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Feature Extraction in Face Recognition using SVM-LBP Detection Technique

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ABSTRACT: Face recognition is a type of biometric software application by using which, we can analyzing, identifying or verifying digital image of the person by using the feature of the face of the person that are unique characteristics of each person. These characteristics may be physical or behavior. The physiological characteristics as like finger print, iris scan, or face etc and behavior characteristics as like hand-writing, voice, key stroke etc. Face recognition is very useful in many areas such as military, airports, universities, ATM, and banks etc, used for the security purposes. There are many techniques or algorithms that are used features extraction in face recognition. In this paper, There are two main categories that may serve as a solution for this problem: feature-based and image-based approaches. In this thesis, three different feature-based oriented solutions are compared. In our, as studied worked on both face detection and recognition methods that is Edge based, Skin fusion based and template matching SVM-LBP method in which we recognize an unknown test image by comparing it with the known training images stored in the database as well as give information regarding the person recognized. These techniques works well under robust conditions like complex background, different face positions. We have taken skin color as a tool for detection. This technique works well for Indian faces which have a specific complexion varying under certain range. We have taken real life examples and simulated the algorithms in MATLAB successfully.

KEYWORDS: - Face recognition, Features extraction, SVM-LBP, feature-based, Image based.

I. INTRODUCTION

The face is our primary focus of attention in social life playing an important role in conveying identity and emotions. We can recognize a number of faces learned throughout our lifespan and identify faces at a glance even after years of separation. This skill is quite robust despite of large variations in visual stimulus due to changing condition, aging and distractions such as beard, glasses or changes in hairstyle.

Computer vision, in general, aims to duplicate (or in some cases compensate) human vision, and traditionally, have been used in performing routine, repetitive tasks, such as classification in massive assembly lines. Today, research on computer vision is spreading enormously so that it is almost impossible to itemize all of its subtopics. Despite of this fact, one can list relevant several applications, such as face processing (i.e. face, expression, and gesture recognition), computer human interaction, crowd surveillance, and content-based image retrieval. All of these applications, stated above, require face detection, which can be simply viewed as a preprocessing step, for obtaining the "object". In other words, many of the techniques that are proposed for these applications assume that the location of the face is preidentified and available for the next step.

Template Matching

Given an input image, the correlation values in predetermined standard regions, such as face contour, eyes, nose and mouth are calculated independently. Although, this approach has the simplicity, it has been insufficient for face detection since it cannot handle variations in scale, rotations pose and shape. Multi-resolution, multi-scale, sub-templates and deformable templates have been proposed to achieve scale and shape invariance template matching [20, 50].



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In [20], Miao et al. proposed a hierarchical template matching method for face detection. Initially, the input image is rotated from -20° to 20° degrees to handle rotation. Then, each rotated image form a mosaic at different scales in which edges are extracted using Laplacian operator. The face template consists of six facial components of two eyebrows, two eyes, nose, and mouth.

Face candidates are located by matching templates of face models represented in edges. In the final step, some heuristics are used to determine existence of a face. Experiments show better detection performance for images containing single face, rather than multiple. Kwon et al. [50] proposed a detection method based on *snakes* and templates. In this approach, an image is first convolved with a blurring filter then with morphological operator to enhance edges. A modified *n-pixel* snake is used to find and eliminate small curve segments. Each candidate is approximated using an ellipse and for each of these candidates, a deformable template method is used to find detailed features. If a sufficient number of facial features are found, and their ratio satisfies the ratio tests based on the template, a face is considered to be detected. Lanitis et al. [3] established a detection method utilizing both shape and intensity information. In this approach, training images are formed in which contours are manually labeled with sampled points, and vector sample points are used as shape feature vectors to be detected. They use a point distribution model (PDM) together with the principal components analysis (PCA) to characterize the shape vectors over an ensemble of individuals. A face shape PDM can be used to detect face in test images using active shape model search to estimate face location and shape parameters. The shape patch is then deformed to the average shape, and intensity parameters are extracted. Then the shape and intensity parameters are used together for measuring Euclidian distance from the face-ness.

