



Intermittent Detection of Diabetic Retinopathy

Ibrahim Patel¹, V. Sripathi Raja², Ashok Shigli³

Assoc. Prof., Dept. of ECE, B. V. Raju Inst. of Tech., Narsapur, Medak, T. S. India¹

Assoc. Prof., Dept. of BME, B. V. Raju Inst. of Tech., Narsapur, Medak, T. S. India²

Prof. & HOD, Dept. of BME, B. V. Raju Inst. of Tech., Narsapur, Medak, T. S. India³

Abstract: Diabetic retinopathy is retinopathy *i.e.*, injury to the membrane, caused by complications of polygenic disorder, which may eventually cause cecity. People suffering with type-II polygenic disorder of people suffering with type-I polygenic disorders for more years are prone to diabetic's retinopathy. It's associate ocular manifestation of polygenic disorder a general sickness that affects up to eighty p. c of all patients who have had polygenic disorder for ten years or a lot of. Despite these daunting statistics, analysis indicates that a minimum ninetieth of those new cases may well be reduced if there have been correct and argus-eyed treatment and observation of the eyes. Robotic image process has the potential to help within the early detection of d polygenic disorder. Exudates, haemorrhages and small aneurysms square measure the first signs of diabetic retinopathy. Image process techniques will scale back the work of ophthalmologists and therefore, the tools used mechanically find the abnormalities. This paper proposes a technique for the retinal image analysis through professional detection of exudates, small aneurysms and haemorrhages and recognizes the membrane to be traditional or abnormal.

Keywords: Diabetic retinopathy, Blindness, Ocular, Exudates, Retinal Image.

I. INTRODUCTION

Diabetic retinopathy (DR) could be a polygenic disorder complication that affects eyes. It's caused by injury to the blood vessels of the photosensitive tissue at the rear of the attention (retina). At first, diabetic retinopathy could cause no symptoms or solely delicate vision issues. Micro-aneurysms area unit among the initial signs of diabetic retinopathy they arise because of high sugar levels within the blood. Consistent with United Nations agency (World Health Organization) there'll be seventy nine million individuals with polygenic disorder by 2030, creating the Bharat Diabetic capital of the planet. Eventually, it will cause visual impairment. The condition will develop in anyone United Nations agency has type-I or type-II polygenic disorder. The longer you have got polygenic disorder and also the less controlled your glucose is, the additional doubtless you're to develop this eye complication. Over time, an excessive amount of sugar in your blood will result in the blockage of the small blood vessels that nourish the tissue layer, alienating its blood provide. As a result, the attention makes an attempt to grow new blood vessels. However, these new blood vessels don't develop properly and may leak merely. Early detection likewise as a result of the periodic screening of DR likely helps in reducing the progression of this un wellness and in preventing the following loss of visual capability one won't have symptoms within the early stages of diabetic retinopathy as shown in figure 1. Because the condition progresses, diabetic retinopathy symptoms may include:

- Spots or dark strings floating in your vision (floaters)
- Blurred vision
- Fluctuating vision
- Impaired colour vision
- Dark or empty areas in your vision
- Vision loss



Fig.1: Image showing Normal vision and DR vision



II. REVIEW ON EARLY DETECTION

A lot of analysis work revealed on early diagnosing of diabetic retinopathy. Most of them are supported the detection of micro-aneurysms at delicate stage diabetic retinopathy has classified in seven categories:

- Mild Non-proliferative Retinopathy: At this earliest stage, small aneurysms occur. They're tiny areas of balloon-like swelling within the retina's small blood vessels.
- Moderate Non-proliferative Retinopathy: because the malady progresses, some blood vessels that nourish the tissue layer are blocked.
- Severe Non-proliferative Retinopathy: more blood vessels are blocked, depriving many areas of the tissue layer with their blood provide. These areas of the tissue layer send signals to the body to grow new blood vessels for nourishment
- Proliferative Retinopathy: At this advanced stage, the signals sent by the tissue layer for nourishment trigger the expansion of recent blood vessels. This condition is named proliferative retinopathy.
- Bleeding within the clear, jelly-like substance that fills the centre of the attention (vitreous)
- Retinal detachment
- Abnormalities in your optic nerve.

