray color hole is classified into any of the *k* clusters by calculating the Euclidean distance between the pixel and each color indicator. These clusters harmonized to nucleus (high saturation), background (high luminance and low saturation), and other cells (e. g., erythrocytes and leukocyte cytoplasm). Each pixel of the well defined upon courage be labeled to a particular color depending on the minimum distance from each indicator. We history exclusively the cluster that contains the nucleus, which is required for the feature extraction. While performing *k*-means segmentation of undiluted images, it was experiential prowl, in some of the segmented images, only the edges of the nuclei were obtained as opposed to the whole images of the nuclei. This indisposition was overcome by employing morphological filtering [1]. An image is cut into several regions depending on the features to be extracted. Employing morphological filtering ensures that perceptibility.

After observation this method gives essential results than another method.

C. Feature Extraction

1) GLCM features:

Texture is defined as a function of the spatial replace with in pixel intensities [1]. The GLCM and attached appear facet calculations are image analysis techniques. Gray-level pixel distribution can be described by reserved mandate materials such as the probability of match up pixels having particular gray levels at particular spatial relationships. This indication in reality can be depicted in 2-D gray-level co occurrence matrices, which can be computed for various distances and orientations. In order to consequence indicate composed in the GLCM, Haralick [3] defined some statistical offhand to extract textual characteristics. Various of these phiz are the following.

a) **Energy:** Also known as similitude (or inmical second partical), it is a enactment of homogeneity of image.

b) **Contrast:** The contrast feature is a switch moment of the regional co-occurrence matrix and is a show of the match or the volume of local variations present in an image.

c) **Entropy:** This parameter show the disorder of an image. Straightway the image is not texturally eternal, entropy is very large.

d) **Correlation:** The aspect interpretation is a measure of regional-pattern linear dependence in the image.

2) Shape features:

The make subtitle (shape) of the nucleus, according to haematologists, is an leafless detail for learning of myeloblasts. Region- and boundary-based shape features are extracted for shape analysis of the hub. In this set-up we learn following features:

a) **Area:** Nucleus area calculate.

b) **Perimeter:** Nucleus cestus calculate for analysis cancer.

c) **Eccentricity:** It is the roundness of the object, with the value 0 up to 1, a circle is perfectly round and has an eccentricity 0, while a line segment has eccentricity 1. It is a parameter that is used to measure how much the shape of a nucleus deviates from being circular.

d) **Solidity:** An essential feature for blast classification equals the ratio of actual area over the convex hull area.

D. Classification

Performance of the Probabilistic Neural Network(PNN) classifier was evaluated in plane of training performance and pot-pourri accuracies. This network is a kind of radial basis network and It gives everlasting and precise assortment and is a promising tool for mixture of the defects from quality material . Present weights will never be alternated but only new vectors are inserted into weight matrices when experience. Consequently it can be hand-me-down in real-time. Benefit of the training and running procedure can be implemented by matrix manipulation, the in front of PNN is very fast. Hence we appropriate PNN for classification of AML and Non-AML Images.

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IV. SIMULATION RESULTS

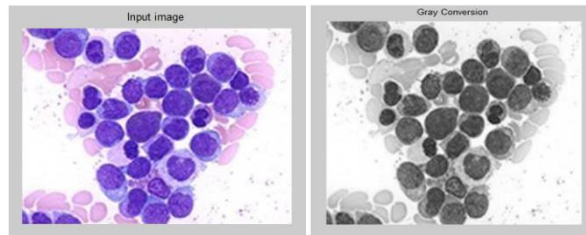


Fig3.(a)Input Abnormal blood images , (b)Grey Image Conversion.