Generalized Knowledge Rules

In generalized knowledge-based approaches, the algorithms are developed based on heuristics about face appearance. Although, it is simple to create heuristics for describing the face, the major difficulty is in translating these heuristics into classification rules in an efficient way. If these rules are over detailed, they may come up with missed detections; on the other hand, if they are more general they may introduce much false detection. In spite of this, some heuristics can be used at an acceptable rate in frontal faces existed on uncluttered back-grounds. Yang and Huang [12] used a hierarchical knowledge-based method to detect faces. Their system consists of three level rules going from general to detail. This method does not report a high detection rate, their ideas for mosaicing (multi-resolution), and multiple level rules have been used by more recent methods.

II. LITERATURE SURVEY

P. Dhoke, et al. (2014), the author proposed a method by using Principal Component Analysis (PCA) with Back Propagation Neural Networks (BPNN) method for identification and verification of a person for face recognition system in this paper. The PCA algorithm is used for the reduction of dimensionality of face mage and the recognition is done by the BPNN. The method that is proposed by the author is applied on a database that consists of a set of facial patterns. The PCA method is used to reduce dimensionality of an image and each face image may be represented as a feature vector of the eigen faces, which are stored in a 1D array. By using this feature vector of Eigen faces, the test image can be constructed. The distance between the feature vectors of the test image and that of the database images are then compared. Thus one can reconstruct original image with the help of Eigen faces so that it matches the desired image. The back-propagation algorithm is a multi-layer network and it is a fully feed-forward network connection. There are main two layers in this first is input layer and second is output layer. Each layer is connected with another layer and the activation travels from input layer to output layer. Back-propagation algorithm consists of two sweeps of the network which are the forward sweep and the backward sweeps. Forward sweep defines the network from the input layer to the output layer and the backward sweep defines network from the output layer to the input layer. The author through this paper shows that the PCA for feature extraction and BPNN for image classification and recognition provide the fast computation and high accuracy rate in face recognition system. The execution time of this is only few seconds and acceptance ration s more than 90%. F. Bellakhdhar, et al. (2013), the author conveyed that Face recognition is an important research field that is used in pattern recognition. It is necessary for us to give attention to feature extractor and classifier. Because the performance of face recognition system is depend on how to extract feature vector and to classify them into a class correctly. In this paper, the author propose a methodological improvement to



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raise face recognition rate by fusing the phase and magnitude of Gabor's representations of the face as a new representation, by using the face recognition algorithm, the principal component Analysis approach and Support Vector Machine(SVM) that are used in this paper as a new classifier for pattern recognition. The algorithms that are proposed in by the author are tested on the public and largely used databases of FRGCv2 face and ORL databases. The algorithm PCA is a global method that is used to implement contrasts with a strong sensitivity to changes in lighting, poses and facial expression (by using the number of poses for each person). The author combines the magnitude and the phase of Gabor is used to extract the characteristic vector, the algorithm PCA used for recognition and SVM used to classify faces. Through this approach, it makes the PCA an algorithm effective and commonly used in reducing dimensionality where it can then be used to upstream other algorithms to improve the results of our application. V.Pali, et al. (2013), the author proposes a method of genetic algorithm (GA) that is on based neural network for feature and in this paper, there are four stages or algorithms are used to get the best results. PCA is used for dimensionality reduction that is Eigen face approach; LDA is used for feature extraction. LDA and PCA methods used for dimensionality reduction and feature extraction which overcomes the small sample size (SSS) problem of LDA. Genetic Algorithm based feature selection and last is Back propagation Neural Network (BPNN) is used for the classification of face images to a particular class. The author do this experiment on database of 80 images and a test database of 40 images and achieved a recognition rate of 97.5% with a small execution time of 10 ms to 20 ms. These methods increase the rate of execution of the system. A. Singh, et al. (2012, the author proposed a method that is combination of four face recognition algorithms that are PCA, LDA, ICA and SVM on AT&T and IFD database to study their performance in terms of accuracy, training time, testing time, total execution time and model size. The author gives the comparison of these methods of face recognition. In this paper, the author observed that recognition rate of the ATT database is higher as compare to IFD database, is due to the nature of images contain in the IFD. The four algorithms of face recognition i.e. PCA, LDA, ICA and SVM which are used in this paper gives the result that are, SVM has the highest (95.6%) rate of accuracy of recognition on ATT database while in case of IFD database LDA (86.3%) marginally ahead from SVM (85.4%).