III. METHOD OF DETECTION OF DIABETIC RETINOPATHY

Screening is meted out through fundal examination performed by medical or optometric employees or by observation of assorted photographic ways. In those units exploitation photography, trained personnel area unit needed to screen the retinal pictures. Manual analysis and diagnosing needs a good deal of your time and energy to review retinal pictures that area unit obtained by structure camera. A screening methodology that doesn't need trained personnel would be of nice profit to screening services by decreasing their prices. A completely robotic approach involving machine analysis of structure pictures might offer an instantaneous classification of retinopathy while not the requirement for specialist opinions.

The employment of digital fundus imaging in eye doctor provides digitised provides digitized knowledge that might be exploited for computerized detection of un wellness. The presence of exudates could be a main hallmark of diabetic. Hence, detection of exudates is a very important diagnostic task that plays a serious role. The manual detection of those red lesions is created by ophthalmologists and could be a slow and erring activity. Some lesions have a really tiny size and should go neglected even by trained specialists. Thus, it's evident the necessity of strategies for automatic detection of red lesions in structure pictures as shown in figure 2.

Fluorescent X ray pictures square measure sensible for observant some pathology like small aneurysms and haemorrhages that square measure indicators of diabetic retinopathy. It's not a perfect methodology for associate automatic screening system since it needs associate injection of fluorescent into the body. Fluorescent X ray pictures square measure sensible for observant some pathologies like small aneurysms and haemorrhages that square measure indicators of diabetic retinopathy. It's not a perfect methodology for associate automatic screening system since it needs associate injection of fluorescent into the body.

Automated image processing has the potential to assist in the early detection of diabetes, by detecting changes in blood vessel patterns in the retina. Automated analysis techniques for retinal images have been an important area of research for developing screening programmers.

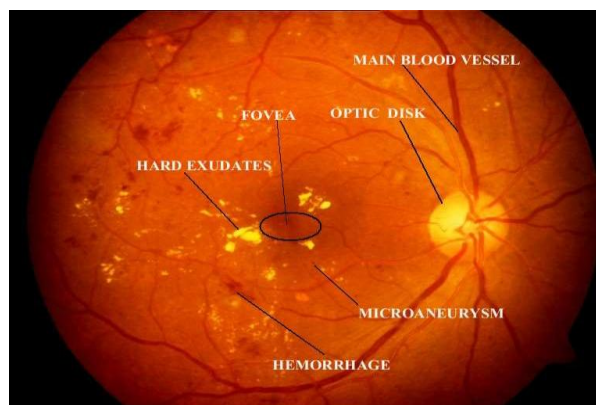


Fig.2: Retinal Image show abnormalities present in DR condition.



IV. PROPOSE METHOD

In our method, MATLAB software is used for programming and code is developed for diagnosis of retinal abnormalities. The abnormal regions are identified using segmentation, enhancement and morphological operations. The algorithms are simple, require less time, less complex and easy to implement. Algorithm does not depend on geometrical characteristics and the non-exudates regions are eliminated to a good extent. Thus, the characterised retinal abnormalities are identified and enhanced, thereby analyzing the DR conditions and providing an efficient diagnosis.

Acquisition of Retinal snapshots is foremost and crucial demand for the project. Once playacting ophthalmic anatomical structure photography for diagnostic functions, the pupil is expanded with eye drops and a special camera referred to as a anatomical structure camera is employed to specialize in the anatomical structure. The ensuing pictures may be spectacular, showing the optic tract through that visual "signals" square measure transmitted to the brain and also the retinal vessels that provide nutrition and chemical element to the tissue set against the red-orange colour of the pigment epithelial tissue.

