

# USING UNIMODAL FRAMEWORK WITH BIOMETRIC CLASSIFICATION TO ENHANCE MULTI-MODEL SECURITY SYSTEM

Hardik Chaudhary

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## ABSTRACT

The usage of Biometric devices nowadays is very common in every field. It helps to identify the persons by their behavioural or physiological features. Multi biometric system has their advantage due to high accuracy of matching, increasing population coverage, deleting spoofing attacks and imparting fault tolerance to biometric applications. The unimodal framework depends on the proof of a solitary wellspring of data while multi-biometric frameworks if merge various wellsprings of biometric proof. The reconciliation of proof is known as a combination. In a multi biometric framework, the wellspring of biometric data utilized different biometric characteristics that can be discussed in this paper.

## I. INTRODUCTION

A Biometric framework can either is utilized for recognizable proof or check of a person. Customarily, verification strategies utilizing passwords (information-based security) and ID cards (ownership-based security) have been utilized to confine access to applications. However, these frameworks are defenceless against assaulted and security can be broken. As per Satyavarapu et al. breach of biometric, frameworks can be commonly isolated into certain classifications. There is a breach at the UI, a breach at the Interfaces between modules, assaults on the breach of the module on the format database. Biometric frameworks allude to the programmed acknowledgment of people dependent on their physiological and social attributes. Physiological attributes (unique finger impression, iris, hand gestures, face, DNA, and so on.) use estimations from the human body. Social attributes (signature, keystroke, voice, and so on.) utilize dynamic estimations dependent on human activities [1]. These are uni-biometric which depend on the proof of a solitary wellspring of data for confirmation, which needs to keep up with an assortment of issues, for example, (clamor in detected information, between class similitudes, and infraclass varieties, and so forth).

It happens that a solitary biometric isn't adequate to meet the assortment of prerequisites portrayed by a few enormous scope validation frameworks conceivable answer for make up for the bogus order issue due to between class similitudes and infraclass varieties that can be found in the combination of biometric frameworks or specialists [2]. This alludes to Multi biometric.

1. **Universality.** No two persons has same traits
2. **Distinctness.** Two persons must have different features
3. **Performance.** FAR/FRR rates must be low.
4. **Collectability.** Biometrics can be quantitatively measured.
5. **Acceptability.** Acceptability of biometrics by user.
6. **Resistant.** Avoidance of fraud.

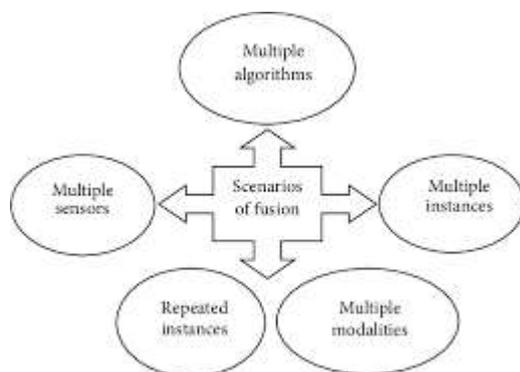


FIGURE 1: Fusion scenarios of multimodal systems.

Fig.1: Recognition system

This framework which wires data from numerous biometric sources can be ordered into various classes: Multi-sensor frameworks, Multi-modular frameworks, Multi-test frameworks, Multi-occasion frameworks, Multi-calculation frameworks. Relies upon the degree of data that is combined, the combination plans can be delegated the levels are sensor level, include level, score level, and choice level combination. There is wide assortment of utilizations through a biometric framework with various degrees of security is alluring. Right now, productive biometric security utilizing multi biometric has been proposed to guarantee the distinctive degree of security.

- **Fusion of modalities must take place in synchronous manner.**
- **Fast processing time fusion strategy must have been adopted.**
- **Modalities are independent to each other.**
- **Different confidence level: like to recognize the crying voice is much easier in video than in audio.**
- **The cost may be incurred in units of time, money or other units of measure.**

There are number of ways of fusion as mentioned below:

- 1) **Fusion prior to matching**
- 2) **Fusion after matching [4]**

## II. IRIS RECOGNITION

Recently, many researchers have been focusing on iris recognition. As a fingerprint is unique, similarly iris is also unique for every single person, which helps in the identification of persons. To detect the iris pattern we localize boundaries of inner and outer part of the iris which is shown in figure 3. The segmented iris is disclosed into polar form in the normalization process after that features are analysed after extraction. Above all processes, localization plays an important role. Miss [5] detection of inner and outer boundaries sometimes causes inadequate segment that causes failure in further research. CHT is the most common algorithm which is used and proved as the best algorithm in iris detection.

