

Automated Dengue Detection

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Abstract - Dengue is a major health problem in tropical and Asia-Pacific regions which typically spreads rapidly in number of infection patients. Knowing that most of the world's population living in risk areas, in order to diagnose and treat the disease, high skilled experts and human resources are needed. However, in some cases human error potentially may occur. Therefore, a new method of automated dengue detection is developed. For that white blood cell classification and classified the dengue as infected or not. In this paper, comparison based on SVM and two layer feed forward network. From that, two layer feed forward network has better accuracy than SVM. This work is also implemented in CNN.

Key Words: infected, non- infected, white blood cell, SVM, CNN

1. INTRODUCTION

The Dengue is a major health problem in tropical and Asia Pacific regions, and it has been spreading for rapidly more than 50 years. The diseases are caused by viruses transmitted of mosquito borne. Each year, 50-100 million people had been dengue infection. 75% of the world's population lives in high risk areas [1]. The health and economy affected. The major symptom of patients appears as a fever and plasma of blood leakage that affected shock syndrome and causes of death. Risk reduction is important factors of patients are getting diagnosed and should be treated correctly and quickly.

Dengue is spread by several species of mosquito of the Aedes type, A. aegypti. The virus has five different types; infection with one type usually gives lifelong immunity to that type, but only short-term immunity to the others. Subsequent infection with a different type increases the risk of severe complications. A number of tests are available to confirm the diagnosis including detecting antibodies to the virus or its RNA. A novel vaccine for dengue fever has been approved and is commercially available to confirm diagnose including detecting antibodies to virus or its RNA.

Symptoms typically begin three to fourteen days after infection. This may include a high fever, headache, vomiting, muscle and joint pains, and a characteristic skin rash.

Recovery generally takes two to seven days. In a small proportion of cases, the disease develops into the life-threatening dengue hemorrhagic fever, resulting in bleeding, low levels of blood platelets and blood plasma leakage, or into dengue shock syndrome, where dangerously low

blood pressure occurs.

Dengue has become a global problem since the Second World War and is common in more than 110 countries. Each year between 50 and 528 million people are infected and approximately 10,000 to 20,000 die. The earliest descriptions of an outbreak date from 1779. Its viral cause and spread were understood by the early 20th century. Apart from eliminating the mosquitoes, work is ongoing for medication targeted directly at the virus. It is classified as a neglected tropical disease.

2. LITERATURE SURVEY

Previous research is used to classify white blood cells based on the edges of cells. The focus is selection of feature extraction and classification method.

S.S.Savkare, S.P.Narote [2] proposed median filter which is used to remove noise.

Kaushiki Roy, Ratnadeep Dey, Debotosh Bhattacharjee, Mita Nasipuri, Pramit Ghosh [3], average filter is developed. It has advantage that single pixel representation and disadvantage that autocorrelation.

Prabhjot Kaur, Vishakha Sharma, Nishatha Garg [4] proposed a method called Gaussian low pass filter. It has advantage that it preserves noise but not preserves details.

Sheng-Fuu Lin, Yu-Bi Hong [5] method of region growth segmentation is presented. It has advantages that correctly separate the region and identify and set the boundary. It disadvantage that choice of kernel and inexpensive and sensitive to noise.

Ashwini Rejintal, Aswini N [6] proposed GLCM method. It advantage has Computational complexity of this method is very less. It disadvantage that ignores the spatial relationship between the texture patterns and sensitive towards noise.

Sarach Tantikittil, Sompong Tumswadi, Wichian Premchaiswadi [7] proposed Decision tree method. It advantage has Assign values to problem. It has disadvantage that Very complex.

Alfred R. J. Katz [8] did a research for classification of five major types of white blood cells. The best accurate technique for classification of supervised learning is a neural network technique.

Ms. Minal D. Joshi, Prof Atul H. Karode, and Prof S.R.Suralkar [9] did a research and studied white blood cell detection and classification of acute Leukemia by image of blood cells. And classification using a K-nearest neighbor (kNN), with 93% of accuracy.