Anatomical structure cameras with digital capture capabilities square measure typically used as screening devices for diabetic retinopathy and eye disease. Digital anatomical structure photography is a good methodology of retinopathy screening that's capable of detective work macular puffiness and proliferative diabetic retinopathy.

V. OPTICAL PRINCIPLE

The optical plan of fundus cameras is based on the principle of monocular indirect ophthalmoscopy. A fundus camera provides a vertical, magnified view of the fundus. A typical camera views 30 to 50° of retinal region, with a amplification of 2.5x, and allows some alteration of this association during zoom or auxiliary lenses from 15°, which provides 5x amplification, to 140° with a wide angle lens, which minifies the snapshot by half. The optics of a fundus camera associate alogous (is comparable) to those of an indirect ophthalmoscope in that the observation and illumination systems follow unrelated paths. The surveillance light is focused via a series of lenses through a doughnut shaped opening, which then passes through a central aperture to form an annulus, before passing through the camera objective lens and through the cornea onto the retina.

The light reflected from the retina passes through the un-illuminated hole in the doughnut formed by the illumination system. As the light paths of the two systems are independent, there are minimal reflections of the light source captured in the formed image. The image forming rays continue towards the low powered telescopic eyepiece. When the button is pressed to take a picture, a mirror interrupts the path of the illumination system allow the light from the flash bulb to pass into the eye. Simultaneously, a mirror falls in front of the observation telescope, which redirects the light onto the capturing medium, whether it is film or a digital CCD. Because of the eye's tendency to accommodate while looking through a telescope, it is imperative that the exiting vergence is parallel in order for an in focus image to be formed on the capturing medium.

VI. PERFORMANCE ANALYSIS

Algorithm development process for the detection of abnormalities is divided into three sections i.e., identification of:

- Exudates
- Micro aneurysms
- Haemorrhages

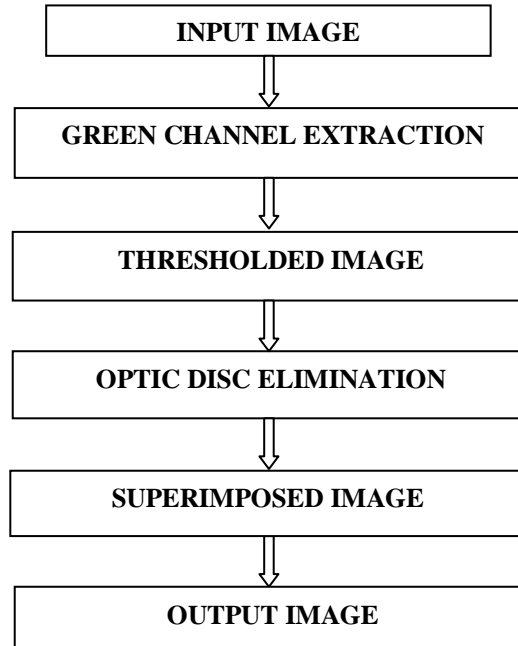
So, three separate algorithms are developed for tracing exudates, micro aneurysms and haemorrhages each, based on their characteristics and features. The detection of these features is essential for diagnosing eye diseases such as diabetic retinopathy. The standard reference for the stages and severity of DR is taken from the International standard reference scale, which was provided by the ophthalmologist.

A. Segmentation of Exudates: The process flow for identification of exudates includes the following steps. The algorithm is depicted in the form of flowchart -1:

Input Image: In the original fundus image as shown, the intensity variation between the bright objects (i.e. the optic disc and the exudates) and the blood vessels is relatively high and the vessels usually have poor local contrast with respect to the background. To isolate the optic disc and other bright parts is a tedious task and hence, pre-processing of the image for subsequent analysis becomes indispensable.



Green Channel Extraction: The green channel of the RGB colour space is extracted, which has better contrast when compared to the other channels. The gray-scale image $f1$ of each original fundus image f is obtained.



Flow chart-1: Segmentation of Exudates.