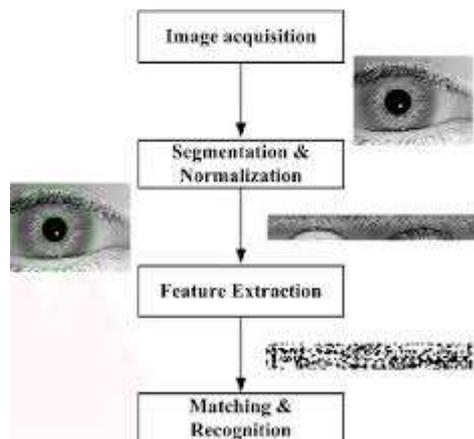


Fig.2 Human iris

### III. FACIAL RECOGNITION

The human face assumes a critical job in our social [6] interface, passing on individuals' character. With the human face as a key to safety, biometric face acknowledgment aptitude has gotten huge consideration in the previous hardly any years because of its feasible for a wide inconstancy of utilizations in both law authorization and non-law requirement. As compared with different biometrics plots by unique finger impression/palm print and iris, face acknowledgment has unmistakable advantages on account of its non-contact method. Face pictures can be gotten from a separation without contacting the individual being recognized, and the documentation doesn't require interrelating with the individual. Likewise, face acknowledgment serves the wrongdoing cautioning assurance since face pictures that have been checked and documented can later assistance recognize an individual. Most present facial gratefulness frameworks effort with numeric codes called face prints. Such plans distinguish 80 nodal realities on a human face. Right now, focuses are endpoints used to gauge factors of an individual's face, for example, the chilliness or width of the nose, the multifaceted nature of the eye attachments and the state of the cheekbones. These structures work by catching data for nodal focuses on an advanced picture of a substance's face and putting away the subsequent data as a face print. The face example would then be able to be utilized as a reason for appraisal with information caught from faces in an image or video. Facial acknowledgment structures dependent on face prints can quickly and precisely recognize target people when the conditions are ideal. [7].



Fig 3: face recognize

### IV. FEATURE EXTRACTION

SIFT algorithm for feature extraction for iris and face for biometric has been used in this section. Unique properties have been used. The SIFT approach [8], for picture feature time, snaps a photo and changes it into a "broad social affair of neighbourhood highpoint vectors" (From "Article Recognition from Local Scale-Invariant Features", David G. Lowe). Every one of these feature vectors is invariant to any scaling, upheaval or understanding of the

image [9]. This system offers various features with neuron responses in primate vision. To help the extraction of these features the SIFT computation applies a 4 phase isolating strategy: [10]

- Detection of Scale-Space Extreme

This period of the isolating undertakings to perceive those territories and scales those are recognizable from different points of view of a similar thing. This can be capably cultivated using a "scale-space" capacity. Further, it has been built up underneath reasonable suppositions it must be considering the Gaussian limit.

- Key point Localisation

This stage tries to take out more spotlights on the summary of key focuses by finding those that have a low distinction or are deficiently limited on an edge. This is practiced by processing the Laplacian, acknowledge math domain. [wolfram.com/Laplacian.html](http://wolfram.com/Laplacian.html) for unpretentious components, regard for each key point found in organizing 1.

- Introduction Assignment

This progression expects to consign an anticipated prologue to the key focuses considering close by picture properties. The key point descriptor, delineated underneath, would then be able to be addressed as for this presentation, accomplishing invariance to rotate.

- Key point Descriptor

The close-by tendency data, used above, is furthermore used to make key point descriptors. The incline information is turned to agree with the presentation of the key point and after that weighted by a Gaussian with a change of  $1.5 * \text{key point scale}$ . This data is then used to make a course of action of histograms over a window concentrated on the key point.

