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# A Dimensionality Reduced Iris Recognition System with Aid of AI Techniques N.MURALI KRISHNA<sup>1</sup> and GITAM INSTITUTE OF SCIENCE<sup>2</sup> <sup>1</sup> kite college of professional engineering sciences/JNTUH,Hyderabad *Received: 11 December 2013 Accepted: 4 January 2014 Published: 15 January 2014*

#### 7 Abstract

Technologies that exploit biometrics have the potential for the identification and verification of individuals designed for controlling access to secured areas or materials. One of the 9 biometrics used for the identification is iris. Many techniques have been developed for iris 10 recognition so far. Here we propose a new iris recognition system utilizing unbalanced wavelet 11 packets and FFBNN-ABC. In our proposed system, the eye images obtained from the iris 12 database are preprocessed using the adaptive median filter to remove the noise. After 13 removing the noise, iris part is localized by using contrast adjustment and active contour 14 technique. Then unbalanced wavelet packets coefficients and Modified Multi Text on 15 Histogram (MMTH) features are extracted from the localized iris image. Then MMTH 16 features extracted are clustered by using the MFCM technique. After clustering, the 17 dimensionality of the features is reduced by using PCA. Then the dimensionality reduced 18 features unbalanced wavelet packet coefficients are given to FFBNN to complete the training 19 process. During the training, the parameters of the FFBNN are optimized using ABC 20 Algorithm. The performance of our proposed iris recognition system is validated by using 21 CASIA database and compared with the existing systems. Our proposed iris recognition 22 system is implemented in the working platform of MATLAB.Keywords: feed forward back 23 propagation neural network (FFBNN), adaptive median filter, unbalanced haar wavelet, 24 modified multi text on histogram (MMTH), iris recognition, artificial bee colony algorithm 25 (ABC), principle component analysis (PCA), modified fuzzy c-means (MFCM). 26

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Index terms — feed forward back propagation neural network (FFBNN), adaptive median filter, unbalanced
 haar wavelet, modified multi text on histogram (MMTH), iris r

# 30 1 A Dimensionality Reduced Iris Recognition

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exploit biometrics have the potential for the identification and verification of individuals designed for controlling
access to secured areas or materials. One of the biometrics used for the identification is iris. Many techniques
have been developed for iris recognition so far.

Here we propose a new iris recognition system utilizing unbalanced wavelet packets and FFBNN-ABC. In our proposed system, the eye images obtained from the iris database are preprocessed using the adaptive median filter to remove the noise. After removing the noise, iris part is localized by using contrast adjustment and active contour technique. Then unbalanced wavelet packets coefficients and Modified Multi Text on Histogram (MMTH) features are extracted from the localized iris image. Then MMTH features extracted are clustered by using the MFCM technique. After clustering, the dimensionality of the features is reduced by using PCA. Then

<sup>41</sup> the dimensionality reduced features & unbalanced wavelet packet coefficients are given to FFBNN to complete the

42 training process. During the training, the parameters of the FFBNN are optimized using ABC Algorithm. The 43 performance of our proposed iris recognition system is validated by using CASIA database and compared with

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wavelet, modified multi text on histogram (MMTH), iris recognition, artificial bee colony algorithm (ABC),

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 traditional biometrics such as finger prints and facial features [13]. Also, the probability of finding two people

with identical iris pattern is almost zero [7]. That's why iris recognition technology is becoming an important 49 biometric solution for people identification in access control [14]. More technically, the iris is part of the unveil, 50 or middle, coat of the eye. It is a thin diaphragm stretching across the interior portion of the eye and supported 51 by the lens [4]. Iris recognition is a method of biometric authentication that uses pattern-recognition techniques 52 based on high-resolution images of the irises of an individual's eyes [2]. There are four main techniques in 53 Iris Recognition System Namely: Segmentation, Normalization, Feature Extraction And Matching [12]. Iris 54 recognition begins with finding an iris in an image, demarcating its inner and outer boundaries at the pupil 55 and sclera, detecting the upper and lower eyelid boundaries if they occlude and detecting and excluding any 56 57 superimposed eyelashes or reflections from the cornea or eyeglasses. These processes may collectively be called 58 segmentation [1]. Iris normalization mainly involves two basic operations, one is to detect eye lids and the other 59 is boundary detection. The first step involves extraction of circular shaped iris rim by removing the noisy regions. The second step is to detect the inner and outer boundaries of iris. [5]. The matching module generates a match 60 61 score by comparing the feature sets of two iris images [11].