Contrast Adjustment: The contrast of the green channel image is adjusted. It maps the intensity values in gray-scale image $f1$ to new values in $f2$ such that 1% of data is saturated at low and high intensities of $f1$. This increases the contrast of the output image $f2$. Only the darker regions have their intensity values enhanced slightly while the brighter regions of the image remain more or less unchanged.

$$f2 = \text{Contrastadjust}(f1)$$

Threshold image: A threshold value T is chosen to obtain a binary image from image after contrast adjustment that isolates the bright parts from the background. The binary image obtained with $T=220$.

$$f3 = \begin{cases} 0 & \text{if } f2(i, j) < T \\ =20 & \text{else} \end{cases}$$

Optic Disc Elimination: Since the optic disc region of the retina non-exudate portion it is necessary to eliminate this portion. Thresholding of an image results in number of connected components such as part of optic disc, some noise and other bright features. These connected components are candidate regions for optic disc. The entire image is scanned to count the number of connected components. Each of the connected components in the threshold image is labelled, total number of pixels in the component and mean spatial coordinates of each connected component is calculated. The component having the maximum number of pixels is assumed to be having the optic cup part of disc and it is considered to be the primary region of interest. The maximum diameter of optic disc can be of 2mm. Therefore, in an image, if any of the components whose mean spatial coordinates are within 50 pixels distance from the mean spatial coordinates of the largest component, then they are merged with it and new mean spatial coordinate is calculated.

$$f4 = (\text{Max}(\text{Connected Components}(f3)) = 0)$$

Superimposed image: The exudates portion obtained from above step $f4$ is super imposed of original image f for clear perception.

$$f5 = \text{Superimpose}(f, f4)$$

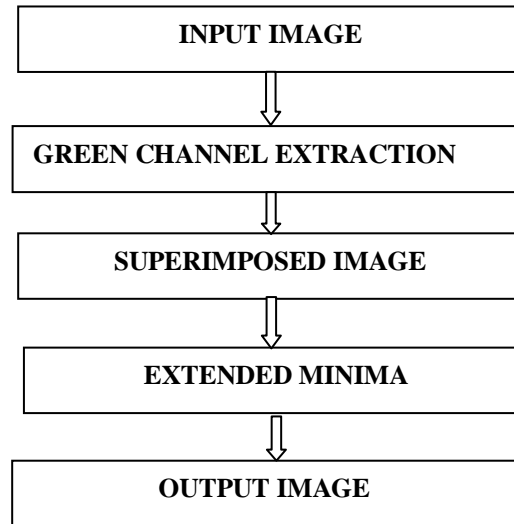
Thus, the exudates present in the retina of DR patient is segmented and viewed clearly.



Micro Aneurysms: The identification of the micro aneurysms is accomplished by the following steps:

- ❖ Input Image
- ❖ Extraction of Green channel
- ❖ Extended Minima
- ❖ Superimposing micro aneurysms of original image
- ❖ Output Image

The algorithm developed for Micro aneurysms is depicted in the following flow chart-2:



Flow chart-2: Identification of Micro-Aneurysms.

Input Image: The input image consisting of micro aneurysms is accessed from the library files. Micro aneurysms usually have a diameter less than 125 μm , and this small size and intensity variations in the background.

$f = \text{ReadImage}(\text{Original})$

Green Channel Extraction: As the Green channel has high intensity as compared to Red and Blue of RGB colour space, only this band is used for further processing. Also micro aneurysms are characterized as high contrast regions.

$f_2 = \text{GreenChannel}(f)$

Extended Minima: In order to highlight the red lesions which are low intensity structures the high intensity structures (optical disc and exudates) of f_2 are to be eliminated. This is done using H-minima transform. This operator removes connected basins with contrast less than a threshold h , using for this purpose a morphological reconstruction based on erosion. In the next step the image is binarized using the morphological operator of regional minimum. This operator converts a gray-scale image to binary format without using any threshold, and is also based on morphological reconstruction by erosion.

