

A Survey on Face Recognition in Present Scenario

Sneha R. Madane, Prof. S.T. Khandare

Abstract— Face recognition has been a fast growing, challenging and interesting area in real time applications. A large number of face recognition algorithms have been developed in last decades. In this paper an attempt is made to review a wide range of methods used for face recognition comprehensively. This include PCA, LDA, ICA, SVM, Gabor wavelet soft computing tool like ANN for recognition and various hybrid combination of this techniques. This review investigates all these methods with parameters that challenges face recognition like illumination, pose variation, facial expressions. In the age of rising crime face recognition is enormously important in the contexts of computer vision, psychology, surveillance, fraud detection, pattern recognition, neural network, content based video processing, etc. Face is a non-strong biometrics for identification intrusive and hence criminals always try to hide their facial organs by different artificial means such as plastic surgery, disguise and dummy.

Index Terms— Artificial Neural Networks (ANN), Face Recognition, Independent Component Analysis (ICA), Linear Discriminate Analysis (LDA), Principal Component Analysis (PCA).

I. INTRODUCTION

Face recognition is an important part of the capability of human perception system and is a routine task for humans, while building a similar computational model of face recognition. The computational model not only contribute to theoretical insights but also to many practical applications like automated crowd surveillance, access control, design of human computer interface (HCI), content based image database management, criminal identification and so on.

II. LITERATURE SURVEY

The earliest work on face recognition can be traced back at least to the 1950s in psychology [1] and to the 1960s in the engineering literature[2]. Some of the earliest studies include work on facial expression emotions by Darwin [3]. But research on automatic machine recognition of faces started in the 1970s [4] and after the seminal work of Kanade [5]. In 1995, a review paper [6] gave a thorough survey of face recognition technology at that time [7]. At that time, video-based face recognition was still in a nascent stage. During the past decades, face recognition has received increased attention and has advanced technically. Many commercial systems for still face recognition are now available. Recently, significant research efforts have been focused on video-based face modeling/tracking, recognition and system integration. New databases have been created and

evaluations of recognition techniques using these databases have been carried out. Now, the face recognition has become one of the most active applications of pattern recognition, image analysis and understanding. Face detection and recognition are challenging tasks due to variation in illumination, variability in scale, location, orientation (up-right, rotated) and pose (frontal, profile). Facial expression, occlusion and lighting conditions also change the overall appearance of face. Face detection and recognition has many real world applications, like human/ computer interface, surveillance, authentication and video indexing.

III. INPUT IMAGE NORMALIZATION

Image normalization is the first stage for all face recognition systems. Firstly face area is detected in the image. We used template matching to localize a face. Then the eye (iris) centers should be detected because the distance between them is used as a normalization factor. We located the eyes in facial images of different size using the luminance component. We required the eyes must be open in input images. The gray-scale image was processed by an edge detector (we used Sobel). Then it was binarized by Otsu method [8]. We calculated the vertical integral projection [9] of the binary image and smoothed the projection profile by averaging of neighboring values. The profile was searched for a large valley with an intervening high peak. The peak indicates to the eye area in the image (see the right part of Fig.1). To locate eyes we applied Hough transform to a small strip of the binary image (the shaded area in Fig.1) using a half circle mask as often the upper part of the iris is covered by eyelid. To speed up the processing, we store several masks corresponding to different radii and then use for Hough Transform.

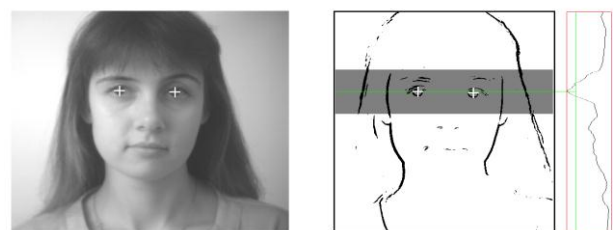


Fig.1. Original face image and located eye centers

Among several peaks in the Hough space we find two highest scoring candidates. The crosses in Fig.1 show the centers of eyes and some details of their location. The output of this step is the coordinates of the eyes. In the next step we transformed the initial images by rotation, scaling and cropping of the central face part. We tried to remove background and hair, and keep the most invariant (in time) part of a face. The most important fiducially features are situated around eyes, eyebrows, and nose. We have observed that the nose shape

Manuscript received April 23, 2015.