The great advantage of the authentication using iris recognition is the irreplaceable nature. It has various 62 applications to high-security facilities, but it is now being widespread developed in information systems such 63 as network, e-commerce, and retail applications [3]. Although, a number of iris recognition methods have been 64 proposed, it has been found that several accurate iris recognition algorithms use multiscale techniques, which 65 provide well-suited representation for iris recognition [10]. The main difficulty of human iris recognition is that 66 it is hard to find the apparent feature points in the image and to keep their represent ability high in an efficient 67 way [17]. The data are unique to the individual and remain so throughout one's life [8]. The performance of iris 68 recognition systems highly depends on the Introduction oday, many countries are considering or even announced 69 procurement of bio-metrically enabled national identity (ID) card schemes, one of whose purposes will be to detect 70 71 and prevent multiple IDs [1]. Applications such as passenger control in airports, access control in restricted areas, 72 border control, database access and financial services are some of the examples where the biometric technology has been applied for more reliable identification and verification [6]. Biometric is unique to each individual and 73 is reliable [16]. Iris recognition is the most reliable biometric system available because of iris uniqueness [19], 74 stability, permanency and easily taking [3]. It is based recognition has been gaining popularity in recent years, 75 and it has several advantages compared to other II. 76

#### 77 2 Related Works

Fernando et al. [21] have used a modular neural network architectures as systems for recognizing persons based 78 on the iris biometric measurement of humans. In that system, the human iris database was enhanced with image 79 processing methods, and the coordinates of the center and radius of the iris were obtained to make a cut of the area 80 of interest by removing the noise around the iris. The input to the modular neural networks was the processed 81 82 iris images and the output was the number of the person identified. The integration of the modules was done with 83 a gating network method results demonstrate that the use of the human iris biometric measurement worked with modular artificial neural networks and favorable results of person identification were obtained. Kodituwakku et 84 al. [22] have attempted to develop an algorithm for iris recognition based on Fuzzy logic incorporated with the 85 visible properties of the human iris function. They were considered the visible features of the human iris such as 86 pigment related features, features controlling the size of the pupil, visible rare anomalies and pigment frill. First 87 they extracted the important and essential feature of a human iris image. Secondly, as an AI technique, Fuzzy 88 logic was applied for iris recognition and person identification. The final system was a very successful at a rate 89 of 98.6% accuracy in recognition with small mistakes. 90

Hariprasath et al. [23] have presented an iris recognition system based on Wavelet Packet Analysis. With 91 an adaptive threshold, WPT sub images coefficients were quantized into 1, 0 as iris signature. Those signatures 92 93 presented the local information of different irises. By using wavelet packets, the size of the iris signature of 94 code attained was 1280 bits. The signature of the iris pattern was compared against the stored pattern after 95 computing the signature of iris pattern Identification was performed by computing the hamming distance. The 96 accuracy of the proposed system varied when different feature vector was chosen. Pushpalatha et al. [25] have proposed an iris recognition system with iris localization to segment and recognize cooler iris with highest speed 97 and accuracy. Frequency domain magnitude and phase features were used for image feature representation. For 98 classification process, support vector machines with "winner takes it all" configuration were used. Tests have 99 shown 97% accuracy with average time of 31 milliseconds seconds for classifying each test image. They developed 100 the iris recognition system using C#.Net (.Net 3.5). 101

#### <sup>102</sup> 3 III. Proposed Iris Recognition System using ai Techniques

In the proposed methodology, the given input image is preprocessed using adaptive median filter for removing salt 103 and pepper noise at the first stage. Following that, by adjusting the contrast and applying active contour technique 104 on the preprocessed eye image, iris is localized. Then Unbalanced Wavelet Packet coefficients and MMTH features 105 are extracted from the localized iris image and the extracted features are clustered using MFCM. Following that 106 the dimension of the features are condensed using PCA. The Unbalanced Wavelet Packet coefficients and the 107 dimension reduced MMTH features are given to train FFBNN. While training the parameters of the FFBNN 108 are optimized using ABC. During the testing process the same procedure is done here till the feature extraction 109 process. Then the output obtained from the feature extraction process is given to well-trained FFBNN-ABC to 110 validate whether the given input iris image is recognized or not. The architecture diagram of the proposed Iris 111 Recognition System is shown in Fig. 1. 112

proposed method. Section 3 discusses about the proposed technique. Section 4 shows the experimental result 113 of the proposed technique and section 5 concludes the paper. Naresh Babu et al. [24] have proposed an efficient 114 Fuzzy based Iris Recognition Scheme (FIRS). That scheme has four stages namely Segmentation, Normalization, 115 Feature extraction and classification using fuzzy logic. Hough transforms used for detection of Region of Interest 116 (ROI), and combination of Discrete Wavelet Transform (DWT) and Independent Component Analysis (ICA) 117 was used for feature extraction. Using mean and standard deviation as parameters a fuzzy classifier was used to 118 classify the IRIS images. The results were quite convincing and encouraging. The input eye image is initially 119 changed into grey level format. After that using Adaptive median filter, the grey level eye image is preprocessed 120 to take away salt and pepper noise. The input image may have noises which destroy the good pixels in the 121 image. The noise must be eradicated from the input image in order to attain good precision. We are applying 122 adaptive median filter to salt and pepper noise in our suggested work. It identifies the impulse by calculating the 123 difference between the standard deviation of the pixels inside the filter window and the concerned current pixel. 124 Let the iris database () I contains many eye images and let j i 125

126 x , be one of the grey level images taken from the database. The lower and upper bounds x are ()?? +? 127 = +? = k i k i m k j k i n n m s j i su , ,(2)

### $_{128}$ 4 ()()()

Next by means of these local mean, standard deviation and as well a user defined multiplier upper and lower bounds are computed. j i m j i s , , 1 min ?  $\mu \times$  ? = (6) ( ) ( ) j i m j i s , , 1 max ?  $\mu \times$  + = (7)

The noise candidates only substituted by the median s ws med j i , , in the above adaptive median filter algorithm, while staying behind are unaltered. By means of the above adaptive median filter algorithm the salt and pepper noise is eliminated from the specified input eye image and the preprocessed eye image is indicated as '

- 137 x . This preprocessed eye image ( )'
- 138 x is subsequently subjected to iris localization process.