## V. PROPOSED WORK

Step 1: First, search the dataset for face and iris recognition from the UCI AI site. Select the 15 pictures in the specific database [15].

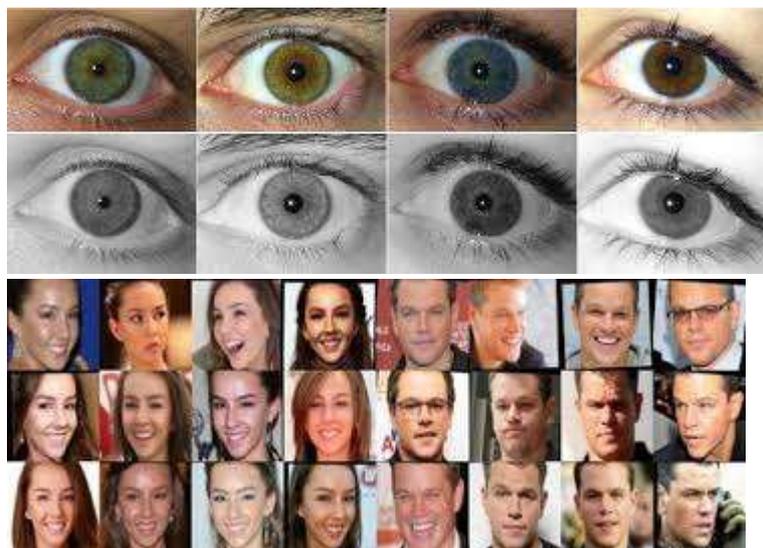


Fig 4: Iris Recognition and Facial Recognition Dataset Image

Step 2: Transfer the face picture and iris and convert the first picture to the Grayscale picture. In Grayscale picture lessen the first pixel and yield picture speaks to that the high contrast picture. In the wake of changing over in Grayscale picture, apply the edge location strategy utilizing the vigilant property. In watchful property recognize the most extreme, least and normal qualities in the edge picture.

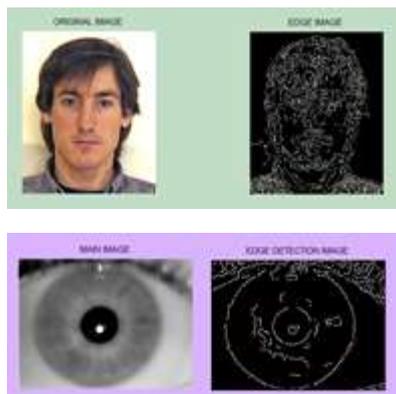


Fig.4(i) Facial Image Upload and Edge Detector and 6(ii) Iris Image Upload and Edge Detector

Step 3: In iris acknowledgment utilizing Hough Circle Transformation for figuring the inward and external range of the first picture. The assurance of the system is to discover hovers in deficient picture inputs. The circle competitors are contorted by "casting a ballot" in the Hough parameter space and afterward select the neighbourhood maxima in a supposed aggregator framework.

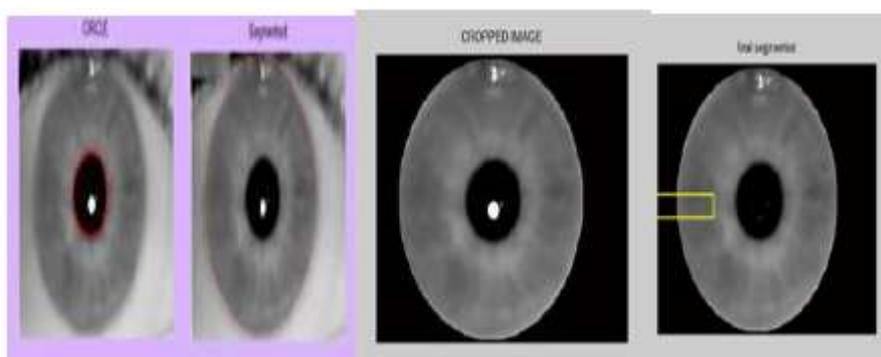


Fig.5 Hough Circle Output

Step 4: Apply the feature extraction algorithm utilizing the Sift calculation to identify the key focuses. Comparable highlights across changed pictures in a typical issue in PC vision. At the point when all pictures are comparable in nature, for example, the same scale, direction, and so forth, basic corner identifiers can work. In any case, when you have pictures of various scales and revolutions, you have to utilize the Scale Invariant Feature Transform. The filter isn't unprejudiced scale-invariant. You can change the ensuing, and still get great outcomes:

- Scale form
- Rotation
- Illumination
- Viewpoint

Well, that is some genuine lively picture coordinating going on. The large squares mark coordinated pictures. The littler square shapes are for singular structures in those districts. Note how the enormous squares are slanted. They follow the heading and point of view of the article in the scene.

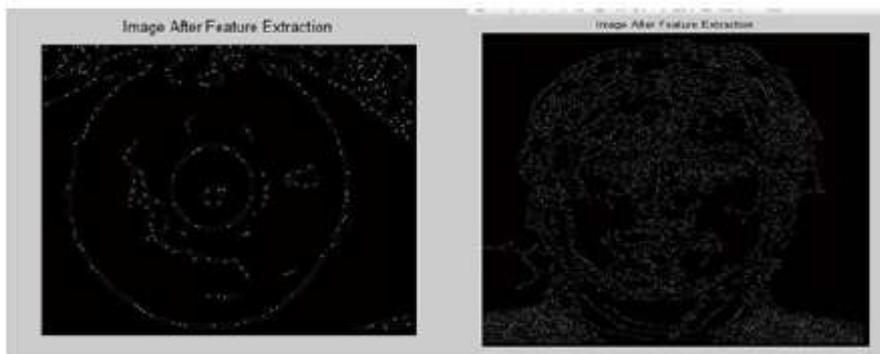


Fig.6 Iris and Face Feature Extraction Image using (SIFT)

Step 5: applying feature extraction, we applied the hereditary calculation for highlight decrease. The Genetic Procedure is a model of machine information that gets its presentation from a picture of the procedures of Evolution in the earth. This is finished by the creation inside a machine of a Populace of Individuals spoke to by Chromosomes, in the soul a lot of character strings that are like the base-4 chromosomes that we understand in our own DNA. The people in the masses at that point experience a procedure of advancement.

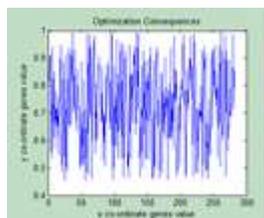


Fig.7 Optimized Output (GA)

Step 6: Apply score level Fusion: Fusing the scores of a few biometric frameworks is a promising way to deal with improves the general framework's precision. Regardless of numerous works in the writing, it is astonishing that there is no planned exertion in making a benchmark database accessible.

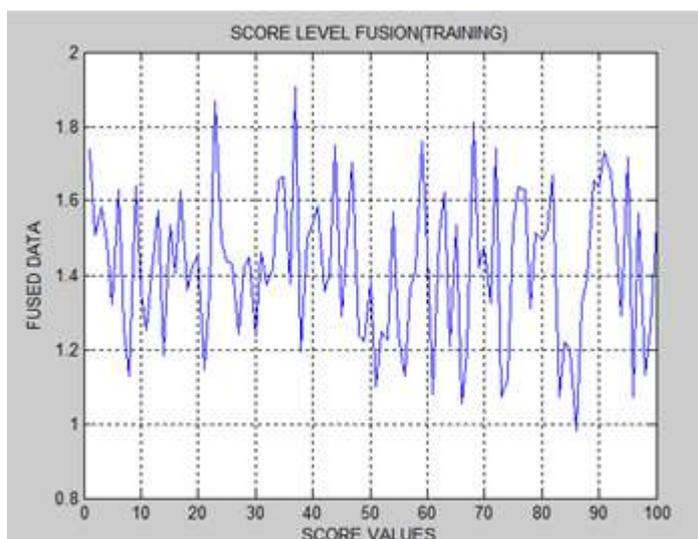


Fig.8 Score Level Fusion

Step 7: Apply for coordinating Euclidean Distance implies the greater part of these picture detachment methods depend on old style (for example Euclidean) measurements. Utilizing "quicker" separation work with lower edge levels and "more slow" separation work with a higher one, comparative outcomes can be acquired.