$f_3 = \text{RegionalMinima}(\text{H-Minima}(f_2))$

Superimposed Image: The image f_3 obtained after applying extended minima consists of only the traces micro aneurysms. These regions are superimposed on the original image for clear vision.

$f_4 = \text{Superimpose}(f, f_3)$

Hence, the high contrast featured micro aneurysms are efficiently segmented and shown.

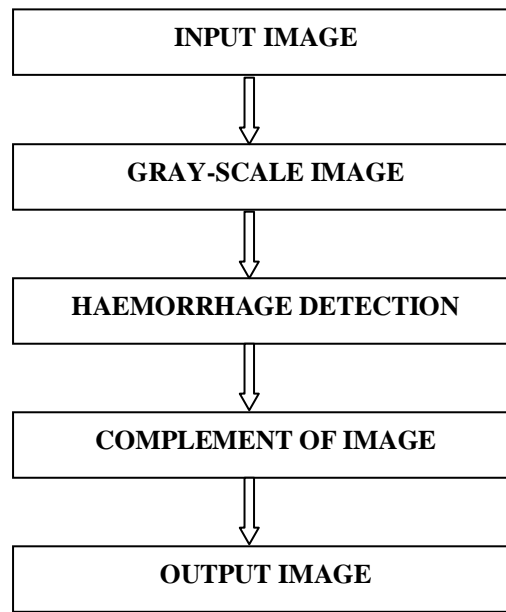
Haemorrhages: Haemorrhages are detected according to following steps:

- i. Input Image



- ii. Gray-Scale Image
- iii. Detection of Haemorrhage
- iv. Complement of Image
- v. Output Image

The algorithm developed for the recognition of haemorrhages is presented in the flow chart-3:



Flow chart-3: Detection of Haemorrhages.

Input Image: The retinal image having haemorrhages are taken as input image as initial step.

$$f = \text{Read Image}$$

Grey-Scale Image: The true colour image RGB f is converted to gray-scale intensity image $f1$. This is done by eliminating the hue and saturation information while retaining the luminance. Gray RGB colour code has equal red, green and blue values.

$$f2 = \text{rgbToGray}(f)$$

Detection of haemorrhage: The haemorrhage portion is detected using Image Segmentation technique, Sobel Edge Detector. Edges are pixels which carry important information in an image. Edge detection is the process of localizing pixel intensity transitions. The Sobel operator is an algorithm for edge detection in images discovers the boundaries between regions also it determine and separate objects from background in an image. It's an important part of detecting features and objects in an image.

$$f3 = \text{SobelEdgeDetector}(f2)$$

Complement of Image: The image obtained after applying sobel edge detector is complemented. This is done to get an enhanced image of the haemorrhage identified.

In complement of a binary image, zeros become ones and ones become zeros; black and white are reversed. In the complement of an intensity or RGB image, each pixel value is subtracted from the maximum pixel value supported by the class (or 1.0 for double-precision images) and the difference is used as the pixel value in the output image. In the output image, dark areas become lighter and light areas become darker.

$$f4 = \text{Complement}(f3)$$



VII. RESULT & DISCUSSION

The Sobel method finds edges using the Sobel approximation to the derivative. It returns edges at those points where the gradient of f_2 is maximum; where the gradient of the considered image is maximum. The horizontal and vertical gradient matrices whose dimensions are 3×3 for the Sobel method has been generally used in the edge detection operations. The mask used as shown in fig 3:

-1	0	1
-2	0	2
-1	0	1

Fig.3: Horizontal and Vertical Mask for Sobel Edge Detection.

Image Results:

➤ Input Image



Fig. 4: Retinal Image with Exudates.