#### <sup>139</sup> 5 b) Iris Segmentation and Normalization

Iris segmentation is the main part in the process of iris recognition. In order to segment the iris from the eye image, here enhanced iris segmentation technique by considering the adaptive thresholding is utilized. The proposed iris segmentation technique has four phases namely,? Removing Holes ? Pupil Detection ? Iris Detection ? Adaptive Normalization the preprocessed image () '

- 144 x is binarized. The process of removing the hole from the pupil is detailed in the below steps:
- 145 Step 1: Set the threshold value (?) as 0.1.
- 146 Step 2: Obtain the binary image ( 'Bx ).
- 147 Step 3: Take the complement image ( 'Cx ) of the binarized image ( ' Bx ).
- 148 Step 4: Take the binary image (' Bx ) with all zeros and consider it as hole. 0, ' = q where Hx q
- 149 Step 5: Catch a point (po) inside the hole.
- 150 Step 6: Check whether Bx Bx then go to step 7
- Step 7:') '(1 Cx Bx B q q ? Î?"? = ? where Î?" is the structuring element defined as ?????????????? 152 0 1 0 1 1 1 0 1 0 Step 8: If 1 ''? = q q Bx Bx ,then discover the hole filled image ('Hx) where '' Bx Bx Hx 153 k? =
- 154 Step 9: Find the number of connected components (?) from the hole filled image ('Hx).
- 155 Step 10: Increment the threshold value (?) as

#### <sup>156</sup> 6 i. Hole Filling

The eye image has holes in the pupil region which is the darkest region in the eye with nearly circular shape. In order to remove the holes from the pupil, binarized image is obtained by applying adaptive thresholding technique. The range of the threshold value (?) is between 0.1 to 0.5. The binary images are obtained by

adaptive thresholding technique. The maximum pixel value in the preprocessed image ()'

- x is multiplied with the threshold value (?). Then by considering the value obtained after the multiplication, 161 () ( c L md 162 ) by 2. By using the obtained center (CPpl) and the radius (RPpl), pupil (Ppl) is detected. 163 iii. Iris detection For iris identification, the preprocessed iris image is upgraded to have sharp variety at 164 the image limits utilizing histogram evening out. This difference upgraded image is utilized for discovering the 165 external iris range by drawing concentric loops of diverse radii from the understudy focus and the intensities 166 lying over the border of the loop are summed up. Among the applicant iris loops, the loop having most extreme 167 Here, scale based normalization approach [29] is utilized to normalize the iris image ) (I in order to preserve the 168 texture property of the features in the iris region ) (I. In the normalization process, the obtained iris part ) (I 169 is converted into Cartesian space to nonuniform polar space. After that, the points lying on the perimeter of the 170 iris 171
- )) ( ( I P and pupil circle )) ( ( ppl P are obtained. Subsequently, the range of radius between the pupil and
- iris boundaries is obtained and it is mapped to a rectangle by considering the distance between the pupil and iris boundaries [29]. Finally, the obtained normalized iris image (), ..., 1 /, n i n i f X i i = + = ? (8) Where s i '
- 175 ? are random variables with ( ) . 0 = i E ?
- We first give a description of the construction of the UH vectors. Suppose that our domain is indexed by , ,..., 177 1 n i = as is the case in ( ), n b i ? + = The breakpoint n b < 1 , 0
- is to be chosen by the analyst. The positive change in power as for the long ago drawn round is the iris external limit. The sweep of the iris location steps is itemized in the accompanying steps.
- 180 Input: radius of the pupil (RPpl), center of the pupil (CPpl), preprocessed image ()' x
- 181 Output: Radius of the iris (IR)
- 182 Step 1: Obtain the preprocessed image ()' x .
- 183 Step 2: Find the histogram equalized image ()' x HE
- 184 Step 3: Compute the size of the preprocessed image ( )C R x  $\times$  ? '
- Step 4: Calculate the radius of the iris as 5.  $1 \times =$  Rppl IR Step 5: Check whether, 2 R IR ?
- 186, then go to step 6.
- 187 Otherwise go to step 10
- Step Step 7: Find the coordinates ( j i, ) of the image)  $\cos()$  (?  $\times + = IR l Rpp i i) \sin(? \times + = IR Rpp l j)$
- 189 j)'()( x HE IR Sum = +
- 190 Step 8: Increment the angle ()? by 10
- 191 Step 9: If 360?? go to step 7 otherwise go to step 5
- 192 Step 10: Change the intensity over circumference
- 193 Step 11: For i=1 to IR , do the following,1 + ? = i i i

