

(A High Impact Factor, Monthly, Peer Reviewed Journal)

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Comparative Analysis of Classifiers for Gender Identification Using Fingerprints

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ABSTRACT: Gender identification of the person can be distinguish by certain forms, as human have different characteristics and inimitable traits. Based on the aspect of individuals anatomy of biometrics like palm print or fingerprint interleaved ridges and valleys measured to identify the person. Biometric traits are essential to build applications such as human computer interaction searching, surveillance, content based indexing, biometric and demographic studies are necessary for gender identification. Several previous studies have investigated the gender information from the face image; some appreciable results have been obtained. Due to the age varying it limits the gender identification through machine learning approaches. In this proposed work an attempt is made of study and comparing different classifiers for gender information from the fingerprints.

KEYWORDS: Gender identification, fingerprint, LDA, QDA, DWT, KNN, SVM.

I. INTRODUCTION

Human vision system is self-sufficient and quite adaptive in identifying the approximated attributes of individuals such as age and gender. For example, the first moment of life we try to recognize mother and father. We constantly perform gender recognition, often without being aware of it. However, the computer vision system is not fully mature to do all these tasks. Consequently, gender identification of human beings play a vital role in bisects to the search space and developing various applications such as human-computer interaction, surveillance, targeted advertising and demographic studies etc[1][9][12].

Classification and identification of fingerprint have been used by many researchers; some researchers have studied the fingerprint gender classification problem. The gender of the person can be identified using fingerprints which is unique, based on the ridges counts gender of the person can be judged[2,4,7,10,11]. In [8] shows both males and females have higher right word directional asymmetry in the ridge count. According to the survey the average ridges count of the male is slightly higher the females, fingerprints are widely used biometric for personal individual authentication. As fingerprint acquisitions is a simple and is one of the prominent biometrics as fingerprints are unvarying but its size and shape may change and differ with age, but basic patterns of the fingerprint remain unchanged.

Many algorithms for fingerprint identification are implemented for security and personal authentication but not well established. In this paper, we aim to examine the impact of gender identification on the enhancement of the performance of the biometric system by proposing comparative analysis of classifiers using fingerprint for gender identification. The rest of this paper is organized as follows: In Section 2, the related work is reported. The feature



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extraction and the flow of work carried out is placed in Section 3. Experimental results are discussed in section 4 and concluding remarks are reported in Section 5.

II. RELATED WORK

Preceding research have shown that it is possible to authenticate an individual by obtaining information from fingerprint images but however a limited work is carried out on identifying individuals gender using fingerprint images. However a few studies are reported on gender identification using fingerprint.

The authors in [13] have worked on integrating faced, fingerprints and soft biometrics traits, using gender, height, weight, eye color, ethnicity and age for user identification. Dataset of ten images were collected for 263 users. The result of 96.3% for authenticity 89.6% for gender classification is obtained.

In the work [14] the authors have worked on gender classification from fingerprints, which is important in forensic anthropology. A fingerprint data set of 2200 person of different age groups were collected, where 1100 were males and 1100 females. Features like ridges count, ridge thickness and valley thickness ratio (RTVTR), white lines count, ridge count asymmetry and pattern type concordance were extracted the dominant feature like Fuzzy-C means (FCM), linear discriminate analysis (LDA) and neural network (NN) were used for classification. The results of 80.39%, 86.5% and 88.5% were obtained for FCM, LDA and NN respectively.

In[15] authors have worked on gender classification from fingerprints. Many features like ridge width, ridge thickness to valley thickness ratio and ridge density [7] were extracted. Support vector machine is used as classifier. The internal data set of 400 fingerprints of male and female were used for experiments and achieved 91% classification rate.

In [16] authors have proposed the method for gender classification using discrete wavelet transform and singular value decomposition. The authors have computed the energy from all the sub-bands of DWT and combined with special features of non-zero singular values. Obtained SVD of fingerprints were extracted for gender classification using K nearest neighbor (KNN) as a classifier. The experiments was conducted on internal database of 3570 fingerprints where 1980 are male and 1590 are female. The result of 94.32% and 95.46 is obtained for left hand little finger for female and male respectively and also the result of gender classification for any finger for male and female was 91.67% and 84.69% respectively which gives the classification rate of 88.28%.