Image results:

➤ Output Image

Input Image

Original Image

Exudates superimposed on original image

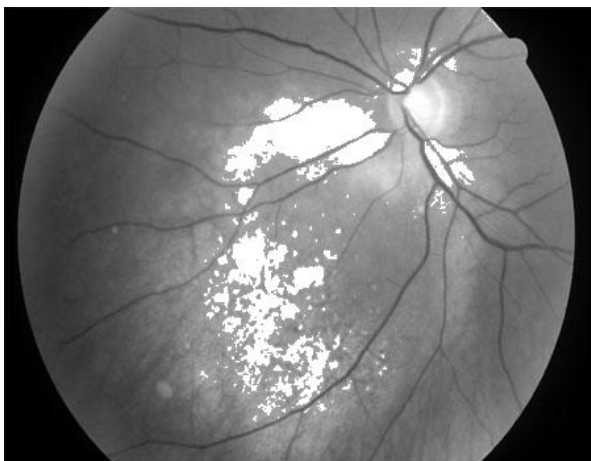


Fig.5: Output Image with Exudates superimposed.

Fig.6: Retinal Image with Micro aneurysms



➤ Image results:

Output Image

Micro aneurysms superimposed on original image



Fig.7: Output Image With Micro aneurysms superimposed.

Input Image

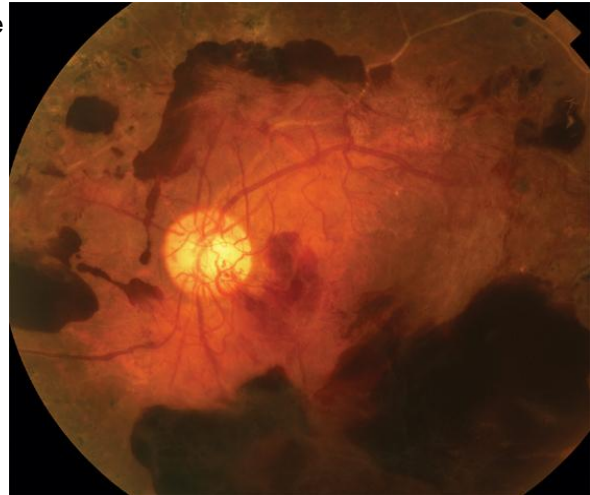


Fig.8: Retinal Image with Haemorrhage blot.

➤ Output Images

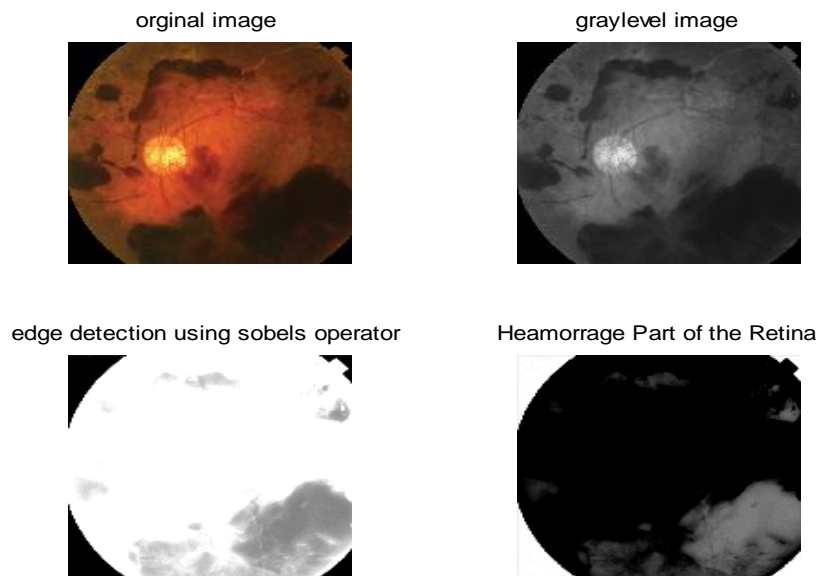


Fig.9; Output Images showing detected Haemorrhage.