#### <sup>194</sup> 7 S S differnce

- 195 Step 12: Find the maximum change in the intensity
- Step 13: Obtain the radius of the iris ( ) IR way that ( ) The recursion then continues in the same manner for as long as feasible, with each vector k j, ? having at most two "children" vectors  $\ref{eq:second}$  ).a the elements of 1 1 1 1 1
- 199 The inner product between X and n b, 1 + + + ? < = l j s l j e b X b l jmax arg, 1, 1, 1 > 1 = 1 + 1 + 2 = 1 + 2
- 200 where k k l 2, l 2? = again  $H_{1}^{(1)}$  () l : l : CDUUT

 $_{\rm 203}$  , are the true UH coefficients of f which are known and need to be estimated.

#### 204 8 ?

Estimate each k j d , by means of a suitable "universal" shrinkage rule( ) , , , , ? k j k j Y h d = . , , , k j k j k j d Y ? + = A Dimensionality Reduced Iris Recognition System with Aid of AI Techniques © 2014 Global Journals Inc. (US) ( ) ( ). , , , i X DUHT X K J K J k j i ? ? = (12)

where the function h has the property that ( )

# $_{209}$ 9 ) ( 2 V H

210 ) To remove the attributes from the images, MTH (Liu, et al., 2010) [26] is a dominant device which extorts the 211 feature from the iris image by combining the benefits of co-occurrence matrix and histogram. Besides with these 212 benefits, mean and variance measures are applied to develop the feature extraction process.

- By using the sobel operator on the iris image along both the horizontal and the vertical directions, the gradient
- 214 images () The MMTH feature extraction process consists of following three steps:y gx x gx ', '? Computing
- Original Image Feature ( ) ( 1 V H ) ? Computing Orientation Image Feature ( ) ( 2 V H ) ? Modified Histogram Features ( ) (V H ) a. Computing Original Image Feature ( ) ( 1 V H )
- Initially, the unique iris image is fragmented in to a number of grids where the grid may have the size of 3x3,
- $_{218}$   $\,$  5x5 and so on. Subsequently for every grid, mean ( m ) and variance ( v ) are computed and by means of those

calculated mean ( mean ) and variance ( var ), threshold value ( v t ) is calculated.v t = { var var, ? + mean 220 mean (16)

Then for each grid, the center pixel value is compared with the threshold value (t v )??????? + = = otherwise Unchanged i i mean t i i mean t mean i C v v p , ) var() (() var() ( $\{(, ), ((17))\}$ 

But the center pixel value lies in between the threshold value (v t), it is substituted with the mean value of the grid or else not as shown in equiv. (11). The grids are partly covered and this process is used for all the grids. The histogram vector (

# $_{ m 226}$ 10 ) ( 1 V H

) is attained after completing the interchanging process, by finding the frequency of grids (not pixels) based on
every grey levels only from the recognized areas. b. Computing Orientation Image Feature () c After obtaining
the orientation image i.e. the gradient image as mentioned above using equiv. (??) & (9), the same process done
for the original image as explained in section 3.2.1. is repeated for the orientation image. Finally, the histogram
vector is obtained, denoted as) (2 V H

only from the identified regions.

233 []?? = = ? ? = N r c j j i ij m c x O 1 1 2 ) () 1 ( $\mu$ ? (18)

234 Where, m is any real number greater than 1, u ij is the degree of membership of x i in the cluster j, x i is the ith of d-dimensional measured data, c j is the d- To attain the cluster, the resultant MMTH features are 235 subsequently passed to the MFCM. Fuzzy c-means (FCM) is a technique of clustering which permits one piece 236 237 of data to belong to two or more clusters. This technique is often applied in pattern recognition. To develop the clustering result adapted FCM is applied based on minimization of the objective function specified in equiv. (12): 238 In our suggested method, the texture attributes computed are clustered in to 2 clusters by means of MFCM.? =239 240 (20) This iteration will end when { } ?  $\mu$   $\mu$  < ? + k ij k ij ij 1 max 241

, where ? a termination criterion between 0 and 1, whilek is the iteration step. This process unites to a local minimum or a saddle point of O. The collected attributes are subsequently passed to the next process that is dimensionality reduction.

#### $_{245}$ 11 ) (V H )

The determined vectors achieved such as () ( 1 V H ) and ( ) ( 2 V H

247 ) are concatenated to acquire the MMTH feature (

# <sup>248</sup> 12 ) (V H

). The attained MMTH features are subsequently focused to clustering process. assesses a novel set of variables, 249 called principal components. Each principal component is a linear mixture of the real values. The entire principal 250 components are orthogonal to each other, so there is no unnecessary information. The principal components as 251 a total form an orthogonal basis for the space of the information. Principal component analysis is a changeable 252 reduction process. It is constructive when you have attained data on a number of variables and consider that 253 there is some idleness in those variables. In this case, redundancy represents that some of the variables are linked 254 with one another, probably because they are measuring the similar construct. As of this redundancy, you consider 255 that it should be probable to decrease the observed variables into a smaller number of principal components that 256 will report for most of the variance in the examined variables. For analyzing information, PCA is a dominant 257 device. This will obtain you through the steps you required to execute a Principle Components Analysis on a set 258 of data. At this point, attributes in each cluster are decreased and as a result the reduced cluster features are 259 employed for additional process. 260

Thus the dimension reduced features are then passed in to FFBNN to continue the recognition process. The obtained feature vector has the length of 6.

