



# An ICA Based Feature Level Fusion of Iris and Ear Biometrics

<sup>1</sup>Chandini Sharma

Student

M.Tech Scholar, C.S.E. Department  
Adesh Institute of Engineering & Technology  
Faridkot, Punjab, India  
[chandinisharma20@gmail.com](mailto:chandinisharma20@gmail.com)

<sup>2</sup>Navdeep Singh Sethi

Assistant Professor

C.S.E. Department  
Adesh Institute of Engineering & Technology  
Faridkot, Punjab, India  
[navdeepsethi@gmail.com](mailto:navdeepsethi@gmail.com)

**Abstract:** In this paper, a new multimodal biometric system is introduced; this system aim to fuse iris and ear features. A new technique i.e. Independent Component Analysis (ICA) is implemented at feature level for feature extraction and fusion of iris and ear increase the accuracy of the authentication system. Additionally the feature reduction of iris and ear has been done by Genetic Algorithm (GA) and training of the system is done by Neural Network (NN). The performance evaluation of proposed method is done by using metrics like False Acceptance Rate (FAR), False Rejection Rate (FRR) and Accuracy in MATLAB environment.

**Keywords:** Independent Component Analysis (ICA), neural network (NN), Biometric Fusion, Iris, Ear.

## 1. INTRODUCTION

Bio metric fusion is always done to enhance the security levels of the access in the modern world. Uni-modal biometric systems do not provide sufficient levels of accuracy and security and are not to prevent the attacks against data modification and authentication [1]. There has only been limited research done into the application of multimodal biometric systems. Most studies have only made use of small population size for their biometric samples [2]. This cannot be accurate enough, as a larger population size might produce a very different result.

A lot of previous work has already been done in the contrast of combining different biometric techniques which is termed as fusion of the biometrics [3]. The biggest problem in any fusion technique is the matching of the features after the extraction for better alignment of the feature value. There are a lot of algorithms which are available for the feature extraction but there is no such method mentioned through which we can align them all together [4]. To make the alignment of the features better optimization techniques could have been used. Previous research workers have put their effort in the combining different biometric techniques using

Optimization Techniques like BCO, ACO [5]. Neural Network is one of most advanced classifiers of this arena. Use of neural network would enhance the security system to a large extent [6].

Remaining paper is organized as Section II includes the proposed work methodology, Section III contains the results and implementation and Section IV contains the conclusion and future scope.

## 2. WORK MODEL

The below figure shows the various steps that are being followed to get the biometric fusion of ear and ear and this will address two critical issues in the design of a multibiometric system, namely, fusion methodology and matching methodology [7].