In the work [17] gender identification based on fingerprints using frequency domain analysis, gender classification is achieved by analyzing fast Fourier transform (FFT), discrete cosine transform (DCT) and power spectral density (PSD). An internal dataset of 220 person of different age group were collected. Using pre determined threshold frequency domain collection are compared the results of 70.07% and 90% for male and female obtained respectively.

In [18] have presented the worked on real time approach to determine the gender using fingerprints and they have focused on extracting the features based on frequency domain analysis, Discrete wavelet transform (DWT) and singular value decomposition(SVD). A dataset of 300 persons of different age and gender is collected. The proposed algorithm produces an accuracy decision of 92% for female and 80.02% for male.

In [19] authors have proposed a novel method for gender classification using discrete wavelet transform and singular value decomposition techniques" authors have focused on 2D-discrete wavelet transformation (DWT) to find frequency domain vector and singular value decomposition in order to find spatial features using K-nearest neighbour classifier. A dataset of 100 left hand index finger fingerprints were collected belong into same age group. The success rate is more than 80% is obtained in this experiments.



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In [20] presented a method for gender identification by means of combined feature have been proposed using FFT, Major Axis Length on the dataset combination of 450 male and 550 female samples of total 1000 numbers. Analysis of left thumb impression of each sample has been done for internal database. Optimal threshold was chosen to get promotable result. The algorithm produces accurate classification of 80% of male and 78% of female.

In the work [21] the authors have proposed a method for fingerprints based gender classification through frequency domain analysis and work with the features of 2D discrete wavelet transform (DWT) and principle component analysis (PCA) with the dataset of 2400 person fingerprints of different age groups. Classification is obtained using minimum distance method and overall success rate of 70% is obtained for gender classification.

The Neural net-work based gender classification [23], using Discrete wavelet transform (DWT) the directional images of fingerprints are obtained. Then the feed-forward back-propagation neural network is applied for male-female fingerprints recognition. The dataset of 300 fingerprints were collected. Success rate of classification was 91.3%.

In the work [24] authors have focused on gender classification through finger print biometric using discreet curvelet transform and back propagation with feed forward using neural network as classifier. The extracted features from fingerprints data base used for learning the stage of neural networks, which acts as automatic classifier two identify the gender. Image orientation wedges are decomposed using fast discreet curvelet transform which provides texture pattern and intensities. Using these pattern haralick or glcm features are estimated to get better gender classification using fingerprints.

In [25] have carried out a research work on fingerprint based gender classification using minutiae extraction by considering the minutiae details like intersecting points, ridge counts, number of blobs, terminating points and ridge counts. Acceptable and spirited results were observed which are helpful and practically applicable in forensic anthropology.

In Gender classification using fingerprints based on support vector machine (SVM) with 10-cross validation techniques [26] 89% and 91% classification rate is achieved for DWT using SVM(RBF sigma) and SVM(polynomial) using SVM classifier respectively.

In [27] have proposed heralick features descriptors for gender classification using fingerprint, using machine learning approach. The classifier like linear discriminant analysis and quadratic discriminant analysis classifier are used to extract the result of 92% and 94& are obtained for male and female respectively. In a proposed method[28] authors have tried to identifying individual's gender along with their age parameter from 280 fingerprint images, with combining the features of DWT and Principal Component analysis(PCA) over Back propagation Neural Network they showed accuracy of 82.14%.

In [29] took advantage of lab setup and extracted ridge related local features like white line count, ridge thickness to valley ratio(RTVTR),ridge density are calculated with multilayer perception neural network classifier over 3000 fingerprint database ,overall obtained accuracy is of 97% with complex setup.