#### <sup>263</sup> 13 f) Recognition

Feed Forward neural Network (FFBNN) is applied to identify the iris. In the training phase, uneven wavelet coefficients and the dimension reduced features are specified as the input to the FFBNN. Using these texture features, the neural network is well educated in order to identify the iris. The neural network contains n number of input units, h hidden units and one output unit. The structure of the FFBNN is specified as below:

268 () ()? = + + + + = H n tn tw f w ep w bp w sp w t x 1 1 ..., ?(21)

269 ()() t x e a x ? + = 1 1 (22)

Er is the FFBNN network output, n n A and D are the preferred and actual outputs and h is the total number of neurons in the unseen layer.

#### <sup>272</sup> 14 In bias function

273 t<br/>n sp , t<br/>n bp , tn ep , 1 t f , 2 t f ... tn f

are the uneven coefficients such as starting point, break point, ending point and features attained after dimension reduction correspondingly. The activation function for the output layer is specified in Eq. (16).

? Get the learning error. ii. Optimization of FFBNN parameters by ABC Now we are applying the ABC
algorithm for optimizing the parameters of FFBNN while training to acquire competent iris recognition result.
ABC algorithm is a swarm based meta-heuristic algorithm which was motivated by the sharp foraging behavior

279 of the honey bees. It contains three components namely, employed bees, onlooker bees and scout bees. The

employed bees are combined with the food sources in the region of the hive and they shift the data to the onlookers about the nectar quality of the food sources they are utilizing. Onlooker bees are looking the dance of

- the employed bees within the hive to pick one food source to use according to the data offered by the employed
- bees. The employed bees whose food source is discarded turn into Scout and look for novel food source randomly.
- The number of food sources indicates the location of probable solutions of optimization problem and the nectar
- amount of a food source represents the quality of the solution. The FFBNN parameters (). This generation process is called as initialization process. The fitness value of the produced food sources is computed by equation
- (??5) to assess the best food source.

#### <sup>288</sup> 15 b. Employed Bee Phase

Using the beneath equation, novel population parameters are produced in the employed bee phase, ( )j k j i ij j i j i x x x V , , , ? + = ?(27)

? Compute BP mistake for every node and revise the weights as follows:) () () ( tn tn tn w w w ? + = (24)292 ) (tn w ? is attained as, () B t x w n tn . .) (? = ? (25)

Where ? is the learning rate, which usually ranges from 0.2 to 0.5, and Be is the Back Propagation fault.

294 ? Next do again the steps (2)

Using ABC, the FFBNN parameters () for the onlooker bees from the solutions (j i x,) based on the probability value (j P). After that the fitness function is computed for the novel solution. In order to choose the best parameter, use the greedy selection process later.

### <sup>298</sup> 16 d. Scout Bee Phase

Find out the abandoned parameters for the scout bees. If any abandoned parameter is present, after that substitute that with the novel parameters found out by scouts by means of the equation (??8) and assesses the fitness value. After that memorize the best parameters accomplished so far. Afterward the iteration is increased and the process is prolonged till the stopping criterion is arrived.

## 303 17 IV.

# 304 18 Experimental Results

Our proposed iris recognition system with FFBNN-ABC is implemented in the working platform of MATLAB (version 7.13).Our proposed iris recognition

#### 307 **19 J**

Where, k and j is an arbitrary chosen index, ? is randomly produced number in the range [-1, 1] and j i V, is the novel value of the th j position. Next the fitness value is calculated for every novel generated population parameters of food sources. From the calculated fitness value of the population, best population parameter is chosen i.e. the population parameter, which has the highest fitness value by using greedy selection process. Probability of the chosen parameter is calculated by the equation (??2) after choosing the best population parameter.