Step 8: After the coordinating procedure, order utilizing Back Propagation Neural Network. At that point compute the exhibition parameters like bogus acknowledgment rate, bogus dismissal rate, and precision.

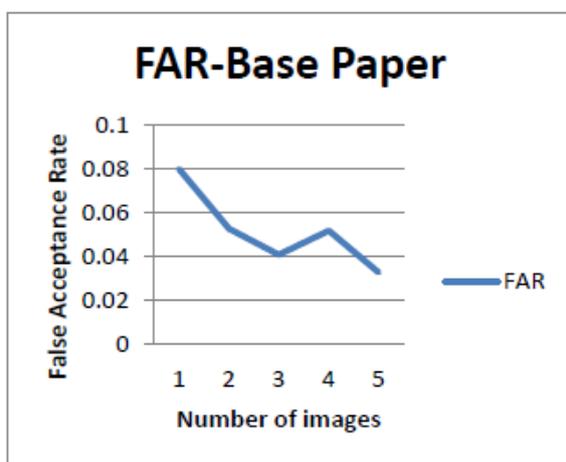
Step 9: After that, we contrast the current execution parameters and proposed work parameters i.e. precision.

**VII. RESULT AND DISCUSSIONS**

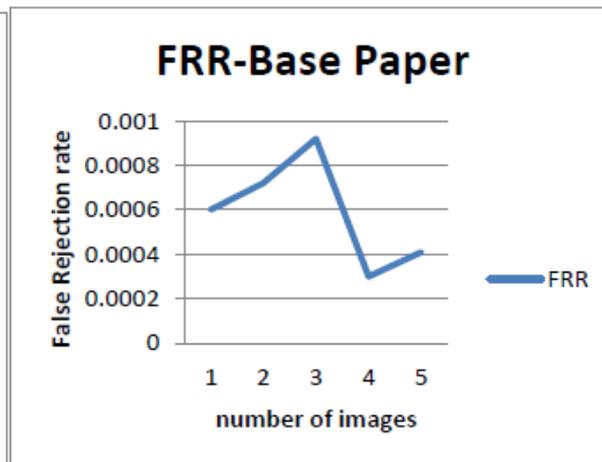
The below table characterized that the Iris and Ear Recognition combination of the Biometric verification. It shows that the correlation between the diverse distinctive presentation parameters like Far, Frr. As per the comparative multi-modular biometric verification framework.

Table 1: Fusion Iris and Ear (Base Paper)

Iris and Ear	FAR	FRR
Image 1	0.0799	0.00060
Image 2	0.0527	0.00072
Image 3	0.0407	0.00092
Image 4	0.0517	0.00030
Image 5	0.0327	0.00041



(i)



(ii)



(iii)

Fig.10 (i) False Acceptance Rate (Base Paper), (ii) False Rejection Rate (Base Paper) and (iii) Comparison Between Existing and Proposed Work (Far and Frr)

The above figure 10(i) shows that the current False acknowledgment rate. The above figure 10(ii) shows that the current bogus dismissal rate. The above figure 10(iii) characterizes the bogus acknowledgment and bogus dismissal case in the examination organize (Previous Work). Bogus acknowledgment intends to recognize the satisfactory mistake and bogus dismissal dismiss the non-adequate blunder.

Table 2 : Comparison between Existing Work and Proposed Work in Fusion FAR

Iris and Ear, Face	FAR(Base)	FAR(Proposed )
Image 1	0.0781	0.0090
Image 2	0.0562	0.0010
Image 3	0.046	0.0011
Image 4	0.0589	0.0013
Image 5	0.0331	0.0015

Table no: 3 Comparison between Existing Work and Proposed Work in Fusion in FRR

Iris and Ear, Face	FRR(Base )	FRR(Proposed)
Image 1	0.00060	0.0081
Image 2	0.00071	0.0090
Image 3	0.00094	0.0095
Image 4	0.00030	0.0014
Image 5	0.00044	0.0037