Monica Madhukar and colleagues [10] research decision support tools for the classification of acute lymphoblastic leukemia. They used K-means Clustering techniques of the Lab color model for image segmentation at accuracy 93.5%.

Different techniques are applied and for preprocessing median filter, segmentation morphological thersholding, feature extraction SIFT and classification SVM. A white blood cells classification processing to apply the detection of dengue.

3. RESEARCH METHODOLOGY

In the proposed system, blood images are captured by digital microscope. In preprocessing, these image contain noise so median filter is used to remove noise. It is segmented by Morphological thersholding segmentation. In feature extraction, features of white blood cell nucleus is extracted by SIFT algorithm .By SVM classifier classify as infected or non infected.

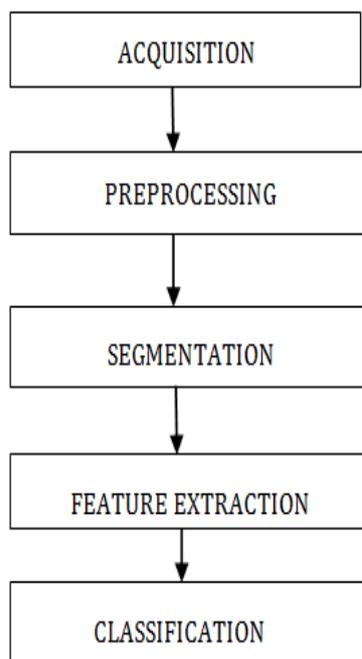


Fig -1: Systematic Flow

A. Acquisition

Images of blood smear are captured by microscope attached with camera. Blood stain used is Leishman blood smear.

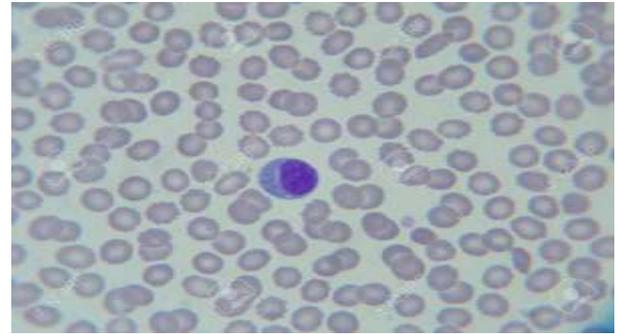


Fig -2: Image Acquisition

B. Preprocessing

Poisson noise: Poisson noise or Shot noise is a type of electronic noise which can be modeled by a Poisson process. Shot noise results from discrete nature of electric charge. Shot noise may be dominant when the finite number of particles that carry energy is sufficiently small so that uncertainties due to the Poisson distribution, which describes the occurrence of independent random events. The magnitude of shot noise increases according to the square root of the expected number of events, such as the electric current or intensity of light. But since the strength of the signal itself increases more rapidly, the relative proportion of shot noise decreases and the signal to noise ratio (considering only shot noise) increases anyway.

Median filter: The median filter is non-linear filter. It removes noise effectively as well as preserving sharp edges. A median filter is more effective than convolution when the goal is to simultaneously reduce noise and preserve noise[2]. It simply replaces each pixel value by the median of the intensity level in the neighborhood of the pixel.

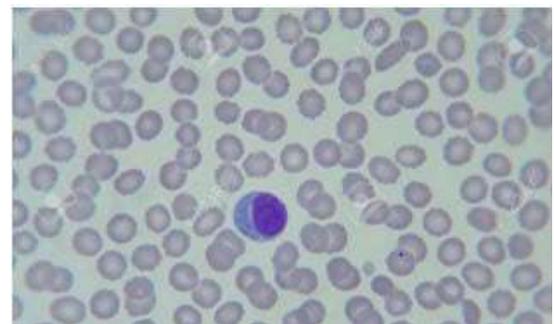


Fig -3: Preprocessed input image.

C. Segmentation

For segmentation thersholding is used to extract infected wbc cell. The simplest thersholding methods replace each pixel in an image with a black pixel if the image intensity

$I(i,j)$ is less than some fixed constant $T(I(i,j) < T)$, or a white pixel if image is greater than constant.