Sneha R. Madane, Student, Computer department, B. N. College of Engineering, Pusad, Dist- Yavatmal (India).

Prof. S. T. Khandare, Computer department, B. N. College of Engineering, Pusad, Dist- Yavatmal (India).

varies a lot as a consequence of head rotation and lighting conditions. Finally mouth is the most variant part of the face. We can rely on the fact that face expression in document photos is usually minimal, at least discreet, while for a camera control image the test person can be requested to restrain from excessive expressions. Additional features from the original image can be gleaned in the form of a gray-scale edge map using Deriche algorithm. Examples of the input images are given in the top of Fig.2.

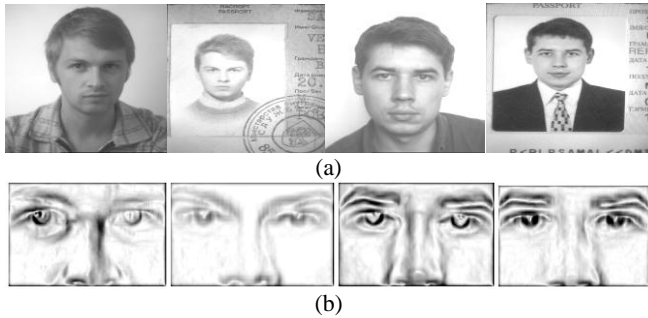


Fig.-2. (a) Original images, (b) Normalized images

IV. FACE RECOGNITION ALGORITHMS

A. Principal Component Analysis (PCA):

PCA also known as Karhunen-Loeve method is one of the popular methods for feature selection and dimension reduction. Recognition of human faces using PCA was first done by Turk and Pentland [10] and reconstruction of human faces was done by Kirby and Sirovich [11]. The recognition method known as eigen face method defines a feature space which reduces the dimensionality of the original data space. This reduced data space is used for recognition. But poor discriminating power within the class and large computation are the well-known common problems in PCA method. This limitation is overcome by Linear Discriminant Analysis (LDA). LDA is the most dominant algorithms for feature selection in appearance based methods. But many LDA based face recognition system first used PCA to reduce dimensions and then LDA is used to maximize the discriminating power of feature selection. The reason is that LDA has the small sample size problem in which dataset selected should have larger samples per class for good discriminating features extraction. Thus implementing LDA directly resulted in poor extraction of discriminating features. In the proposed method [12] Gabor filter is used to filter frontal face images and PCA is used to reduce the dimension of filtered feature vectors and then LDA is used for feature extraction. The performances of appearance based statistical methods such as PCA, LDA and ICA are tested and compared for the recognition of colored faces images in. PCA is better than LDA and ICA under different illumination variations but LDA is better than ICA. LDA is more sensitive than PCA and ICA on partial occlusions, but PCA is less sensitive to partial occlusions compared to LDA and ICA. PCA is used as a dimension reduction technique in [13] and for modeling expression deformations in a recursive algorithm for calculating the discriminate features of PCA-LDA procedure is introduced. This method concentrates on challenging issue of computing discriminating vectors from an incrementally arriving high