VIII. CONCLUSION AND FUTURE WORK

The project proposes efficient methods based on segmentation and enhancement techniques which differentiate between original diabetic image and processed image detected. These may be exploited for the examination of patterns of disease. This may have particular relevance to the management of common ophthalmological disorders such as diabetic retinopathy. The optic disc was particularly reliably detected.

Segmentation is one of the important modules of any image processing technique. The method uses the threshold value to separate the exudates, extended minima to segment the micro aneurysms and sobel edge detector to identify the haemorrhage portion of diabetic retinopathy Human eye. The optic disc was particularly reliably detected. The proposed technique shows significant results.



The main focus of this work is on segmenting the diabetic retinopathy image and making the Exudates, micro aneurysms and haemorrhages clear enough for perception. These methods give almost good results. Exact detection of the condition of a retina whether it is normal or abnormal was determined successfully. Thus Image processing techniques can reduce the work of ophthalmologists and the tools used automatically locate the exudates. Exact detection of the condition of a retina whether it is normal or abnormal was determined successfully.

Future Work: In diabetic retinopathy further algorithms will be required to detect features which indicate risk to the patient's sight such as neovascularisation, cotton wool spots, venous changes, and para-foveal exudation. The detection of other features of retinopathy in diabetes, such as senile macular degeneration, will be facilitated by the removal from the image data set of complex regional features such as the blood vessels. Blood vessels can be detected prior to the detection of pathologies and subtract them from the image.

In diabetes, grading of a patient's retinopathy by fundus imaging and computer analysis, at the site of acquisition of the image, would allow an immediate opinion for the patient on the urgency of referral for an ophthalmological opinion. The deficiency of missing some thin and minute lesions is because of utilizing a simple thresholding method. In retinal images containing severe lesions, the algorithm needs to benefit from a higher level thresholding method or a more proper scheme. Hence, future work is to deal with the problem of the presence of minute lesions in retinal fundus images and also try to localize faint and small exudates.

ACKNOWLEDGMENT

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BIOGRAPHIES



Mr. Ibrahim Patel Working as Associate Professor in BME department, B. V. Raju Institute of Technology, Vishnu, Narsapur, Medak, (Dist), Telangana, India. He has received B. Tech. (ECE) M. Tech Degree in Biomedical Instrumentation and currently pursuing Ph.D at Andhra University. He is having 20 years of teaching and research experience and published 75 research papers in the International conference & journals and His main research interest includes Voice to sign language. I had received "Best Paper Award" and in ICSCI –International Conference in 2011 and I honoured with best paper award from IETE-ACCT-2012 at Naval Science & Technological Laboratory (NSTL) Visakhapatnam (Dist) A. P. and I had received "Best Technical Session Paper Award in 2nd International Conference Biomedical Engg. and Assistive Technologies (BEATS-2012) at National Institute Jalandhar Punjab India.



Mr. V. Sripathi Raja, is currently working as Associate Professor, department of BME, at B. V. Raju Institute of Technology, Narsapur, Telangana, INDIA. He has received B.Tech (Electronics and Communications Engineering) from JNTU Kakinada in 1999, M.Tech (ECE) from Anna University, Chennai, India in 2002. His main research interests are in Biomedical Instrumentation and Signal and Image processing. He has a total experience of 13 years in teaching at various cadres. He has published one journal and three papers in international proceedings. He is a life member of ISTE, BMESI.



Prof. Ashok Shigli, is currently working as Professor and Head, department of BME, at B. V. Raju Institute of Technology, Narsapur, Telangana, INDIA, where he teaches, among other things. He has received B.E (Instrumentation Technology) from Mysore University in 1992, M.Tech (ECE) from IASE University, Sardarshar, Rajasthan, India in 1995 and pursuing his Ph.D (ECE) in Rayalaseema University, Kurnool, Andhra Pradesh state. His main research interests are in Instrumentation and Speech processing. He has a total experience of 21 years in teaching at various cadres. He has published 6 journal and 10 papers in international proceedings. He is a life member of ISTE, MIE and BMESI.