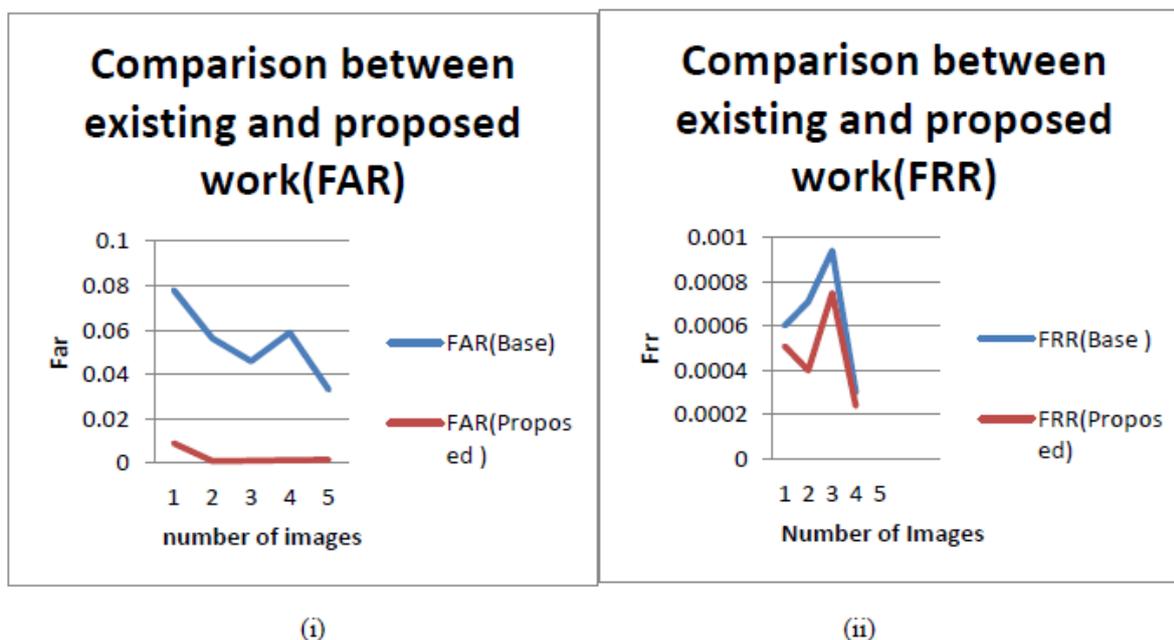


Fig.11(i) Comparison between Existing and Proposed Work (False Acceptance Rate) and (ii) Comparison between Existing and Proposed Work (False Rejection Rate)

The above figure 11(i) shows that the Comparison among Existing and Proposed Work in False Acceptance rate. We improve the exhibition among Previous and Proposed Work. The above figure 11(ii) shows that the Comparison among Existing and Proposed Work in False Rejection rate. We improve the exhibition among Previous and Proposed Work.

## VIII. CONCLUSION AND FUTURE SCOPE

This paper has proposed a framework that is based on iris and face. In the proposed system another procedure is produced at score level combination to expand the presentation of the iris and ear detection framework. Right now the framework is created utilizing SIFT and GA as it was. After that FAR, FRR and precision have been assessed in which SIFT performs well-having results like for SIFT Accuracy = 99.97 %, FAR= 0.0013, FRR= 0.00051. From the diagrams, it has been reasoned that the SIFT system functions admirably. Future works could go toward utilizing a Genetic calculation or ICA in hybridization with BFO. Free Component Analysis (ICA) is a computational strategy to get shrouded estimations of arbitrary factors. ICA fundamentally intended for multivariate information. The information utilized for investigating utilizing ICA can be started from numerous fields like financial aspects, advanced pictures, report databases, and so on. Likewise, the EGA Optimization Algorithm is all the more impressive for the issues with a few measures of factors given. EGA is very efficient in finding the entire hunt space or any of the arrangement space, which is exceptionally enormous and troublesome. The Enhanced Genetic calculation is executed utilizing PC re-enactment, procuring inhabitants of people, which is the arrangement space. The people experience the determination procedure by assessing the wellness work, utilizing a few administrators, for example, transformation and hybrid.