From a grayscale image, thresholding can be used to create binary images. This is the method of separating foreground and background of image. All pixels with a value higher than the threshold value are classified as feature pixels. All pixels with a lower value are classified as background pixels.

Such a distribution is called bimodal because there are two mode values: one for the background and one for the feature

$$I_T(x,y) = \begin{cases} 1 & \text{for } I(x,y) \geq T \\ 0 & \text{for } I(x,y) < T \end{cases}$$

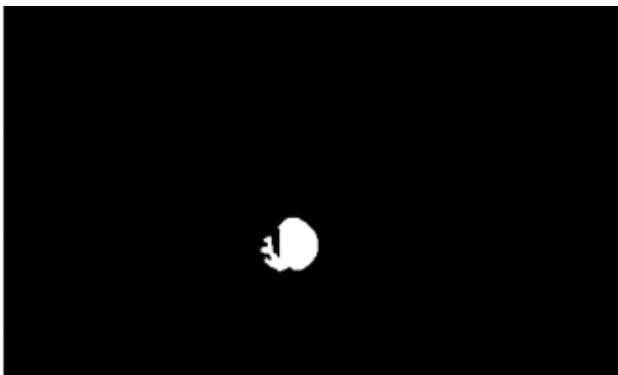


Fig -4: Segmented image

D. Feature Extraction

SIFT is quite an involved algorithm. It is used to find local features.

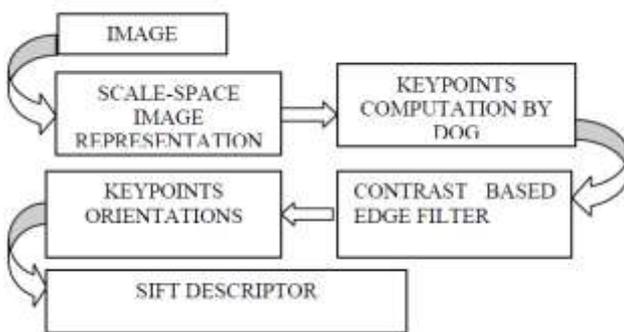


Fig -5: SIFT Algorithm

1. Constructing scale space: This is the initial preparation. You create internal representations of the original image to ensure scale invariance. This is done by generating a "scale space".

2. LoG Approximation: In the previous step, we created the scale space of the image. The idea was to blur an image progressively, shrink it, blur the small image progressively

and so on. Now we use those blurred images to generate another set of images, the Difference of Gaussians (DoG). These DoG images are a great for finding out interesting key points in the image.

3. Finding Key points: Finding key points is a two part process. a) Locate maxima/minima in DoG images: The first step is to coarsely locate the maxima and minima. This is simple. You iterate through each pixel and check all its neighbours. If pixel is larger than its neighbors then it is maxima.

$$D(x) = D + \frac{\partial D^T}{\partial x} x + \frac{1}{2} x^T \frac{\partial^2 D}{\partial x^2} x$$

4. Get rid of bad points: Key points generated in the previous step produce a lot of key points. Some of them lie along an edge, or they don't have enough contrast. If the magnitude of the intensity (i.e., without sign) at the current pixel in the DoG image (that is being checked for minima/maxima) is less than a certain value, it is rejected.

5. For each key point P a squared region R around P is considered and partitioned in 4x4 parts. An histogram with 8 bins is used for representing the orientation of the points in each of the sub-regions of R. The final descriptor associated to P is a vector that concatenates the histograms of the sub- regions of R.

The descriptor vector has $(4 \times 4) \times 8 = 128$ entries.

E. Classification

This is the final step to detect the given input is infected or not. Support Vector Machine (SVM) algorithm was chosen as it is a promising nonlinear, non parametric classification technique. The reduced feature set was given as input to the algorithm. In SVM, the input is considered as an n-dimensional 12 vector, and the question is whether such points can be separated with a hyperplane. Multiple hyper planes might be generated that can classify the data.

