INTRODUCTION

The Retinal vessel segmentation can be utilized to diagnosis, screening, treatment, and evaluation of cardiovascular and ophthalmologic diseases like diabetes, hypertension, arteriosclerosis and choroidal neovascularisation. We can utilize attributes of retinal blood vessels like length, width, tortuosity, angles and branching pattern [2]. The retinal vessels automated segmentation is considered as a first step for ophthalmic disorder in computer aided diagnosis system development [3]. The diabetic retinopathy [4], vessel diameter measurement in relation with diagnosis of hypertension [5], computer assisted laser surgery [6] etc are screening programs. They are implemented by analysis and automatic detection of the vasculature [7]. A large number of death and visual loss has been occurred due to spreads of diabetes on retina [8].

The supervised and unsupervised are the two categories on which automatic segmentation method of retinal blood vessels is categorised. The manually labelled samples are used in case of supervised learning for already trained classifier. The vessels and non vessels are the two classes on which samples pixels are classified. They depend on pre-classified data that may not be available in real life applications. Mainly in health images, the training used in this method makes it better than unsupervised methods in terms of performance.

The use of retinal vascular tree in biometric identification is unique for every individual data [9,10]. The manual segmentation of retinal blood vessels is very tedious, long task and it also requires training. The retinal blood vessels automatic segmentation is proved to be very challenging due to various structures complication that is influenced by other sources.

OPTIMIZED RETINAL BLOOD VESSEL SEGMENTATION TECHNIQUE FOR DETECTION OF DIABETIC RETINOPATHY

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Abstract: Most of the retinal diseases namely retinopathy, occlusion etc., can be identified through changes exhibited in retinal vasculature of gathered images. Thus, segmentation of retinal blood vessels aids in detecting the alterations and hence the disease. In one data set there are large numbers of features but not all are useful in classification. The useless redundant features reduce the performance of output. In our case we have used only small relevant features rather than using all the features that contain both relevant and irrelevant features. To achieve better results there is need to minimize the number of features along with maximizing the performance of classification. In this paper, to achieve the objective firstly we have extracted the features using the Principle component analysis (PCA) for both series of vessels and non vessel images (normal and abnormal). Then further extracted features are optimized using various optimization methods which actually optimize the selected features of images taken from DRIVE dataset for both vessels and non-vessels to improve the recognition accuracy. Afterwards, the naïve bayes classifier has been applied to test the results for both normal and abnormal images dataset. The accuracy of PSO optimization is 0.99537 which is 31.42% better than LION and 53.09% better than Firefly. The simulation results show that PSO is better in terms of accuracy, sensitivity, entropy, etc., as compared to other optimization methods for both normal and abnormal set of images.

Keywords: Vessel images, Non vessel images, Particle Swarm Optimization (PSO), Firefly, LION, Accuracy, Sensitivity
range they have used Gamma correction techniques. In last they have used CLAHE to avoid amplification of unwanted noise present in image and its parameters also help to limit the contrast in homogeneous area. In x-ray images and in other images it is critical to perform image enhancement for getting good segmentation. In autonomous disease diagnosis system, subsequent modules are used to increase image sharpness using magnification of the contrast. In this paper [13], authors have analyzed various pre-processing techniques for segmentation of vertebral bone. They have considered histogram equalization (HE), gamma correction (GC) and contrast limited adaptive histogram equalizer (CLAHE) three methods. The experimental results show that most accurate technique is CLAHE as compared to other techniques. The use of CLAHE gives good result that is why in this paper we have used this technique for image enhancement [14]. In first step we have uploaded an image and then pre-processed it using Contrast Limited Adaptive Histogram Equalization (CLAHE) technique.

Section 2: Edge detection of segmentation and Feature Extraction (PCA) The second section and step is to detect the edges.

In automated retinopathy analysis system an important clinical role is played by blood vessel detection and segmentation technique. There is need of expertise and time to manually segment an image [15]. An image is divided into essential regions using various methods. The local or abrupt changes are considered for segmenting image using edge detector. The retinal analysis will become efficient using computational intelligence. In our paper, we have used MATLAB graphical user interface (GUI) tool to segment a blood vessels and then extracted the features using Principle component analysis (PCA). In paper [16], authors have used DRIVE database and achieves accuracy of 94.61% with 73.38% of sensitivity as compared to existing techniques of blood vessel segmentation. The image can be detected using various existing techniques which identify an image and then locate sharp discontinuities in it. The discontinuities are nothing it is just abrupt changes in pixel density that characterized boundaries [17]. In case of noisy images it is difficult to perform edge detection because there are a large frequency content which is available in both edges and noises. There are number of edge detection techniques available such as Sobel, Robert, Canny, Prewitt and Laplacian of Gaussian (LoG).

Figure 1: Features extracted for normal image

After segmenting an image using edge detection the next step is feature extraction. The features are extracted using Principle component analysis (PCA) from series of vessels and non vessel images [18]. There is some information which is present in one single image and to extract that we have used feature extraction method. The pattern recognition, feature extractions are the parts of pre-processing steps that helps in extracting and building features which are of foremost importance. In feature extraction we mainly reduce the dimensionality that restricts the resources which is the main set of required data. This reduced and required information will be further manipulated according to our requirement [19]. In our work we want to enhance blood vessels images for this we have used PCA as a feature extractor.