From the above literature review, we realized that, authors have presented different types of features and techniques among wavelets have given promising results, with different age groups, but which classifier is more suitable is still interesting question[3][22]. In this paper, as preliminary study we investigated the performance of different classifiers for gender determination using finger prints.



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III. PROPOSED WORK

Our method consists of three steps namely pre-processing, feature-extraction and classification. First an input image is gone though preprocessing which involves usage of adaptive Histogram equalization technique for improving the quality of the fingerprint image. Further for feature computation we employed discrete wavelets transformation. Lastly behavior of different classifiers is observed for gender identification task. We have shown the diagrammatic representation of proposed method in the figure 1





A. Image Acquisition

As there is no separate standard dataset for male and female fingerprints, a dataset of fingerprints of different age group were collected from rural and urban areas. Hence experiment was conducted using our own dataset. The main purpose for the creation of database is to check the robustness in the gender identification. Database contains of totally 74 healthy persons fingerprints, of which 37 are male volunteers and 37 female volunteers. Each individuals ten fingers print images are collected giving them a prior directions. Thus totally we have 740 fingerprints images, which are further been considered for the experiment. Sample from our database are shown below in figure 2.



Fig. 2. Sample of fingerprints

B. Pre-Processing

Pre-Processing is considerable step and application oriented. A technique Contrast limited and adoptive histogram equalization [CLAHE] is applied for improving the quality of the fingerprint image then normalized the image to size of 164x164 which is empirically fixed, further. The process of preprocessing can be seen in figure 3.



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Fig. 3. Preprocessing Step a) Input Fingerprint image b) Enhanced Image after applying CLAHE, c) Image resized by 164*164

C. Feature extraction

Wavelets have been shown to be useful for texture analysis and most popular tool in multi-resolution image analysis and in object recognition research area . in this proposed methodology two dimensional DWT over fingerprints are implemented. Image representations of low-high, high-low and high-high sub bands of horizontal, vertical and diagonal representation are obtained respectively. Finally low-low sub band presents the approximation of all three details. A fingerprint images are decomposed into four levels with returns 16 sub-bands by applying DWT. Let us consider 'a' be the mother wavelet transform represented by:

$$\Psi_{(\mathbf{x},\mathbf{y})}(\mathbf{a}) = \frac{1}{\sqrt{x}} \psi \left[\frac{a - x}{y} \right].$$
 (1)

Where x & y are shifting parameters. By multiplying one dimensional scaling with wavelet function, one dimensional wavelets extends to two dimensional scaling to get two dimensional wavelet transform, one two dimensional scaling function $\Phi_A(x,y)$ and three two dimensional scaling function $\Psi_H(c,d)$, $\Psi_V(c:d)$ and $\Psi_D(c:d)$ are used. The product of one-dimensional scaling Φ and corresponding wavelet ψ are followed in equation 1-5.

$$\begin{split} \Psi^{A}(\mathbf{c}, \mathbf{d}) &= \Psi(\mathbf{c}) \oslash (\mathbf{d}) \quad \dots \quad (2) \\ \Psi^{V}(\mathbf{c}, \mathbf{d}) &= \oslash (\mathbf{c}) \Psi(\mathbf{d}) \dots \quad (3) \\ \Psi^{D}(\mathbf{c}, \mathbf{b}) &= \Psi(\mathbf{c}) \Psi(\mathbf{d}) \dots \quad (4) \\ \Psi^{H}(\mathbf{c}, \mathbf{d}) &= \Psi(\mathbf{c}) \oslash (\mathbf{d}) \dots \quad (5) \end{split}$$

Where Φ_A represents global information of the preserved image under low frequency components. Approximation co-efficient Ψ_H , Ψ_V and Ψ_D represents horizontal, vertical and diagonal directions variation respectively. A two level decomposition of fingerprint image is shown in figure 4.

Discreet wavelet transform is applied to each input image. Upto fourth level decomposition we got 16 sub bands. From each sub band 2 statistical measures namely standard deviation and entropy are calculated. Using 2 statistical measures from each sub band we got 16*2=32 dimensional feature vectors for each input fingerprints.