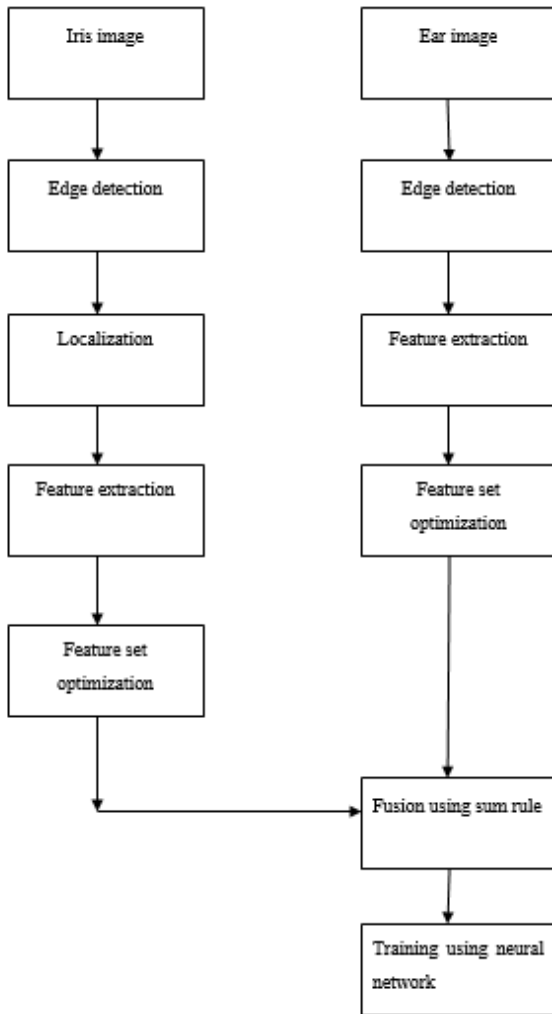


Figure.1 proposed work model

## A. IRIS RECOGNITION

### Step.1 Image acquisition

Image acquisition of iris biometric. Image acquisition of iris image is the first step of proposed technique which is collected from IIT Delhi Iris Image Database version 1.0.



Figure 3. Iris Biometrics

### Step.2 Gray scale conversion

Intensity image  $I$  is given as:

$$I = R(.23) + G(.50) + B(.31)$$

Image  $I$  is then preprocessed with ICA.  $I$  is actually the hue- Saturation-Value color space. During conversion Red-Green-Blue color space is converted into HSV color space.

### Step.3 Canny edge detection

Canny edge detector is applied to get the edges of the iris image. Edge detection is mainly done to get the useful data and to remove the extra part of ear biometric from this the edge gradient and direction can be determined:

$$G = \sqrt{|G_x| + |G_y|}$$

The edge detector returns a value for the first derivative in the horizontal direction ( $G_x$ ) and the vertical direction ( $G_y$ ).

### Step.4 Iris localization using HCT

Iris localization is done mainly to get the pupil and outer part of iris biometric [6]. The first step of iris localization is to get the pupil and then the outer part of iris using following formula:

$$G_{abs} = G_{vertical} + G_{horizontal}$$

where  $G_{vertical}$  is the convolution of image with  $C_{vertical}$  and  $G_{horizontal}$  is the convolution of image with  $C_{horizontal}$ .

### Step.5 Feature extraction using ICA

Then feature extraction will takes place. Independent Component Analysis (ICA) is a computational method to get hidden values of random variables. ICA basically designed for multivariate data. ICA is somewhat related to Principal Component Analysis (PCA). But it is capable when PCA fails. ICA will helps to find features of independent components [8].

**Input:**  $C1$  (contrast function),

$Z1$  (whitened mixtures),

$\beta 1$ (convergence parameter),

$\tau 1$ (number of iterations).

**Output:**  $W1$  (separation matrix).

**Auxiliary:**  $\alpha 1$ (an angle),  $n1$  (number of sources),  $i1, j1, t1$  (iteration indices).

Begin . Initialize  $W1$  to the identity matrix.

$W1 \leftarrow In1$  .

Deflation approach: estimate each source sequentially.

for  $i1 \leftarrow 1$  to  $n1$  do .

Iterate for the  $i$ -th source

for  $t \leftarrow 1$  to  $\tau 1$  do .

Set the current angle variation.

$\alpha 1 \leftarrow \pi \beta t$  .

For each perpendicular direction.

for  $j1 \leftarrow i+1$  to  $n1$  do .

Determine best contrast value.

if  $C1(wi\uparrow jz) > C1(wiz)$  and  $C1(wi\uparrow jz) > C1(wi\downarrow jz)$  then .

Rotate the  $i$ -th and  $j$ -th rows of  $W$  accordingly  $(+\alpha)$ .  
 $wi, wj \leftarrow wi\uparrow j, wj\downarrow i$  else if  $C1(wi\downarrow jz) > C1(wiz)$  and  $C1(wi\downarrow jz) > C1(wi\uparrow jz)$  then .

Rotate the  $i$ -th and  $j$ -th rows of  $W$  accordingly  $(-\alpha)$ .