system is the combination of FFBNN and ABC. In order to reduce the computation complexity and get higher 314 performance, the dimensionality of features is reduced with the help of the well-known optimization algorithm 315 PCA. Then the dimensionality reduced features are given to the FFBNN to achieve the training process. So as 316 to get more accuracy in the process of recognition, the FFBNN parameters are optimized using ABC algorithm. 317 In the testing process, more data are given to the well trained FFBNN-ABC to validate the performance of the 318 proposed technique. The performance of the proposed iris recognition system is evaluated using CASIA database 319 and the proposed technique's performance is compared with the existing iris recognition systems given in [21], 320 [23] and [24]. a) Performance Analysis By applying the statistical measures which is specified in [27], the concert 321 322 of our suggested iris recognition system is examined. We employ CASIA iris thousand -NG database which has 323 788 number of iris images to complete the performance analysis process. For one dataset, our proposed technique 324 takes 0.3225 seconds for training and 0.0054 seconds for testing. Totally our database consists of 51 dataset. 325 The performance of the proposed technique is compared with other classifiers such as FFBN, FFBN\_GA, Fuzzy, ANFIS&KNN and the corresponding statistical measures are given in Table 1(i). Then the performance of the 326 proposed technique is analyzed by using Unbalanced Haar Wavelet and it is compared with other wavelets such 327 as Haar, Coif let, Symlet & Bi-orthogonal wavelet and the corresponding statistical measures are given in Table 328 1(ii). Also our suggested iris recognition system performance is assessed and compared with the conventional 329 iris recognition system given in [21], [23] & [24] and the corresponding statistical measures are given in Table 330

1(iii). Figure ??, 4 and 5illustrate the sample of iris images, preprocessed images and iris segmented images 331 correspondingly. In Table ??1(i) and Figure 6.(i), the performance of the proposed technique is compared with 332 various classifiers such as FFBNN, FFBNN-GA, Fuzzy, ANFIS and KNN. By seeing both table and graph, 333 we can say that the proposed technique yields higher rate of accuracy than the proposed technique. From 334 the measurement of the accuracy, we can say that our proposed technique recognize the iris images effectively. 335 In addition to that, the sensitivity and specificity are the two measurements which can provide the additional 336 details about the performance of a technique. On looking at the sensitivity and specificity measures, our proposed 337 technique has given better rate than the other classifiers. In specificity measure, our proposed technique is yielded 338 100% specificity. Also, when looking at the other measurements such as FPR and FDR, the proposed technique 339 obtained 0% FPR and In Table ??1 (ii) and Figure (ii), the performance of the proposed technique is compared by 340 changing wavelets such as Haar, Coif let, Symlet and Bi-Orthogonal. In our proposed technique, Unbalanced Haar 341 Wavelet is utilized. On looking at both table and graph, we can say that the proposed technique yields higher 342 rate of accuracy, sensitivity and specificity when compared to the other wavelet techniques. All the performance 343 measures are showed that our proposed technique recognize the iris images efficiently. Similarly, the performance 344 of the proposed technique is compared with the existing techniques such as [21], [23] and [24] and it is given in 345 Table1.(iii) and figure 6 system in iris recognition process presents an incredible rate of accuracy (98.8317757), 346 347 sensitivity, (98.69451697), specificity (100), FAR (0) and FRR (1.305483029). The high value of these measures 348 illustrates that our suggested technique more precisely identifies the iris images from the specified test images. 349 Based on FFBNN-ABC, the comparison result illustrates that our suggested iris recognition system has specified high accuracy than existing methods. Hence our suggested iris recognition system competently identifies the iris 350 imaged by applying the FFBNN and ABC techniques. 351

Discussion: Comparison of the performance of the proposed technique with the other techniques in terms of 352 FAR and FRR. 353

In Figure ??7, the performance of the proposed technique is compared with other techniques in terms of FAR 354 and FRR. Our proposed technique has less FRR rate when compared to the other techniques. While seeing the 355 value of FAR, our proposed technique offers 0% of FAR. It adds additional strength to our proposed technique 356 in its performance. Thus our proposed technique proved its efficiency in the recognition of iris. 357 V.

#### 20Conclusion 359

We have suggested an iris recognition system based on FFBNN and ABC at this point. The suggested system 360 was executed and CASIA iris thousand -NG database is employed to examine the results of the suggested iris 361

recognition system. The presentation study confirmed that the suggested iris recognition J e XIV Issue IV Version 362 I



Figure 1: Figure 1 :

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358

<sup>&</sup>lt;sup>1</sup>Year 2014 J

 $<sup>^{2}</sup>$ © 2014 Global Journals Inc. (US) (MMTH). After that, gradient map ( )

<sup>&</sup>lt;sup>3</sup>© 2014 Global Journals Inc. (US)

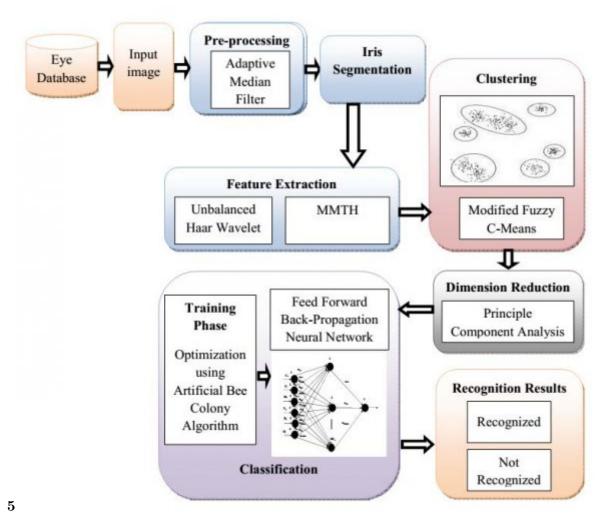


Figure 2: 5 .