III. METHDOLOGY

Algorithm for face detecting using Template Matching with LBP and SVM

Various Steps of that are used for face detection are as below:

Step 1: Load the image and get the image into RGB color space and process it to convert into Gray scale.

Step 2: Apply filter such as median filter to remove noise from the tested image.

Step 3: Now, Extraction of various parts of the faces such as face, eye, nose, mouth, etc. are detected by using Extract features object detector and places the bounding boxes on the object which is to be detected.

Step 4: Now, apply LBP technique to recognize the face identification for exact matching of face from a tested image with histogram technique is saved in the training dataset that is to be loaded by SVM mean support vector machine. **Step 5:** Tested Image with feature extraction is saved as in above step is loaded for classification to recognize or identify exact face from the database.

IV. RESULTS

In this chapter, experimental results concerning training and evaluation will be introduced. In training experiments, SVM with LBP of the learning methods provide several features enabling a control for learning efficiency. For testing, besides using some different constants, both of the methods share the same testing setup. Here three techniques have been worked out having different results on the base of parameters.

Some of the faces were not detected properly and one non-face area was also detected as face. The experiment was conducted again after coloring the background which provided a 100% result. Also the experiment was conducted on other images. Images having high ranges of color (most of them falling under skin color category) showed false results. After adjusting the filter coefficients and changing the color range, a better result was obtained. The results are shown below with graphical regions and as well as through parameters.



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Implementation using SVM- LBP Face Detection

Graphical representation is shown below using SVM- LBP detecting used for face identifying. In this, various parts of face is also detected. In this, five buttons are shown. First button is used for loading the original image. Second button is used for Gray scale conversion. Third button is used for showing filtered image as filtering through median filter. Fourth button is for extraction of face only. At last, fifth button extract feature shows the various parts of faces to be recognized or identified.

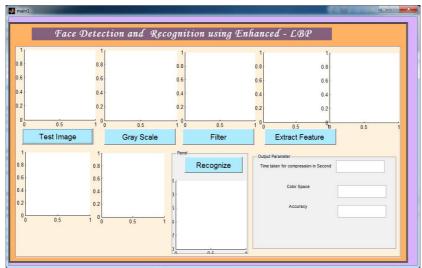


Figure 1.1: Window for detecting face parts using SVM- LBP Technique

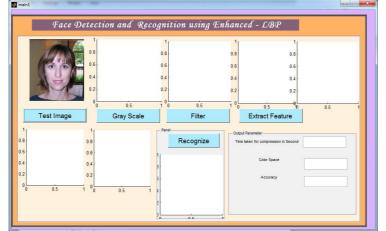


Figure 1.2: Window with Loading Image using SVM- LBP Technique

Above Figure 1.1, shows the original window. Figure 1.2, shows the window with loading or selecting image for dataset.



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Figure 1.3: Window with Gray Scale Image using SVM- LBP Technique

Below Figure 1.3, shows the Gray Scale image. Figure 1.4, shows the window with Filtered image after removing noise form it.

nain1 Face Det	ection and Reco	gnition using Enl	hanced - LBP	
Test Image	Gray Scale	Filter	08 05 06 05 04 04 02 02 00 05 00 05	1
1 0.8 0.6 0.4 0.2 0.2	6 4	Panel	Output Parameter Time taken for compression in Second Color Space Accuracy	
0 0.5 1	0 0.5 1			

Figure 1.4: Window with Filtered Image using SVM- LBP Technique



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Figure 1.5: Window showing various parts of faces using SVM-LBP technique

Figure 1.5, shows the window showing all parts of face as detected in first row with face and nose. Second row with image detecting as mouth and eyes.