Figure 2: Feature extracted for abnormal image

Section 3: Feature Optimization (Firefly + PSO + LION)

In second section features has been extracted for both normal and abnormal images using PCA. In this section we will optimize the extracted features using different algorithms. Firstly we have applied particle swarm optimization (PSO) on extracted features. The PSO is proving to be very efficient for feature extraction due to its ability of rarely falling into local optimal. In paper [20], authors have given different pros and cons of PSO algorithms that need to be resolved. PSO is a meta-heuristic nature inspired technique that simulates the bird flocking behaviour. It uses randomly generated population that search for optimal solution by updating velocity and position according to gbest and pbest value. The MATLAB is used for performing simulation results in terms of different parameters [21]. After the feature optimization process, the test images will be classified as normal and abnormal images using classification process using Naive Bayes classifier and performance will be evaluated. The features are optimized and simulation results are given for both normal and abnormal images [22].

Figure 3: Normal image Feature optimized using PSO
Table I: This table shows the simulation results for normal images using PSO, Firefly & LION:

| Algorithm | Normal images | Sensitivity | Specificity | Accuracy | Positive Predicted | False Positive Rate |
|-----------|--------------|-------------|-------------|----------|--------------------|---------------------|
| PSO       |              | 0.99766     | 0.99539     | 0.99537  | 0.99539            | 0.00463             |
| Firefly   |              | 0.0237      | 0.51186     | 0.46443  | 0.04633            | 0.48814             |
| LION      |              | 0.99435     | 0.68678     | 0.68112  | 0.76045            | 0.31322             |

Table II: This table shows the simulation results for Abnormal images using PSO, Firefly & LION:

| Algorithm | Abnormal images | Sensitivity | Specificity | Accuracy | Positive Predicted | False Positive Rate |
|-----------|----------------|-------------|-------------|----------|--------------------|---------------------|
| PSO       |                | 0.99448     | 0.70136     | 0.69584  | 0.76906            | 0.29864             |
| Firefly   |                | 0.023714    | 0.51186     | 0.46443  | 0.04633            | 0.48814             |
| LION      |                | 0.99435     | 0.68678     | 0.68112  | 0.76045            | 0.31322             |

The above shown Table I and Table II gives the results of PSO, Firefly and LION algorithm in terms of Sensitivity, Specificity, Accuracy, Positive and False positive predicted value for both normal and abnormal images.
After the feature optimization process, the test images will be classified as normal and abnormal images using classification process using Naive Bayes classifier and performance will be evaluated.

Figure 8: Classification as Normal and Abnormal images using Naïve Bayes classifier

Table III: This table shows the simulation results for normal images using PSO, Firefly & LION

| Optimization Technique | Normal image |          |          |          |
|------------------------|--------------|----------|----------|----------|
|                        | Entropy      | Min Pixel| Max Pixel| Intensity |
| PSO                    | 3.3232       | 0.16909  | 2.4084e-07| 0.015802 |
| Firefly                | 0.090793     | 99.6888  | 0.12702  | 49.9441  |
| LION                   | 3.6556       | 0.072513 | 0        | 0.039911 |

Table IV: This table shows the simulation results for abnormal images using PSO, Firefly & LION

| Optimization Technique | Abnormal image |          |          |          |
|------------------------|----------------|----------|----------|----------|
|                        | Entropy        | Min Pixel| Max Pixel| Intensity |
| PSO                    | 3.6803         | 0.4352   | 1.7083e-05| 0.029855 |
| Firefly                | 0.080793       | 99.1421  | 0.95455  | 50.9886  |
| LION                   | 3.6992         | 0.071176 | 0        | 0.036736 |

The above shown Table3 and Table4 gives result of PSO, Firefly and Lion algorithms in terms of Entropy, Min Pixel, Max Pixel and intensity for both normal and abnormal images.

CONCLUSION

In this paper, traditional feature selection methods and evolutionary algorithms such as PSO, Firefly and LION for feature selection has been investigated. We have used DRIVE dataset and all data sets contain both redundant and useful data. Our aim is to only work on useful data that will give fruitful and accurate results. The use of redundant data will reduce the success rate of getting the output. We have used vessels and non vessels images after uploading and processing the input image. The images will be pre-processed using CLAHE algorithm and edges are detected by performing segmentation. The features of images are extracted using PCA algorithm for both normal and abnormal images. The normal images are those that don’t have symptoms of diabetic retinopathy and non vessels or abnormal images are vice versa. The extracted features are optimized by applying different optimization algorithms. We have used MATLAB and analyzed results in terms of sensitivity, Specificity, Accuracy, Positive and false positive predicted rate. The accuracy of PSO optimization is 0.99537 which is 31.42% better than LION and 53.09% better than Firefly. The results are good for normal images but in case of abnormal images the results are not very efficient. To further increase the accuracy for abnormal images a hybrid approach is desirable.

ACKNOWLEDGEMENT

Authors are highly thankful to the RIC department of IKG Punjab Technical University, Kapurthala, Punjab, India for providing the opportunity to conduct this research work.

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