dimensional data stream without computing the corresponding covariance matrix and without knowing the data in advance. The proposed incremental PCA-LDA algorithm is very efficient in memory usage and it is very efficient in the calculation of first basis vectors. This algorithm gives an acceptable face recognition success rate in comparison with very famous face recognition algorithms such as PCA and LDA. Two appearance-based techniques such as Modified PCA (MPCA) and Locality Preserving Projections (LPP) are combined in [14] to give a high face recognition rate. PCA is used as a feature extraction technique in. These feature vectors are compared using Mahalanobis distances for decision making. Tensor based Multi linear PCA approach is proposed in which extracts feature directly from the tensor representation rather than the vector representation. This method shows a better performance in comparison with the well-known methods in distance varying environments. A different approach for face detection was proposed in which minimizes computation time while achieving higher detection accuracy. PCA was used to reduce the dimension extracting a feature vector. GRNN used as a function approximation network to detect whether the input image contains a face or not and if existed then reports about its orientation. The proposed system had shown that GRNN can perform better than back propagation algorithm and give some solution for better regularization. The simulation results show that the proposed method results in a better performance than conventional PCA and LDA approaches and the computational cost remains the same as that of PCA and much less than that of LDA.

B. Support Vector Machine (SVM):

Support Vector Machines (SVM) is one of the most useful techniques in classification problems. One clear example is face recognition. However, SVM cannot be applied when the feature vectors defining samples have missing entries. A classification algorithm that has successfully been used in this framework is the all-known Support Vector Machines (SVM) [15], which can be applied to the original appearance space or a subspace of it obtained after applying a feature extraction method [16]. The advantage of SVM classifier over traditional neural network is that SVMs can achieve better generalization performance.

C. Independent Component Analysis (ICA):

Independent component analysis (ICA) is a method for finding underlying factors or components from multivariate (multidimensional) statistical data. There is need to implement face recognition system using ICA for facial images having face orientations and different illumination conditions, which will give better results as compared with existing systems[17]. What distinguishes ICA from other methods is that, it looks for component that are both statistically independent and non Gaussian. The ICA is similar to blind source separation problem that boils down to finding a linear representation in which the components are statistically independent. The comparisons of face recognition using PCA and ICA on FERET database with different classifiers were discussed [18] and found that the ICA had better recognition rate as compared with PCA with statistically independent basis images and also with

statistically independent coefficients. Face recognition using ICA with large rotation angles with poses and variations in illumination conditions was proposed. A novel subspace method called sequential row column independent analysis for face recognition is proposed. In ICA each face image is transformed into a vector before calculating the independent components. RC_ICA reduces face recognition error and dimensionality of recognition subspace becomes smaller. A novel technique for face recognition combined the independent component analysis (ICA) model with the optical correlation technique was proposed. This approach relied on the performances of a strongly discriminating optical correlation method along with the robustness of the ICA model. Independent component analysis (ICA) model had sparked interest in searching for a linear transformation to express a set of random variables as linear combinations of statistically independent source variables. ICA provided a more powerful data representation than PCA as its goal was that of providing independent rather than uncorrelated image decomposition and representation. A fast incremental principal non Gaussian directions analysis algorithm called IPCA_ICA was proposed. This algorithm computes the principal components of a sequence of image vectors incrementally without estimating the covariance matrix and at the same time transforms these principal components to the independent directions that maximize the non-Gaussianity of the source. IPCA_ICA is very efficient in the calculation of the first basis vectors. PCA_ICA achieves higher average success rate than Eigen face, the Fisher face and Fast ICA methods.

D. Gabor wavelet:

For enhancing face recognition high intensity feature vectors extracted from Gabor wavelet transformation of frontal face images combined together with ICA in [19]. Gabor features have been recognized as one of the best representations for face recognition. In recent years, Gabor wavelets have been widely used for face representation by face recognition researchers [20], because the kernels of the Gabor wavelets are similar to the 2D receptive field profiles of the mammal cortical simple cells, which exhibits desirable characteristics of spatial locality and orientation selectivity. Previous works on Gabor features have also demonstrated impressive results for face recognition. Typical methods include the dynamic link architecture (DLA) elastic bunch graph matching (EBGM) Gabor Fisher classifier (GFC), and Ada Boosted GFC (AGFC) [20]. In this paper [21], it was observed that though Gabor phases are sensitive to local variations, they can discriminate between patterns with similar magnitudes, i.e. they provide more detailed information about the local image features. Therefore, the Gabor phases can work comparably well with the magnitudes, as long as its sensitivity to misalignment and local variations can be compensated carefully. In previous work, authors proposed to represent face images using the local Gabor binary patterns (LGBP), which combines Gabor magnitudes with local binary patterns (LBP) operator. Improved results were achieved when compared with the LBP and the GFC. Since face representation with LGBP based on local histograms, which were insensitive to local variations, similarly local histograms of LGBP can be used to suppress the sensitivity of Gabor phases to local variations. By encoding Gabor phases through