Figure 1.6: Window recognizing exact faces using SVM-LBP technique

Discussion

As above shown, graphical results of three techniques. Now, parameters such as true color space, time and accuracy will evaluate the exact result that which technique is best for detecting face from an image and also their parts. Table below shows the comparison of techniques through various parameters.



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Figure 1.7: Original Images d Dataset

Figure 1.7, shows the original images dataset for detection of face appearance as feature extraction.

Table 1.1: Performance comparison of techniques through parameters										
Sr. No	Color space				Accuracy		Time			
Images Name/ Technique	Edge	Fusion Skin	SVM- LBP	Edge	Fusion Skin	SVM- LBP	Edge	Fusion Skin	SVM- LBP	
Image 1	<mark>972.9</mark>	49152.7	26244	79.01	89.82	<mark>91.63</mark>	<mark>0.46</mark>	1.818	5.43	
Image 2	<mark>882.9</mark>	49152.7	23373	87.07	89.96	<mark>91.28</mark>	<mark>0.33</mark>	1.810	3.23	
Image 3	<mark>1059.4</mark>	49152.3	56656	47.85	89.95	<mark>90.87</mark>	<mark>0.44</mark>	1.816	3.05	
Image 4	<mark>954.8</mark>	49152.1	24040	80.50	89.17	<mark>91.17</mark>	<mark>0.34</mark>	1.848	3.27	
Image 7	<mark>870.1</mark>	49152.7	38280	58.18	89.68	<mark>91.02</mark>	<mark>0.42</mark>	1.880	3.22	

Table 1.1: Performance comparison of techniques through parameters

This thesis is mainly focused on face detection using SVM-LBP, Edge and fusion skin methods which simply take face detection as a pattern recognition problem. Prior face knowledge and any other knowledge-based heuristics are ignored and a prediction is only made according to the learned characteristics from training images. Additionally, theoretic and common practical adaptations of learning field for face detection task are studied and main steps for a face detection learning scheme are outlined ranging from training through testing.

In order to make a comprehensive study, Edge, SVM-LBP and Fusion Skin based face detection systems are implemented, and experimented in training and evaluation phases with a comparative insight. The first Edge detection techniques, shows the better results according to the performance of parameters such as time taken and color space. Second SVM-LBP techniques, shows best results in accuracy as high. Color space means the different colors form representation of RGB as numerically. It specifies the color percentages of red, green, and blue hues mixed together. It should be Low as to have accurate color specification. Time compared to be taken less is shown by edge technique. Accuracy means what is true percentage for accurate to detect objects which referred.

V. CONCLUSION

Experimental results in the training phase showed that both Edge and SVM- LBP face classifiers are able separate examples included in face and non-face training data from each other. Thus, both applied an implicit surface division with the help of adjustments for internal weights. Experiments concerning efficiency of these divisions, showed trained tested images have similar generalization (true detection rate over total number of false detections) abilities. While Accuracy in generalization value of SVM-LBP is slightly higher than other techniques as counterpart. Hence, it is possible to state SVM-LBP and Edge detection techniques is best than Fusion skin technique.

The face recognition and detection algorithms were thoroughly studied taking a number of test images and varying the conditions and variables. All the work mentioned above involved real time data. The Edge and SVM-LBP success rates were given while for face detection, the success rate was different for different images depending on the external factors. The overall success rate was 91% with edge detection method.



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VI. FUTURE SCOPE

There exist several directions for researches in the future. Further improvements can be made in terms of training examples for achieving a better generalization. Additionally, more informative features can be used instead of intensity values. For example, a simple quantization for gray levels may also achieve better evaluation results. Moreover, utilization of color values instead of gray levels may also improve learning mechanism, but in that case lighting variations must be sufficient large in the training sets. Internal dynamics may be further studied in order to achieve improvements, and to design different adaptations for other visual tasks.

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