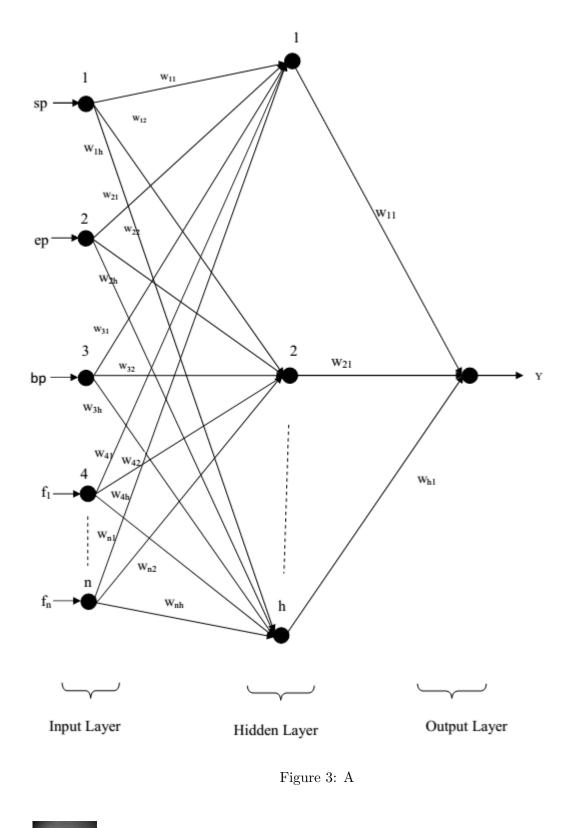




Figure 4: 12 :

13			
		Figure 5: Step 13 :	
		Figure 6:	
		Figure 7: A	
6			
		Figure 8: 6 :	
0			
		Figure 9: ?	
0			
		Figure 10:	
31			
	<b>T</b> . <b>1 1</b>		

Figure 11: the range of **b** is such that assumption 3 1 n



Figure 12:

0		
	Figure 13:	
	Figure 14:	
0	0	
	Figure 15:	
0		
	Figure 16:	
2	Figure 17: Figure 2 :	
0		

Figure 18:

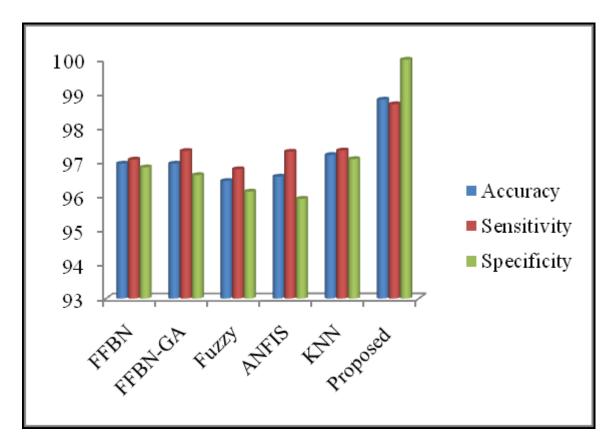


Figure 19:

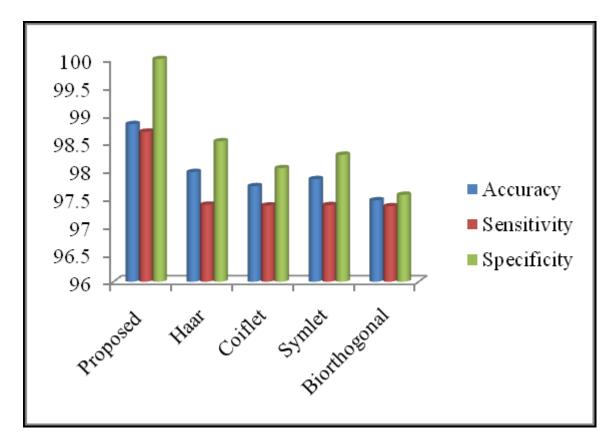


Figure 20: P

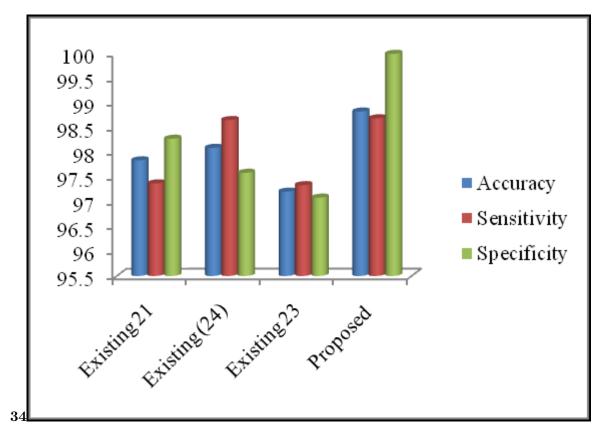


Figure 21: Figure 3 : Figure 4 :