$wi, wj \leftarrow wi\downarrow j, wj\uparrow i$

end if

end for

end for

end for

Return  $W$

End

### Step.6 Iris Optimization using Genetic Algorithm

Initialize generation 0:

$k := 0$ ;

$P_k :=$  a population of  $n$  randomly-generated individuals;

// Evaluate  $P_k$ :

    Compute fitness ( $i$ )

for each  $i \in P_k$ ; do

{ // Create generation  $k + 1$ :

// 1. Copy: Select  $(1 - \chi) \times n$  members of  $P_k$  and insert into  $P_{k+1}$ ;

// 2. Crossover: Select  $\chi \times n$  members of  $P_k$ ; pair them up; produce offspring; insert the offspring into  $P_{k+1}$ ;

// 3. Mutate: Select  $\mu \times n$  members of  $P_{k+1}$ ;

invert a randomly-selected bit in each;

// Evaluate  $P_{k+1}$ :

    Compute fitness( $i$ ) for each  $i \in P_{k+1}$ ;

// Increment:  $k := k + 1$ ; } while fitness of fittest individual in  $P_k$  is not high enough;

return the fittest individual from  $P_k$

## B. EAR RECOGNITION

### Step.7: Image Acquisition

Image acquisition of ear biometric. Image acquisition of ear image is the first step of proposed technique which is collected from IIT Delhi Ear Image Database version 1.0.

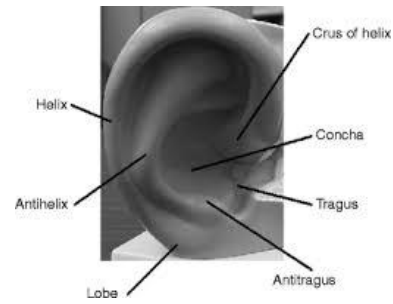


Figure: 2 Ear Biometric

### Step.8: Gray scale conversion

Repeat step.2

### Step.9: Canny edge detection

Repeat step.3

### Step.10: Feature extraction of Ear using ICA

Repeat step.5

### Step.11 Optimization of biometric feature set using Genetic algorithm [5].

Repeat step.6

## C. FUSION OF IRIS AND EAR BIOMETRICS

### Step.12 APPLY SUM RULE

Based on the idea of sum rule based score level fusion, the generated matching scores of iris and ear data are summed up together to form a final fused score using sum rule and it can be described as [7]:

$$F_{\text{score}} = F_{\text{iris}} + F_{\text{ear}}$$

### Step.13 APPLY NEURAL NETWORK

Training using neural network will be done in following way:

setup NN and also prepare the subsequent factors as:  
number\_of\_layers; epochs;

learning\_rate; permissible\_error;

**input:** network, training set

**do**

**for** each image in training set

    fuse the removed characteristics hooked on to a single characteristics matrix;

**until** a solitary characteristic of vector matrix is made;

**do**

    train the given network regarding class labels as well as feature vectors;

**until** ending criterion epochs=2000 is fulfilled

**output:** a trained neural network.

### Step.14

Evaluation of Performance matrices [8,9].

### 3. RESULTS AND IMPLEMENTATION

The system is tested on iris/ear images obtained at IIT Delhi database using MATLAB environment. The below figures shows the implementation results.