LBP and local histograms, a very impressive recognition rates comparable with those of Gabor magnitudes-based methods were achieved, which shows effectiveness of Gabor phases in the discrimination of different faces. A novel method for extraction of facial features is based on Gabor wavelet representation of face images and kernel least squares discrimination algorithm. The experimental results based on XM2VTS and ORL databases shown that Gabor based kernel least squares discrimination approach outperforms feature extraction methods such as PCA, LDA, Kernel PCA or Generalized Discriminate Analysis (GDA) as well as combination of these methods with Gabor representations of face images. A technique is presented in by which high intensity feature vectors extracted from the Gabor wavelet transformation of frontal face images combined together with ICA for enhanced face recognition.

E. Linear Discriminate Analysis (LDA):

The linear discriminate analysis (LDA) is a powerful method for face recognition. It yields an effective representation that linearly transforms the original data space into a low-dimensional feature space where the data is well separated. However, The within-class scatter matrix (SW) becomes singular in face recognition and the classical LDA cannot be solved which is the under sampled problem of LDA (also known as small sample size problem). A subspace analysis method for face recognition called kernel discriminate locality preserving projections (MMDLPP) is based on the analysis of LDA, LPP and kernel functions. A nonlinear subspace which can not only preserves the local facial manifold structure but also emphasizes discriminate information. Combined with maximum margin criterion (MMC) new method called maximizing margin and discriminate locality preserving projections (MMDLPP) was proposed in [22] to find the subspace that best discriminates different face change and preserving the intrinsic relations of the local neighborhood in the same face class according to prior class label information. The proposed method was compared with PCA as well as locality preserving projections (LPP) ORL, YALE, YALEB face database and authors had shown that it provides a better representation of class information and achieved better recognition accuracy. Illumination adaptive linear discriminate analysis (IALDA) was proposed in [23] to solve illumination variation problems in face recognition. The recognition accuracy of the suggested method (IALDA), far higher than that of PCA method and LDA method. The recognition accuracy of the suggested method was lower than that the Logarithmic Total Variation (LTV) algorithm. However, The LTV algorithm has high time complexity. Therefore, the LTV method is not practically applicable. At the same time, this also indicates that the proposed IALDA method is robust for illumination variations. David Monzo, compared several approaches to extract facial landmarks and studied their influence on face recognition problems. In order to obtain fair comparisons, they used the same number of facial landmarks and the same type of descriptors (HOG descriptors) for each approach. The comparative results were obtained using FERET and FRGC datasets and shown that better recognition rates were obtained when landmarks are located at real facial fiducially points.