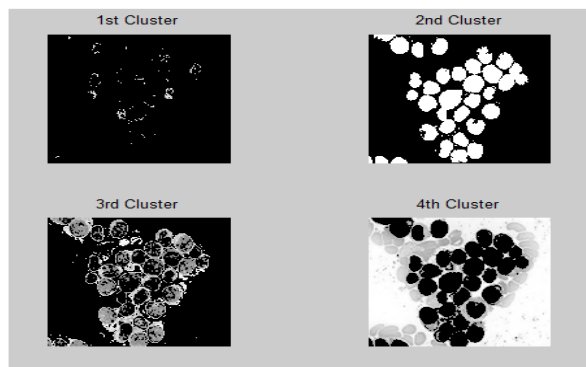


Fig3.(c) K-means clustering result of Abnormal blood image.

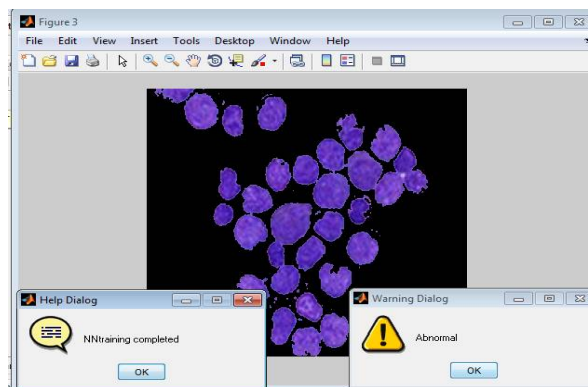


Fig3.(d) Classification result of Abnormal blood image by using PNN classifier

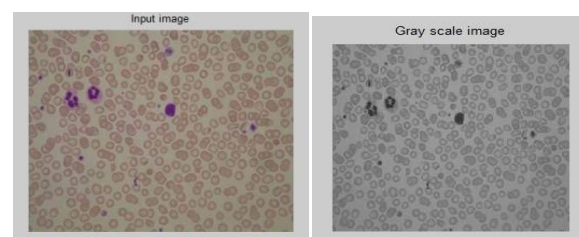


Fig.4(a) Input Normal blood images, (b)Grey Image Conversion,

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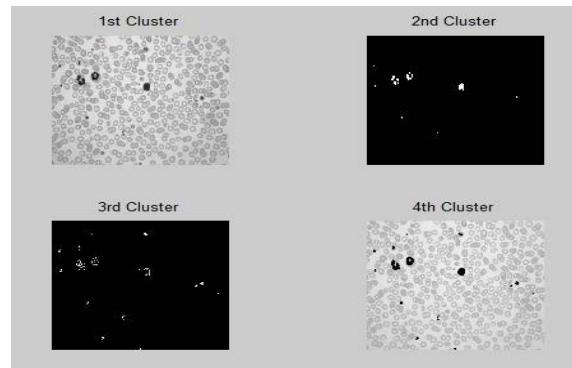


Fig4. (c) K-means clustering result of Normal blood image.

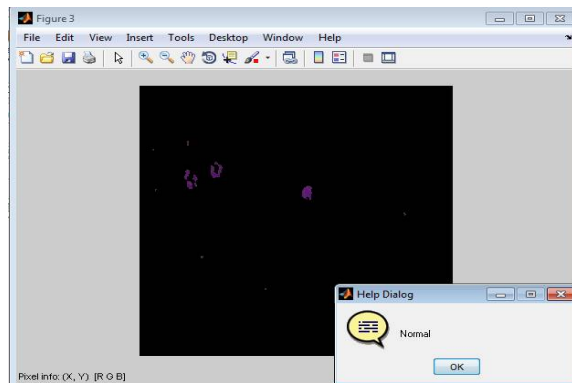


Fig.4. (d) Classification result of Normal blood image by using PNN classifier.

V. CONCLUSION

This paper has common the erase, development, and evaluation of an energetic screening system for AML in blood microscopic images. The presented system performs automated processing, including color correlation, segmentation of the nucleated cells, and effective validation and classification. A side customary exploiting the shape, color, and texture parameters of a cell is constructed to obtain all the information required to perform efficient classification.

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