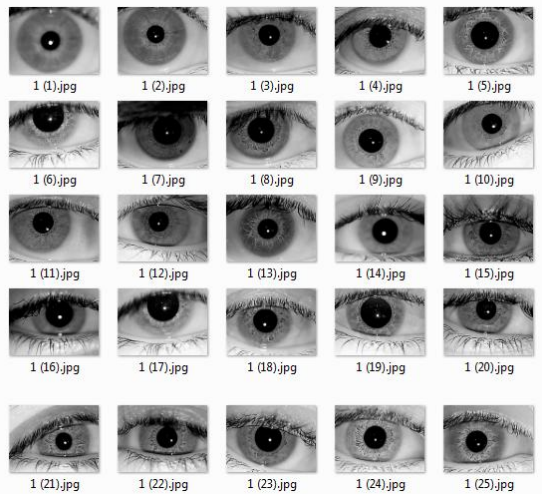


Figure.4 Iris Database

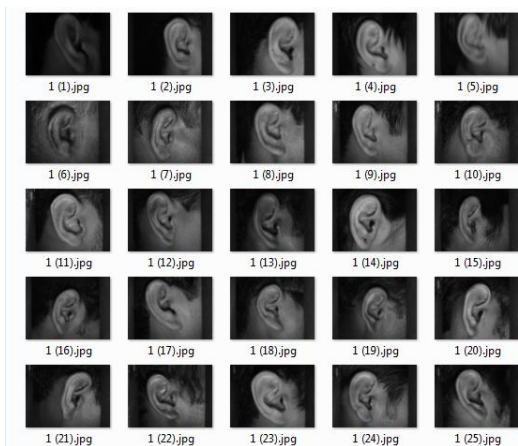


Figure.5 Ear database

#### • Calculate FAR

$$FAR = \frac{(\text{Total Number of Samples} - \text{Number of Samples that falsely accepted})}{\text{Total Number of Samples}}$$

#### • Calculate FRR

$$FRR = \frac{(\text{Total Number of Samples} - \text{Number of Samples that Falsely Rejected})}{\text{Total Number of Samples}}$$

#### • Calculate Accuracy 100– (FAR+FRR) %

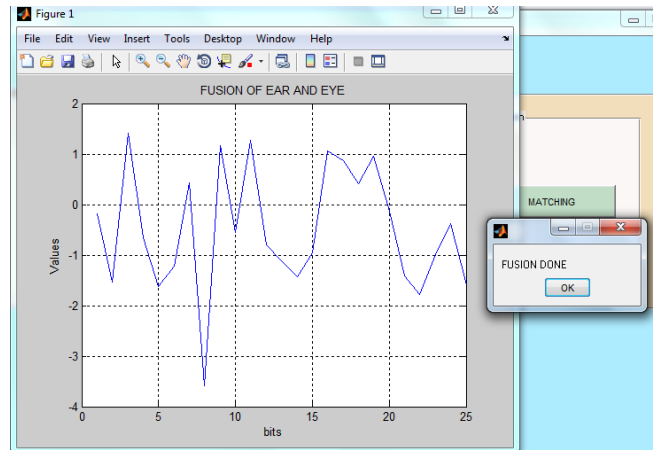


Figure.6 Iris and Ear Biometric Fusion

Above graph shows the fusion of iris and Ear using proposed methodology that has been followed and for 1 image taken from each of iris and ear dataset will led to following parameter values.

The below figure shows the FAR, FRR and Accuracy value for proposed selected image of iris and ear. It has been shown that FAR= .88, FRR= .99 and Accuracy= 99.38 has been achieved.

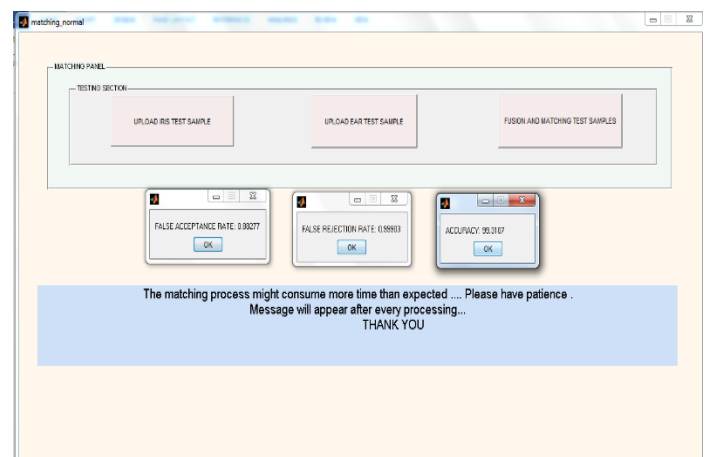


Figure.7 Iris and Ear Biometric Fusion parameter values

### 4. CONCLUSFUTURE SCOPE

In this paper, a new technique is generated at feature level for feature extraction and fusion of iris and ear verification system to increase the accuracy of the authentication systems. In this ICA features are extracted for iris and ear. This proposed method decreased the FAR as well as FRR, & has increases the system performance on the given data set. This system has been measured up to with the other bimodal system where score level fusion is done using the same modalities.

Future works could go in the direction of using more robust modeling techniques against forgeries and hybrid fusion level can be used. Multi-modal modalities can be used together to make forgeries more difficult. Also, the system should be tested on a larger database to validate the robustness of the model.

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