F. Artificial Neural Network (ANN):

Multi-Layer Perceptron (MLP) with a feed forward learning algorithms was chosen for the proposed system because of its simplicity and its capability in supervised pattern matching. It has been successfully applied to many pattern classification problems [24]. The method used Gabor wavelet transform and feed forward neural network for both finding feature points and extracting feature vectors. The experimental results have shown that proposed method achieves better results compared to the graph matching and eigen faces methods, which are known to be the most successful algorithms. A new class of convolutional neural network was proposed in [25] where the processing cells are shunting inhibitory neurons. Previously shunting inhibitory neurons have been used in a conventional feed forward architecture for classification and non-linear regression and were shown to be more powerful than MLPs [26] i.e. they can approximate complex decision surfaces much more readily than MLPs. A hybrid neural network solution was presented in which combines local image sampling, a self-organizing map neural network, and a convolutional neural network. The self-organizing map provides a quantization of the image samples into a topological space where inputs that are nearby in the original space are also nearby in the output space, thereby providing dimensionality reduction and invariance to minor changes in the image sample, and the convolutional neural network (CNN) provides for partial invariance to translation, rotation, scale, and deformation. PCA+CNN & SOM+CNN methods are both superior to eigen faces technique even when there is only one training image per person. SOM +CNN method consistently performs better than the PCA+CNN method. A new face detection method is proposed in [27] using polynomial neural network (PNN). The PNN functions as a classifier to evaluate the face likelihood of the image patterns of the multi-scale shifted local regions. The PCA technique used to reduce the dimensionality of image patterns and extract features for the PNN. Using a single network the author had achieved fairly high detection rate and low false positive rate on images with complex backgrounds. In comparison with a multilayer perceptron, the performance of PNN is superior. To best reflect the geometry of the 3D face manifold and improve recognition. Spectral Regression Kernel Discriminate Analysis (SRKDA) based on regression and spectral graph analysis introduced in proposed method [28]. When the sample vectors are linearly independent, which is usually the case for small sample size problems; SRKDA can efficiently give more exact solutions than ordinary subspace learning approaches. It not only solves high dimensional and small sample size problems, but also enhances feature extraction from face local non-linear structure. Detailed comparisons between SRKDA, PCA, LPP, OLPP, SR, and KDA to show the efficiency of proposed method for 3D face recognition, especially with respect to expression variations. SRKDA only needs to solve a set of regularized regression problems and no eigen vector computation involved, which is a huge saving in computational cost.

V. ADVANTAGES AND DISADVANTAGES

Advantages:

1. The main advantage of PCA technique is that it can reduce the data needed to identify the individual to 1/1000th of the

data presented and also reduce the dimension extracting a feature vector.

2. LDA aims to maximum between class (across users) variance and minimum within class (within user) variance.

3. A Gabor Wavelet technique achieved higher recognition rate and better classification efficiency when feature vectors had low dimensions.

Disadvantages:

1) PCA is usually used but it is very time consuming.

VI. CONCLUSION

In this paper, we have addressed the problems needed to overcome for face recognition such as light intensity variable, facial expression etc. and we have discussed certain requirements for a reliable and efficient face recognition system like accuracy, efficiency. We have reviewed different face recognition algorithm. Our future works is made a comparison of these algorithms and have discussed the advantages and drawbacks of each of them.