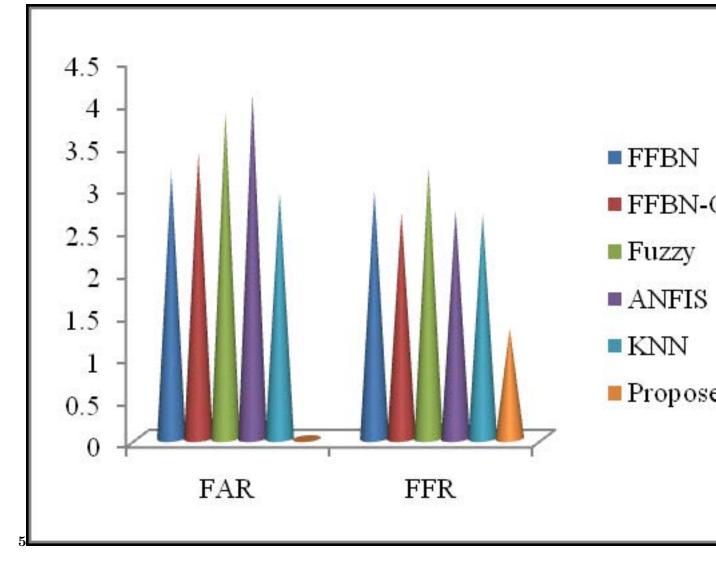


Figure 22: Figure 5 :

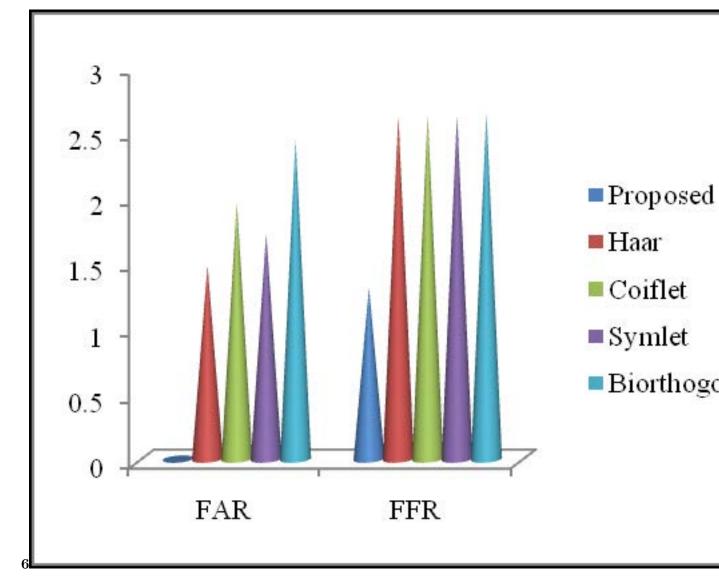


Figure 23: Figure 6 :

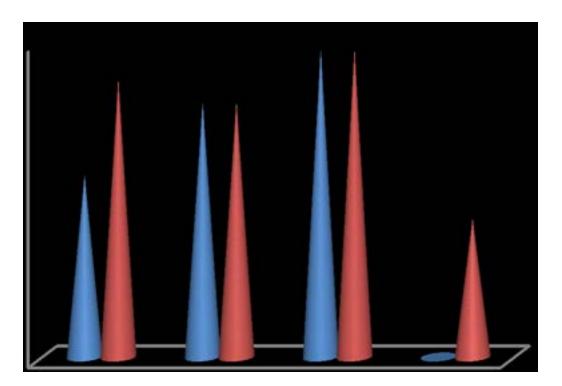


Figure 24: FDR

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Global	Measur	esProposed	l FFBNN	FFBNN-	FUZZY	ANFIS	KNN
Journal	Accu-	Tech-	96.95431472	$\mathbf{GA}$	96.44670051	96.57360406	97.20812183
of Re-	racy	nique	97.07446809	96.95431472	96.79144385	97.30458221	97.34042553
searches	Sensi-	98.83177	596.84466019	97.32620321	96.1352657	95.92326139	97.08737864
in Engi-	tivity	98.69451	6 <b>9</b> 7155339806	96.61835749	3.8647343	4.076738609	2.912621359
neering	Speci-	100 0	96.56084656	3.381642512	95.76719577	95.5026455	96.82539683
	ficity	100		96.2962963			
	$\operatorname{FPR}$						
	PPV						
	NPV	90	97.31707317	97.56097561	97.07317073	97.56097561	97.56097561
	FDR	0	3.439153439	3.703703704	4.232804233	4.497354497	3.174603175
	MCC	94.24705	0 <b>9</b> 3.89852174	93.90090616	92.88352799	93.14569616	94.40708603
	FAR	0	3.155339806	3.381642512	3.8647343	4.076738609	2.912621359
	FRR	1.305483	0 <b>2</b> 9925531915	2.673796791	3.20855615	2.69541779	2.659574468
				(i)			

[Note: Performance measures of Proposed FFBNN-ABC-PCA technique with other (i) other classifiers (ii) other wavelets (iii) existing techniques (i)]

Figure 25: Table 1 :

#### 20 CONCLUSION

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