From the SVM, accuracy is 70%, sensitivity is 80%, specificity is 60%, and precision is 66.7%. In order to improve accuracy, the classification can be done in artificial neural network (ANN).

F. Two layer feed forward network.

A feed forward neural network is an artificial neural network wherein connections between the units do not form a cycle. It consists of a (possibly large) number of simple neuron-like processing units, organized in layers. Every unit in a layer is connected with all the units in the previous layer.

The weights on these connections encode the knowledge of a network. Often the units in a neural network are also called nodes[8]. Data enters at the inputs and passes through the network, layer by layer, until it arrives at the

outputs. During normal operation, that is when it acts as a classifier, there is no feedback between layers.

The operation of this network can be divided into two phases:

1. learning phase: The FFN uses a supervised learning algorithm: besides the input pattern, the neural net also needs to know to what category the pattern belongs. Learning proceeds as follows: a pattern is presented at the inputs. The pattern will be transformed in its passage through the layers of the network until it reaches the output layer. The units in the output layer all belong to a different category. The outputs of the network as they are now are compared with the outputs as they ideally would have been if this pattern were correctly classified: in the latter case the unit with the correct category would have had the largest output value and the output values of the other output units would have been very small. On the basis of this comparison all the connection weights are modified a little bit to guarantee that, the next time this same pattern is presented at the inputs, the value of the output unit that corresponds with the correct category is a little bit higher than it is now and that, at the same time, the output values of all the other incorrect outputs are a little bit lower than they are now. The differences between the actual outputs and the idealized outputs are propagated back from the top layer to lower layers to be used at these layers to modify connection weights.

2. Classification phase: In the classification phase, the weights of the network are fixed. A pattern, presented at the inputs, will be transformed from layer to layer until it reaches the output layer.

Table -1: Comparison between SVM and ANN.

	SVM	ANN
Accuracy	70	80
Precision	66.7	80
Sensitivity	80	80
Specificity	60	80
Prevalence	75	80

4. CNN BASED RECOGNITION SYSTEM

In machine learning, a convolution neural network (CNN, or ConvNet) is a class of deep, feed-forward artificial neural networks that has successfully been applied to analyzing visual imagery. CNNs use a variation of multilayer perceptions designed to require minimal preprocessing. They are also known as shift invariant or space invariant artificial neural networks (SIANN), based on their shared-weights architecture and translation invariance characteristics.

Convolutional networks were inspired by biological processes in that the connectivity pattern between neurons resembles the organization of the animal visual cortex. Individual cortical neurons respond to stimuli only in a restricted region of the visual field known as the receptive field. The receptive fields of different neurons partially overlap such that they cover the entire visual field.

CNNs use relatively little pre-processing compared to other image classification algorithms. This means that the network learns the filters that in traditional algorithms were hand-engineered. This independence from prior knowledge and human effort in feature design is a major advantage.

A CNN consists of an input and an output layer, as well as multiple hidden layers. The hidden layers of a CNN typically consist of convolutional layers, pooling layers, fully connected layers and normalization layers.

CNN generally requires more than 1000 samples. In these cases, dengue sample is less so that output has less accuracy.

5. EXPERIMENTAL RESULTS

Testing method is a method for to measure the performance of each model that was created. The accuracy is 70%, sensitivity is 80%, specificity is 60%, precision is 67% and prevalence is 80%.

The table shows comparison between existing method and proposed method. In these paper, testing is carried out by few samples. If input samples is improved then accuracy is increases.

Table -2: Comparison between Existing and proposed method.

	Existing method	Proposed method
Accuracy	65	75
Precision	62	73.3
Sensitivity	71.2	70
Specificity	60	80
Prevalance	60	77.5

6. CONCLUSION

The proposed is based on white blood cells and it avoids complexity of lab technicians and reduces errors. The steps involved are preprocessing, segmentation, feature extraction and classification. The microscopic blood images contain noise and it can be removed by median filter. The segmentation is followed by Thersholding in which white blood cells is extracted. The nucleus is extracted by SIFT feature extraction technique. Then classify infected or non-infected using SVM classifier.

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