REFERENCES

- [1]. Bruner, I. S. and Tagiuri, R. "The perception of people in andbook of Social Psychology", Vol. 2, G. Lindzey, Ed., Addison-Wesley, Reading, MA, 634-654.1954
- [2]. Bledsoe, W. W. "The model method in facial recognition, Panoramic research , 1964.
- [3]. Ekman, P. Ed., Charles Darwin's "The Expression of the Emotions in Man and Animals, Third Edition, withIntroduction", After words and Commentaries by PaulEkman. Harper- Collins/Oxford University Press, New York, NY/London, U.K.1998.
- [4]. Kelly, M. D. "Visual identification of people by computer", Stanford AI Project, Stanford, CA., 1970.
- [5]. Kanade, T., "Computer recognition of human faces", Birkhauser, Basel, Switzerland, and Stuttgart, Germany 1973
- [6]. Chellapa, R., Wilson, C. L., and Sirohey, S. "Human and machine recognition of faces: A survey" IEEE,83, 705-740.1995.
- [7]. Samal, A. and Iyengar, P., "Automatic recognition and analysis of human faces and facial expressions: A survey. Patt.Recog", 1992.
- [8]. N. Otsu, "A threshold selection method from the gray-level histograms" IEEE Trans. on Syst.,Man, Cybern.-1979.
- [9]. R. Brunelli and T. Poggio, " Face recognition: features versus templates" IEEE Trans. on PAMI.-1993.
- [10]. M. Turk and A. Pentland, "Eigenfaces for recognition," J. Cognitive Neuroscience, 1991.
- [11]. D. L. Swets and J. J. Weng, "Using discriminant eigenfeatures for image retrieval", IEEE, 1996.
- [12]. C.Magesh Kumar, R.Thiyagarajan ,S.P.Natarajan, S.Arulselvi, G.Sainarayanan, "Gabor features and LDA based Face Recognition with ANN classifier", Proceedings.Of ICETECT 2011
- [13]. Y. Cheng, C.L. Wang, Z.Y. Li, Y.K. Hou and C.X. Zhao, "Multiscale principal contour direction for varying lighting face recognition", Proceedings of IEEE 2010
- [14]. J. Shermina,V. Vasudevan, "An Efficient Face recognition System Based on Fusion of MPCA and LPP", American Journal of Scientific ResearchSSN 1450-223XIssue 11,2010.
- [15]. Vapnik,JohnWiley and Sons, "Statistical Learning Theory", New York, 1998.
- [16]. B. Heisele, T. Serre, and T. Poggio, "A component based framework for face detection and identification", IJCV, 74(2):167-181,2007.
- [17]. Pong C.Yuen, J.H.Lai, "Face representation using independent component analysis", PatternRecognition 35, 2002.
- [18]. Jian Yang, David Zhang, Jing-yu Yang, "Is ICA Significantly Better than PCA for Face Recognition?"Proceedings of the Tenth IEEE International Conferenceon Computer Vision(ICCV'05).
- [19]. ArindamKar, DebotoshBhattacharjee, Dipak Kumar Basu, MitaNasipuri, MahantapasKundu, "High Performance Human Face

Recognition using Independent High Intensity Gabor Wavelet Responses: A Statistical Approach”, International Journal of Computer Science & Emerging Technologies 178 Volume 2, Issue 1, February 2011

- [20]. Wenchao Zhang , Shiguang Shan, Laiyun Qing, Xilin Chen, Wen Gao, “Are Gabor phases really useless for face recognition?”, Springer-Verlag London Limited 2008,
- [21]. Shen L, Bai L, “A review on Gabor wavelets for face recognition”, Pattern Anal Appl 9(10):273–292, 2006
- [22]. Xiaohu Ma, Yanqi Tan, Yaying Zhao, Hongbo Tian, “Face Recognition Based on Maximizing Margin and Discriminant Locality Preserving Projection”, Journal of Information & Computational Science, 2010.
- [23]. Zhonghua Liu, Jingbo Zhou, Zhong Jin, “Face recognition based on illumination adaptive LDA”, International Conference on Pattern Recognition, 2010.
- [24]. Li X. and Areibi S., “A Hardware/Software co-design approach for Face Recognition”, The 16th International Conference on Microelectronics, Tunisia 2004.
- [25]. F. Tivive and A. Bouzerdoum, “A new class of convolutional neural network and their application to face detection”, International Joint Conference on neural Networks, vol. 3, 2003.
- [26]. A. Bouzerdoum, “Classification and function approximation using feed-forward shunting inhibitory artificial neural networks”, vol. 6, 2000.
- [27]. Lin-Lin Huang, Akinobu Shimizu, Yoshihiro Hagihara, Hideo Kobatake, “Face detection from cluttered images using a polynomial neural network”, Elsevier Science 2002.
- [28]. D. Cai, X. He and J. Han, “Efficient Kernel Discriminant Analysis via Spectral Regression”, Technical report, Computer Science Department, UIUC, UIUCDCS, August 2007.

Sneha R. Madane is Master of Engineering (M. E.) Student, studying in B. N. College of Engineering, Pusad, Dist- Yavatmal (India). She has published 4 papers in National conferences. Her area of interest includes Image processing, Data analysis, Mobile Computing.

Prof. S. T. Khandare is professor in Computer department, B. N. College of Engineering, Pusad, Dist- Yavatmal (India). His area of interest includes Image processing, Cloud computing.