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Fig 4. Feature extraction from male and female image

D. Classification

K-NN Classifier:

KNN classification is found to be simple and conventional non-parametric classifier; it classifies the data based on measuring the distances between the test data and each of the training data to decide the final classification output. Prediction of class data is done on the basic of its neighbor [17]. In this work, we have used value of k=3, which is fixed empirically with city-block distance as distance measures and 5 fold cross validation.

Let us consider M represents training data and N denotes for testing pattern Then KNN finds the minimum distance d between using its components Mj and Nj

 $d_{City-block}(M,N) = \sum_{i=1}^{m} |Mj - Nj|$

SVM Classifier:

Support Vector Machine (SVM) is a classification technique based on the statistical learning theory proposed by vapnik in 1995 [5]. It is a binary classifier it abstracts a decision boundary in multidimensional space using an appropriate sub set of the training set of vectors; the elements of this sub set are the support vectors. Geometrically SVM are those training patterns that are closest to the decision boundary. As a kernel function is simply a dot product of two feature vectors transformed in such a expanded feature vector spaces. Various kernel functions are available in SVM classifier, in our experiment we have used with cubic Kernel with 3rd degree kernel function.

LDA Classifier:

Fisher discriminant analysis is also called ad discriminant analysis in a classification. Based on different Gaussion distributions different classes are generated. Gaussian distribution parameters are estimated by fitting function for each class in order to train the classifier. A trained classifier predict new classes of data and find class with smallest misclassification cost[1].

Random forest and random tree:

Random forest is a kind of both classification and regression model, as in machine learning approaches is more resistant to over fitting to their training set. Random forest creates number of trees i.e. magnitude of decision tree which try to evaluate the root node .Random forest averages multiple decision trees over trained set, with goal of reducing variance, at the expense of small bias it generally boost the performance in the final model classification. Random forest algorithm is generally is supervised classification algorithm [6].



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Table 1: Detail Result

DETAIL RESULT				
CLASSIFIERS	OVERALL ACCURACY			
Linear Discriminant	85.48%			
Analysis(LDA)				
Quadratic discriminating	92.2			
analysis(QDA)				
KNN	96.1%			
Random forest	97.7			
SVM	99%			

IV. EXPERIMENT AND RESULTS

In the proposed method, we aim to study comparative analysis of classifiers for the task of gender identification using fingerprints. For identifying gender of an individual, we have trained the system to classify male and female based on their fingerprints, creating dataset, which consists of 740 fingerprints belonging to 74 volunteers. We have trained our proposed system with different parametric and non-parametric classifiers to check the performance evaluation of well known classifying techniques, with best accuracy of 99% achieved with support vector machine (Cubic-SVM) classifier. Random Forest Classifier gives an accuracy of 97.7%...But K-NN follows similar tends with accuracy of 96.1%.QDA classifier follows with 92.2% of accuracy and poorest accuracy of 85.48% is obtained with Linear discriminant analysis(LDA). SVM follows similar trend and yield highest accuracy of 99.0% repectively, LDA provides accuracy of 85.48% respectively. The confusion matrix of highest and lowest results is shown in the table 2.

SVM			LDA	
Male	Female	Male	Female	
369	1	299	71	
0	370	46	324	

	Fable	2:	Confusion	Matri
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V. CONCLUSIONS

In this paper, we have attempted human gender identification problem using fingerprints. We have made comparative study with different classifiers. Discrete Wavelets Transformation (DWT) method is used to extract discriminating texture features from fingerprints. These features have exhibited significant performance. Exhaustive experimentations have been done on a relatively larger dataset of 740 fingerprint images with different validating classifiers the performance of the algorithm. This analysis reveals that the proposed method has the remarkable performance with SVM classifier compared with the other classifiers. In future, we concentrate on the evaluation of: 1) the proposed system on different biometric traits and identify gender using different traits. 2) To combine different



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biometric traits fuse them according to their modality and find gender using multimodal biometrics.3) We plan to evaluate the auxiliary information like age along with gender.

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